PENELLOPE – A Code System for Monte Carlo Simulation of Electron and Photon Transport

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PENELOPE

A Code System for Monte Carlo Simulation of Electron and Photon Transport

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FOREWORD

The OECD/NEA Data Bank was established in order to promote the effective sharing of data and software developed in Member countries in the field of nuclear technology and radiation physics applications. It operates a Computer Program Service (CPS) related to nuclear energy applications. The software library collects programs, compiles and verifies them in an appropriate computer environment, ensuring that the computer program package is complete and adequately documented. Internationally agreed quality assurance methods are used in the verification process.

In order to obtain good results in modelling the behaviour of technological systems two conditions must be fulfilled:

1. Good quality and validated computer codes and associated basic data libraries should be used.

2. Modelling should be done by a qualified user of such codes.

One area to which special effort has been devoted in recent years is radiation transport. Workshops and training courses including the use of computer codes have been organised in the field of neutral particle transport for codes using both deterministic and stochastic methods. The area of charged particle transport, and in particular electron-photon transport, has received increased attention for a number of technological and medical applications. At the most recent Monte Carlo 2000 Conference, held on 23-26 October 2000 in Lisbon, Portugal, about half of the papers covered electron-photon transport and its application.

A new computer code was recently released to the NEA Data Bank for general distribution: “PENELOPE, A Code System for Monte Carlo Simulation of Electron and Photon Transport” developed by Francesc Salvat, José M. Fernández-Varea, Eduardo Acosta and Josep Sempau. This code began to be widely used by radiation physicists, who requested that a workshop with hands-on training with the PENELOPE code be organised. The NEA Nuclear Science Committee endorsed this request and the authors agreed to teach a course covering the physics behind the code and to demonstrate, with corresponding exercises, how it can be used for practical applications.

These proceedings contain the teaching notes of the workshop and training course held in November 2001.
Abstract

The computer code system PENELOPE (version 2001) performs Monte Carlo simulation of coupled electron-photon transport in arbitrary materials for a wide energy range, from a few hundred eV to about 1 GeV. Photon transport is simulated by means of the standard, detailed simulation scheme. Electron and positron histories are generated on the basis of a mixed procedure, which combines detailed simulation of hard events with condensed simulation of soft interactions. A geometry package called PENGEOM permits the generation of random electron-photon showers in material systems consisting of homogeneous bodies limited by quadric surfaces, i.e. planes, spheres, cylinders, etc. This report is intended not only to serve as a manual of the PENELOPE code system, but also to provide the user with the necessary information to understand the details of the Monte Carlo algorithm.

Keywords: Radiation transport, electron-photon showers, Monte Carlo simulation, sampling algorithms, quadric geometry

Symbols and numerical values of constants frequently used in the text
(Mohr and Taylor, 2000)

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<thead>
<tr>
<th>Quantity</th>
<th>Symbol</th>
<th>Value</th>
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<tr>
<td>Avogadro’s number</td>
<td>(N_A)</td>
<td>(6.022142 \times 10^{23}) mol(^{-1})</td>
</tr>
<tr>
<td>Velocity of light in vacuum</td>
<td>(c)</td>
<td>(2.997925 \times 10^8) m s(^{-1})</td>
</tr>
<tr>
<td>Reduced Planck’s constant</td>
<td>(\hbar = \hbar/(2\pi))</td>
<td>(6.582119 \times 10^{-22}) MeV s</td>
</tr>
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<td>Electron charge</td>
<td>(e)</td>
<td>(1.602176 \times 10^{-19}) C</td>
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<tr>
<td>Electron mass</td>
<td>(m_e)</td>
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<tr>
<td>Electron rest energy</td>
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<td>Classical electron radius</td>
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<td>(\alpha = e^2/\left(h c\right))</td>
<td>(1/137.0360)</td>
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<tr>
<td>Bohr radius</td>
<td>(a_0 = \hbar^2/\left(m_e e^2\right))</td>
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Radiation transport in matter has been a subject of intense work since the beginning of the 20th century. Today, we know that high-energy photons, electrons and positrons penetrating matter suffer multiple interactions by which energy is transferred to the atoms and molecules of the material and secondary particles are produced\(^1\). By repeated interaction with the medium, a high-energy particle originates a cascade of particles which is usually referred to as a shower. After each interaction of a particle, its energy is reduced and further particles may be generated so that the evolution of the shower represents an effective degradation in energy. As time goes on, the initial energy is progressively deposited into the medium, while that remaining is shared by an increasingly larger number of particles.

A reliable description of shower evolution is required in a number of fields. Thus, knowledge of radiation transport properties is needed for quantitative analysis in surface electron spectroscopies (Jablonski, 1987; Tofterup, 1986), positron surface spectroscopy (Shultz and Lynn, 1988), electron microscopy (Reimer, 1985), electron energy loss spectroscopy (Reimer, et al., 1992), electron probe microanalysis (Heinrich and Newbury, 1991), etc. Detailed information on shower evolution is also required for the design and quantitative use of radiation detectors (Titus, 1970; Berger and Seltzer, 1972). A field where radiation transport studies play an important sociological role is that of radiation dosimetry and radiotherapy (Andreo, 1991).

The study of radiation transport problems was initially attempted on the basis of the Boltzmann transport equation. However, this procedure comes up against considerable difficulties when applied to limited geometries, with the result that numerical methods based on the transport equation have only had a certain success in simple geometries, mainly for unlimited and semi-infinite media (see e.g. Zheng-Ming and Brahme, 1993). At the end of the 1950s, with the availability of computers, Monte Carlo simulation methods were developed as a powerful alternative to deal with transport problems. Basically, the evolution of an electron-photon shower is of a random nature, so that this is a process particularly amenable to Monte Carlo simulation. Detailed simulation, where all the interactions experienced by a particle are simulated in chronological succession, is exact, i.e. it yields the same results as the rigorous solution of the transport equation (apart from the inherent statistical uncertainties).

To our knowledge, the first numerical Monte Carlo simulation of photon transport is that of Hayward and Hubbell (1954) who generated 67 photon histories using a desk calculator. The simulation of photon transport is straightforward since the mean number of events in each history is fairly small. Indeed, the photon is effectively absorbed after a single photoelectric or pair-production interaction or after a few Compton interactions (say, of the order of 10). With present-day computational facilities, detailed simulation of photon transport is a simple routine task.

The simulation of electron and positron transport is much more difficult than that of photons. The main reason is that the average energy loss of an electron in a single interaction is very small (of the order of a few tens of eV). As a consequence, high-energy electrons suffer a large number of interactions before being effectively absorbed in the medium. In practice, detailed simulation is feasible

\(^1\) In this report, the term particle will be used to designate either photons, electrons or positrons.
only when the average number of collisions per track is not too large (say, up to a few hundred). Experimental situations which are amenable to detailed simulation are those involving either electron sources with low initial kinetic energies (up to about 100 keV) or special geometries such as electron beams impinging on thin foils. For larger initial energies, and thick geometries, the average number of collisions experienced by an electron until it is effectively stopped becomes very large, and detailed simulation is very inefficient.

For high-energy electrons and positrons, most of the Monte Carlo codes currently available [e.g. ETRAN (Berger and Seltzer, 1988), ITS3 (Halbleib, et al., 1992), EGS4 (Nelson, et al., 1985), EGSnrc (Kawrakow and Rogers, 2000), GEANT (Brun, et al., 1986), MCNP (Briesmeister, 1997), …] have recourse to multiple scattering theories which allow the simulation of the global effect of a large number of events in a track segment of a given length (step). Following Berger (1963), these simulation procedures will be referred to as “condensed” Monte Carlo methods. The multiple scattering theories implemented in condensed simulation algorithms are only approximate and may lead to systematic errors, which can be made evident by the dependence of the simulation results on the adopted step length (Bielajew and Rogers, 1987). To analyse their magnitude, one can perform simulations of the same arrangement with different step lengths. The results are usually found to stabilise when the step length is reduced, while computation time increases rapidly, roughly in proportion to the inverse of the step length. Thus, for each particular problem, one must reach a certain compromise between available computer time and attainable accuracy. It is also worth noting that, owing to the nature of certain multiple scattering theories and/or to the particular way they are implemented in the simulation code, the use of very short step lengths may introduce spurious effects in the simulation results. For instance, the multiple elastic scattering theory of Molière (1948) which is the model used in EGS4-based codes, is not applicable to step lengths shorter than a few times the elastic mean free path (see e.g. Fernández-Varea, et al., 1993b) and multiple elastic scattering has to be switched off when the step length becomes smaller than this value. As a consequence, stabilisation for short step lengths does not necessarily imply that simulation results are correct. Condensed schemes also have difficulties in generating particle tracks in the vicinity of an interface, i.e. a surface separating two media of different compositions. When the particle moves near an interface, the step length must be kept smaller than the minimum distance to the interface so as to make sure that the step is completely contained in the initial medium (Bielajew and Rogers, 1987). This may complicate the code considerably, even for relatively simple geometries.

In the present report, we describe the version 2001 of PENELOPE, a Monte Carlo algorithm and computer code for the simulation of coupled electron-photon transport. The name is an acronym that stands for PENetration and Energy LOss of Positrons and Electrons (photon simulation was introduced later). The simulation algorithm is based on a scattering model that combines numerical databases with analytical cross section models for the different interaction mechanisms and is applicable to energies (kinetic energies in the case of electrons and positrons) from a few hundred eV to ~1 GeV. Photon transport is simulated by means of the conventional detailed method. The simulation of electron and positron transport is performed by means of a mixed procedure. Hard interactions, with scattering angle $\theta$ and/or energy loss $W$ greater than pre-selected cut-off values $\theta_c$ and $W_c$, are simulated in detail. Soft interactions, with scattering angle or energy loss less than the corresponding cut-offs, are described by means of multiple scattering approaches. This simulation scheme handles lateral displacements and interface crossing appropriately and provides a consistent description of energy straggling. The simulation is stable under variations of the cut-offs $\theta_c$, $W_c$ and these can be made quite large, thus speeding up the calculation considerably, without altering the results. A characteristic feature of our code is that the most delicate parts of the simulation are handled internally; electrons, positrons and photons are simulated by calling the same subroutines. Thus, from the user’s point of view, PENELOPE makes the practical simulation of electrons and positrons as simple as that of photons (although simulating a charged particle may take a longer time).
The present version of PENELOPE is the result of continued evolution from the first version, which was released in 1996. The idea of developing a general-purpose Monte Carlo code, with better modelling than those available at that time, arose during a short course on radiation transport simulation given by F. Salvat at the Radiation Metrology Unit, CIEMAT (Madrid), in 1988. The present version 2001 contains substantial changes/improvements to the previous versions 1996 and 2000. As for the physics, the model for electron/positron elastic scattering has been revised, bremsstrahlung emission is now simulated using partial-wave data instead of analytical approximate formulae, photoelectric absorption in K and L-shells is described from the corresponding partial cross sections, and fluorescence radiation from vacancies in K and L-shells is now followed. Refinements have also been introduced in the electron/positron transport mechanics, mostly to account for the energy dependence of the mean free paths for hard events. The simulation routines have been re-programmed in a more structured (and readable) way and new example MAIN programs have been written, with a more flexible input and expanded output.

This report is intended not only to serve as a manual of the simulation package, but also to provide the user with the necessary information to understand the details of the Monte Carlo algorithm. In Chapter 1 we give a brief survey of random sampling methods and an elementary introduction to Monte Carlo simulation of radiation transport. The cross sections adopted in PENELOPE to describe particle interactions, and the associated sampling techniques, are presented in Chapters 2 and 3. Chapter 4 is devoted to mixed simulation methods for electron and positron transport. In Chapter 5, a relatively simple, but effective, method to handle simulation in quadric geometries is presented. The FORTRAN77 simulation package PENELOPE and other complementary programs, are described in Chapter 6, which also provides instructions to operate them. Information on relativistic kinematics and numerical methods is given in Appendices A and B. Finally, Appendix C is devoted to simulation of electron/positron transport under external, static electric and magnetic fields. The source files of PENELOPE, the auxiliary programs and the database are supplied on a ZIP-compressed file, which is distributed by the NEA Data Bank and the RSICC. The code is also available from the authors, but we would appreciate it if users did try to get the code from these institutions.

In the course of our Monte Carlo research, we have had the fortune of getting much help from numerous friends and colleagues. Since the mid 1980s, we have benefited from discussions with D. Liljequist, which gave shape to our first algorithm for simulation of electrons and positrons. We are particularly grateful to A. Riveros for his enthusiastic and friendly support over the years, and for guiding us into the field of microanalysis and X-ray simulation. A. Sánchez-Reyes and E. García-Toraño were the first external users of the code system; they suffered the inconveniences of using continuously changing preliminary versions of the code without complaining too much. More recently, stimulating collaboration with A.F. Bielajew has led to substantial improvements in the electron transport mechanics and in the code organisation. We are deeply indebted to J.H. Hubbell and D.E. Cullen for kindly providing us with updated information on photon interaction and atomic relaxation data. Thanks are also due to S.M. Seltzer for sending us his bremsstrahlung energy-loss database. L. Sorbier generously prepared most of the photoelectric and atomic relaxation database files and worked on the associated sampling algorithms. We are indebted to many colleagues, especially P. Andreo, for comments and suggestions, which have been of much help to improve the present version of the code. Our most sincere appreciation to the members of our research group; X. Llovet, M. Dingfelder,

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J. Asenjo and C. Campos. They did much more than chasing bugs through the programs and in this write-up. Finally, we would like to thank the staff of the NEA Data Bank, particularly E. Sartori, for kindly organising the first training course on PENEOPE.

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Barcelona, November 2001
Chapter 1

Monte Carlo simulation. Basic concepts

The name “Monte Carlo” was coined in the 1940s by scientists working on the nuclear weapon project in Los Alamos to designate a class of numerical methods based on the use of random numbers. Nowadays, Monte Carlo methods are widely used to solve complex physical and mathematical problems (James, 1980; Rubinstein, 1981; Kalos and Whitlock, 1986), particularly those involving multiple independent variables where more conventional numerical methods would demand formidable amounts of memory and computer time. The book by Kalos and Whitlock (1986) gives a readable survey of Monte Carlo techniques, including simple applications in radiation transport, statistical physics, and many-body quantum theory.

In Monte Carlo simulation of radiation transport, the history (track) of a particle is viewed as a random sequence of free flights that end with an interaction event where the particle changes its direction of movement, loses energy and, occasionally, produces secondary particles. The Monte Carlo simulation of a given experimental arrangement (e.g. an electron beam, coming from an accelerator and impinging on a water phantom) consists of the numerical generation of random histories. To simulate these histories we need an “interaction model”, i.e. a set of differential cross sections (DCS) for the relevant interaction mechanisms. The DCSs determine the probability distribution functions (PDF) of the random variables that characterize a track; 1) free path between successive interaction events, 2) kind of interaction taking place and 3) energy loss and angular deflection in a particular event (and initial state of emitted secondary particles, if any). Once these PDFs are known, random histories can be generated by using appropriate sampling methods. If the number of generated histories is large enough, quantitative information on the transport process may be obtained by simply averaging over the simulated histories.

The Monte Carlo method yields the same information as the solution of the Boltzmann transport equation, with the same interaction model, but is easier to implement (Berger, 1963). In particular, the simulation of radiation transport in finite samples is
straightforward, while even the simplest finite geometries (e.g. thin foils) are very difficult to be dealt with by the transport equation. The main drawback of the Monte Carlo method lies in its random nature, all the results are affected by statistical uncertainties, which can be reduced at the expense of increasing the sampled population and, hence, the computation time. Under special circumstances, the statistical uncertainties may be lowered by using variance-reduction techniques (Rubinstein, 1981; Bielajew and Rogers, 1988).

1.1 Elements of probability theory

The essential characteristic of Monte Carlo simulation is the use of random numbers and random variables. A random variable is a quantity that results from a repeatable process and whose actual values (realizations) cannot be predicted with certainty. In the real world, randomness originates either from uncontrolled factors (as occurs e.g. in games of chance) or from the quantum nature of microscopic systems and processes (e.g. nuclear disintegration and radiation interactions). As a familiar example, assume that we throw two dice in a box; the sum of points in their upper faces is a discrete random variable, which can take the values 2 to 12, while the distance \( x \) between the dice is a continuous random variable, which varies between zero (dice in contact) and a maximum value determined by the dimensions of the box. In the computer, random variables are generated by means of numerical transformations of random numbers (see below).

Let \( x \) be a continuous random variable that takes values in the interval \( x_{\text{min}} \leq x \leq x_{\text{max}} \). To measure the likelihood of obtaining \( x \) in an interval \((a,b)\) we use the probability \( P\{x|a < x < b\} \), defined as the ratio \( n/N \) of the number \( n \) of values of \( x \) that fall within that interval and the total number \( N \) of generated \( x \)-values, in the limit \( N \to \infty \). The probability of obtaining \( x \) in a differential interval of length \( dx \) about \( x_1 \) can be expressed as

\[
P\{x|x_1 < x < x_1 + dx\} = p(x_1) \, dx,
\]

where \( p(x) \) is the probability distribution function (PDF) of \( x \). Since 1) negative probabilities have no meaning and 2) the obtained value of \( x \) must be somewhere in \( (x_{\text{min}},x_{\text{max}}) \), the PDF must be definite positive and normalized to unity

\[
p(x) \geq 0 \quad \text{and} \quad \int_{x_{\text{min}}}^{x_{\text{max}}} p(x) \, dx = 1.
\]

Any “function” that satisfies these two conditions can be interpreted as a PDF. In Monte Carlo simulation we shall frequently use the uniform distribution,

\[
U_{x_{\text{min}},x_{\text{max}}}(x) = \begin{cases} 
1/(x_{\text{max}} - x_{\text{min}}) & \text{if } x_{\text{min}} < x < x_{\text{max}}, \\
0 & \text{otherwise,}
\end{cases}
\]
1.1. Elements of probability theory

which is discontinuous. The definition (1.2) also includes singular distributions such as the Dirac delta, \( \delta(x - x_0) \), which is defined by the property

\[
\int_a^b f(x) \delta(x - x_0) \, dx = \begin{cases} 
  f(x_0) & \text{if } a < x_0 < b, \\
  0 & \text{if } x_0 < a \text{ or } x_0 > b
\end{cases}
\]  

(1.4)

for any function \( f(x) \) that is continuous at \( x_0 \). An equivalent, more intuitive definition is the following,

\[
\delta(x - x_0) \equiv \lim_{\Delta \to 0} U_{x_0 - \Delta, x_0 + \Delta}(x),
\]  

(1.4')

which represents the delta distribution as the zero-width limit of a sequence of uniform distributions centred at the point \( x_0 \). Hence, the Dirac distribution describes a single-valued discrete random variable (i.e., a constant). The PDF of a random variable \( x \) that takes the discrete values \( x = x_1, x_2, \ldots \) with point probabilities \( p_1, p_2, \ldots \) can be expressed as a mixture of delta distributions,

\[
p(x) = \sum_i p_i \delta(x - x_i).
\]  

(1.5)

Discrete distributions can thus be regarded as particular forms of continuous distributions.

Given a continuous random variable \( x \), the cumulative distribution function of \( x \) is defined by

\[
\mathcal{P}(x) \equiv \int_{x_{\text{min}}}^x p(x') \, dx'.
\]  

(1.6)

This is a non-decreasing function of \( x \) that varies from \( \mathcal{P}(x_{\text{min}}) = 0 \) to \( \mathcal{P}(x_{\text{max}}) = 1 \). In the case of a discrete PDF of the form (1.5), \( \mathcal{P}(x) \) is a step function. Notice that the probability \( P\{x|a < x < b\} \) of having \( x \) in the interval \((a,b)\) is

\[
P\{x|a < x < b\} = \int_a^b p(x) \, dx = \mathcal{P}(b) - \mathcal{P}(a),
\]  

(1.7)

and that \( p(x) = d\mathcal{P}(x)/dx \).

The \( n \)-th moment of \( p(x) \) is defined as

\[
\langle x^n \rangle = \int_{x_{\text{min}}}^{x_{\text{max}}} x^n p(x) \, dx.
\]  

(1.8)

The moment \( \langle x^n \rangle \) is simply the integral of \( p(x) \), which is equal to unity, by definition. However, higher order moments may or may not exist. An example of a PDF that has no even-order moments is the Lorentz or Cauchy distribution,

\[
p_L(x) \equiv \frac{1}{\pi} \frac{\gamma}{\gamma^2 + x^2}, \quad -\infty < x < \infty.
\]  

(1.9)

Its first moment, and other odd-order moments, can be assigned a finite value if they are defined as the “principal value” of the integrals, e.g.

\[
\langle x \rangle_L = \lim_{a \to \infty} \int_{-a}^{+a} x \frac{1}{\pi} \frac{\gamma}{\gamma^2 + x^2} \, dx = 0,
\]  

(1.10)
but the second and higher even-order moments are infinite, irrespective of the way they are defined.

The first moment, when it exists, is called the mean or expected value of the random variable \( x \),

\[
\langle x \rangle = \int x \ p(x) \, dx. \tag{1.11}
\]

The expected value of a function \( f(x) \) is defined in a similar way,

\[
\langle f(x) \rangle \equiv \int f(x) \ p(x) \, dx. \tag{1.12}
\]

Since \( f(x) \) is a random variable, it has its own PDF, \( \pi(f) \), which is such that the probability of having \( f \) in a certain interval of length \( df \) is equal to the probability of having \( x \) in the corresponding interval or intervals\(^1\). Thus, if \( f(x) \) is a monotonously increasing function of \( x \) (so that there is a one-to-one correspondence between the values of \( x \) and \( f \)), \( p(x) \, dx = \pi(f) \, df \) and

\[
\pi(f) = p(x) \cdot (df/dx)^{-1}. \tag{1.13}
\]

It can be shown that the definitions (1.11) and (1.12) are equivalent. If \( f(x) \) increases monotonously with \( x \), the proof is trivial: we can start from the definition (1.11) and write

\[
\langle f \rangle = \int f \ \pi(f) \, df = \int f(x) \ p(x) \cdot (dx/df) \, df = \int f(x) \ p(x) \, dx,
\]

which agrees with (1.12). Notice that the expectation value is linear, i.e.

\[
\langle a_1 f_1(x) + a_2 f_2(x) \rangle = a_1 \langle f_1(x) \rangle + a_2 \langle f_2(x) \rangle, \tag{1.14}
\]

where \( a_1 \) and \( a_2 \) are arbitrary real constants.

If the first and second moments of the PDF \( p(x) \) exist, we define the variance of \( x \) [or of \( p(x) \)] by

\[
\text{var}(x) \equiv \langle (x - \langle x \rangle)^2 \rangle = \int (x - \langle x \rangle)^2 \ p(x) \, dx = \langle x^2 \rangle - \langle x \rangle^2. \tag{1.15}
\]

The square root of the variance, \( \sigma \equiv [\text{var}(x)]^{1/2} \), is called the “standard deviation” (and sometimes the “standard uncertainty”); it gives a measure of the dispersion of the random variable (i.e. of the width of the PDF). The Dirac delta is the only PDF that has zero variance. Similarly, the variance of a function \( f(x) \) is defined as

\[
\text{var}\{f(x)\} = \langle f^2(x) \rangle - \langle f(x) \rangle^2. \tag{1.16}
\]

Thus, for a constant \( f(x) = a \), \( \langle f \rangle = a \) and \( \text{var}\{f\} = 0. \)

---

\(^1\)When \( f(x) \) does not increase or decrease monotonously with \( x \), there may be multiple values of \( x \) corresponding to a given value of \( f \).
1.1.1 Two-dimensional random variables

Let us now consider the case of a two-dimensional random variable, \((x, y)\). The corresponding (joint) PDF \(p(x, y)\) satisfies the conditions

\[
p(x, y) \geq 0 \quad \text{and} \quad \int dx \int dy \ p(x, y) = 1.
\]  

(1.17)

The marginal PDFs of \(x\) and \(y\) are defined as

\[
q(x) \equiv \int p(x, y) dy \quad \text{and} \quad q(y) \equiv \int p(x, y) dx,
\]

(1.18)

i.e. \(q(x)\) is the probability of obtaining the value \(x\) and any value of \(y\). The joint PDF can be expressed as

\[
p(x, y) = q(x) p(y|x) = q(y) p(x|y),
\]

(1.19)

where

\[
p(x|y) = \frac{p(x, y)}{q(y)} \quad \text{and} \quad p(y|x) = \frac{p(x, y)}{q(x)}
\]

(1.20)

are the conditional PDFs of \(x\) and \(y\), respectively. Notice that \(p(x|y)\) is the normalized PDF of \(x\) for a fixed value of \(y\).

The expectation value of a function \(f(x, y)\) is

\[
\langle f(x, y) \rangle = \int dx \int dy \ f(x, y) \ p(x, y).
\]

(1.21)

The moments of the PDF are defined by

\[
\langle x^n y^m \rangle = \int dx \int dy \ x^n y^m \ p(x, y).
\]

(1.22)

In particular,

\[
\langle x^n \rangle = \int dx \int dy \ x^n \ p(x, y) = \int x^n \ q(x) \ dx.
\]

(1.23)

Again, the only moment that is necessarily defined is \(\langle x^0 y^0 \rangle = 1\). When the corresponding moments exist, the variances of \(x\) and \(y\) are given by

\[
\text{var}(x) = \langle x^2 \rangle - \langle x \rangle^2 \quad \text{and} \quad \text{var}(y) = \langle y^2 \rangle - \langle y \rangle^2.
\]

(1.24)

The variance of \(x + y\) is

\[
\text{var}(x + y) = \langle (x + y)^2 \rangle - \langle x + y \rangle^2 = \text{var}(x) + \text{var}(y) + 2 \text{cov}(x, y),
\]

(1.25)

where

\[
\text{cov}(x, y) = \langle xy \rangle - \langle x \rangle \langle y \rangle
\]

(1.26)

is the covariance of \(x\) and \(y\), which can be positive or negative. A related quantity is the correlation coefficient,

\[
\rho(x, y) = \frac{\text{cov}(x, y)}{\sqrt{\text{var}(x) \text{var}(y)}},
\]

(1.27)
which takes values from $-1$ to $1$. Notice that $\text{cov}(x, x) = \text{var}(x)$. When the variables $x$ and $y$ are independent, i.e. when $p(x, y) = p_x(x)p_y(y)$, we have

$$\text{cov}(x, y) = 0 \quad \text{and} \quad \text{var}(x + y) = \text{var}(x) + \text{var}(y). \quad (1.28)$$

Moreover, for independent variables,

$$\text{var}\{a_1 x + a_2 y\} = a_1^2 \text{var}(x) + a_2^2 \text{var}(y). \quad (1.29)$$

### 1.2 Random sampling methods

The first component of a Monte Carlo calculation is the numerical sampling of random variables with specified PDFs. In this section we describe different techniques to generate random values of a variable $x$ distributed in the interval $(x_{\text{min}}, x_{\text{max}})$ according to a given PDF $p(x)$. We concentrate on the simple case of single-variable distributions, since random sampling from multivariate distributions can always be reduced to single-variable sampling (see below). A more detailed description of sampling methods can be found in the textbooks of Rubinstein (1981) and Kalos and Whitlock (1986).

#### 1.2.1 Random number generator

In general, random sampling algorithms are based on the use of random numbers $\xi$ uniformly distributed in the interval $(0,1)$. These random numbers can be easily generated on the computer (see e.g. Kalos and Whitlock, 1986; James, 1990). Among the “good” random number generators currently available, the simplest ones are the so-called multiplicative congruential generators (Press and Teukolsky, 1992). A popular example of this kind of generator is the following,

$$R_n = 7^5 R_{n-1} \bmod 2^{31} - 1, \quad \xi_n = R_n/(2^{31} - 1), \quad (1.30)$$

which produces a sequence of random numbers $\xi_n$ uniformly distributed in $(0,1)$ from a given “seed” $R_0 (< 2^{31} - 1)$. Actually, the generated sequence is not truly random, since it is obtained from a deterministic algorithm (the term “pseudo-random” would be more appropriate), but it is very unlikely that the subtle correlations between the values in the sequence have an appreciable effect on the simulation results. The generator (1.30) is known to have good random properties (Press and Teukolsky, 1992). However, the sequence is periodic, with a period of the order of $10^9$. With present-day computational facilities, this value is not large enough to prevent re-initiation in a single simulation run. An excellent critical review of random number generators has been published by James (1990), where he recommends using algorithms that are more sophisticated than simple congruential ones. The generator implemented in the FORTRAN 77 function \texttt{RAND} (table 1.1) is due to L’Ecuyer (1988); it produces 32-bit floating point numbers uniformly distributed in the open interval between zero and one. Its period is of the order of $10^{18}$, which is virtually infinite for practical simulations.
Table 1.1: FORTRAN77 random number generator.

| C | ********************************************************* |
| C | FUNCTION RAND                                           |
| C | ********************************************************* |
| C | FUNCTION RAND(DUMMY)                                    |
| C | This is an adapted version of subroutine RANECU written by F. James |
| C | (Comput. Phys. Commun. 60 (1990) 329–344), which has been modified to |
| C | give a single random number at each call.                |
| C | The 'seeds' ISEED1 and ISEED2 must be initialized in the main program |
| C | and transferred through the named common block /RSEED/.   |
| C | IMPLICIT DOUBLE PRECISION (A-H,O-Z), INTEGER*4 (I)       |
| C | PARAMETER (USCALE=1.0D0/2.0D0**31)                      |
| C | COMMON/RSEED/ISEED1, ISEED2                             |
| C | I1=ISEED1/53668                                         |
| C | ISEED1=40014*(ISEED1-I1*53668)-I1*12211                 |
| C | IF(ISEED1.LT.0) ISEED1=ISEED1+2147483563                  |
| C | I2=ISEED2/52774                                         |
| C | ISEED2=40692*(ISEED2-I2*52774)-I2*3791                 |
| C | IF(ISEED2.LT.0) ISEED2=ISEED2+2147483399                  |
| C | IZ=ISEED1-ISEED2                                        |
| C | IF(IZ.LT.1) IZ=IZ+2147483562                            |
| C | RAND=IZ*USCALE                                         |
| C | RETURN                                                  |
| C | END                                                     |

1.2.2 Inverse transform method

The cumulative distribution function of \( p(x) \), eq. (1.6), is a non-decreasing function of \( x \) and, therefore, it has an inverse function \( P^{-1}(\xi) \). The transformation \( \xi = P(x) \) defines a new random variable that takes values in the interval \((0,1)\), see fig. 1.1. Owing to the correspondence between \( x \) and \( \xi \) values, the PDF of \( \xi \), \( p_\xi(\xi) \), and that of \( x \), \( p(x) \), are related by

\[
p_\xi(\xi) = p(x) \left( \frac{d\xi}{dx} \right)^{-1} = p(x) \left( \frac{dP(x)}{dx} \right)^{-1} = 1, \tag{1.31}
\]

that is, \( \xi \) is distributed uniformly in the interval \((0,1)\).

Now it is clear that if \( \xi \) is a random number, the variable \( x \) defined by \( x = P^{-1}(\xi) \) is randomly distributed in the interval \((x_{\text{min}}, x_{\text{max}})\) with PDF \( p(x) \) (see fig. 1.1). This provides a practical method of generating random values of \( x \) using a generator of random numbers uniformly distributed in \((0,1)\). The randomness of \( x \) is guaranteed by
that of ξ. Notice that \( x \) is the (unique) root of the equation
\[
ξ = \int_{x_{\text{min}}}^{x} p(x') \, dx',
\]
which will be referred to as the sampling equation of the variable \( x \). This procedure for random sampling is known as the inverse transform method; it is particularly adequate for PDFs \( p(x) \) given by simple analytical expressions such that the sampling equation (1.32) can be solved analytically.

![Diagram](image)

**Figure 1.1:** Random sampling from a distribution \( p(x) \) using the inverse transform method.

Consider, for instance, the uniform distribution in the interval \((a, b)\),
\[
p(x) \equiv U_{a,b}(x) = \frac{1}{b - a}.
\]
The sampling equation (1.32) then reads
\[
ξ = \frac{x - a}{b - a},
\]
which leads to the well-known sampling formula
\[
x = a + ξ(b - a).
\]
As another familiar example, consider the exponential distribution
\[
p(s) = \frac{1}{λ} \exp(-s/λ), \quad s > 0,
\]
of the free path \( s \) of a particle between interaction events (see section 1.4.1). The parameter \( λ \) represents the mean free path. In this case, the sampling equation (1.32) is easily solved to give the sampling formula
\[
s = -λ \ln(1 - ξ) = -λ \ln ξ,
\]
The last equality follows from the fact that \( 1 - ξ \) is also a random number distributed in \((0, 1)\).
1.2. Random sampling methods

Numerical inverse transform

The inverse transform method can also be efficiently used for random sampling from continuous distributions \( p(x) \) that are given in numerical form, or that are too complicated to be sampled analytically. To apply this method, the cumulative distribution function \( \mathcal{P}(x) \) has to be evaluated at the points \( x_i \) of a certain grid. The sampling equation \( \mathcal{P}(x) = \xi \) can then be solved by inverse interpolation, i.e. by interpolating in the table \( (\xi_i, x_i) \), where \( \xi_i \equiv \mathcal{P}(x_i) \) (\( \xi \) is regarded as the independent variable). Care must be exercised to make sure that the numerical integration and interpolation do not introduce significant errors.

\[
\begin{align*}
\text{(a)} & \quad p(x) \\
\text{(b)} & \quad p(x)
\end{align*}
\]

**Figure 1.2:** Random sampling from a continuous distribution \( p(x) \) using the numerical inverse transform method with \( N = 20 \) grid points. a) Piecewise constant approximation. b) Piecewise linear approximation.

A simple, general, approximate method for numerical sampling from continuous distributions is the following. The values \( x_n (n = 0, 1, \ldots, N) \) of \( x \) for which the cumulative distribution function has the values \( n/N \),

\[
\mathcal{P}(x_n) = \int_{x_{\text{min}}}^{x_n} p(x) \, dx = \frac{n}{N},
\]

are previously computed and stored in memory. Notice that the **exact** probability of having \( x \) in the interval \( (x_n, x_{n+1}) \) is \( 1/N \). We can now sample \( x \) by linear interpolation:
we generate a random number \( \xi \) and consider the quantity \( y \equiv \xi N \), which takes values in the interval \((0, N)\). We set \( n = [y] \), where the symbol \([y]\) denotes the integer part of \( y \) (i.e. the largest integer that is less than \( y \)). The value of \( x \) is obtained as

\[
x = x_n + (x_{n+1} - x_n)u,
\]

\( u \equiv y - n \in (0, 1) \). \( \tag{1.38} \)

This is equivalent to approximating the PDF by a piecewise constant function (see fig. 1.2a). Since the spacing between the points \( x_n \) (at which the cumulative distribution function is specified) is roughly proportional to \( 1/p(x_n) \), the approximation is more accurate in regions where \( p(x) \) is large.

The algorithm can be improved by storing the values \( p(x_n) \) of the PDF at the points \( x_n \) in memory and approximating the PDF in the interval \((x_n, x_{n+1})\) linearly,

\[
p_{\text{lin}}(x) \simeq C_n \left[ p(x_n) + \frac{p(x_{n+1}) - p(x_n)}{x_{n+1} - x_n} (x - x_n) \right], \tag{1.39} \]

with a normalization constant \( C_n \) such that the integral of \( p_{\text{lin}}(x) \) over the interval \((x_n, x_{n+1})\) equals \( 1/N \). In general, this piecewise linear approximation is not continuous. Of course, \( p_{\text{lin}}(x) \) will differ from the exact PDF \( p(x) \) when the latter is not linear in the interval, but the differences are smaller than for the piecewise constant approximation with the same number \( N \) of grid points (see fig. 1.2). Again, the approximation is better where \( p(x) \) is larger. An exact algorithm for random sampling from the piecewise linear approximation (1.39) is the following,

(i) Generate a random number \( \xi \) and set \( y = \xi N \), \( n = [y] \) and \( u = y - n \).

(ii) If \( p(x_n) \neq 0 \), set \( r = p(x_{n+1})/p(x_n) \) and

\[
t = \begin{cases} 
    \frac{(1 - u + r^2 u)^{1/2} - 1}{r - 1} & \text{if } r \neq 1, \\
    u & \text{if } r = 1.
\end{cases} \tag{1.40} \]

(iii) If \( p(x_n) = 0 \), set \( t = u^{1/2} \).

(iv) Deliver \( x = x_n + (x_{n+1} - x_n)t \).

1.2.3 Discrete distributions

The inverse transform method can also be applied to discrete distributions. Consider that the random variable \( x \) can take the discrete values \( x = 1, \ldots, N \) with point probabilities \( p_1, \ldots, p_N \), respectively. The corresponding PDF can be expressed as

\[
p(x) = \sum_{i=1}^{N} p_i \delta(x - i), \tag{1.41} \]

where \( \delta(x) \) is the Dirac distribution. Here \( p(x) \) is assumed to be defined in an interval \((a, b)\) with \( a < 1 \) and \( b > N \). The corresponding cumulative distribution function is

\[
P(x) = \sum_{i=1}^{\lfloor x \rfloor} p_i, \tag{1.42} \]
1.2. Random sampling methods

where \([x]\) stands for the integer part of \(x\). Notice that \(P(x) = 0\) when \(x < 1\). Then, eq. (1.32) leads to the sampling formula

\[
x = \begin{cases} 
1 & \text{if } \xi \leq p_1 \\
2 & \text{if } p_1 < \xi \leq p_1 + p_2 \\
& \vdots \\
= j & \text{if } \sum_{i=1}^{j-1} p_i < \xi \leq \sum_{i=1}^{j} p_i \\
& \vdots
\end{cases}
\]  

(1.43)

We can define the quantities

\[
P_1 = 0, \quad P_2 = p_1, \quad P_3 = p_1 + p_2, \quad \ldots, \quad P_{N+1} = \sum_{i=1}^{N} p_i = 1.
\]  

(1.44)

To sample \(x\) we generate a random number \(\xi\) and set \(x\) equal to the index \(i\) such that

\[
P_i < \xi \leq P_{i+1}.
\]  

(1.45)

If the number \(N\) of \(x\)-values is large, this sampling algorithm may be quite slow because of the large number of comparisons needed to determine the sampled value. The easiest method to reduce the number of comparisons is to use binary search instead of sequential search. The algorithm for binary search, for a given value of \(\xi\), proceeds as follows:

(i) Set \(i = 1\) and \(j = N + 1\).
(ii) Set \(k = \lfloor (i + j) / 2 \rfloor\).
(iii) If \(P_k < \xi\), set \(i = k\); otherwise set \(j = k\).
(iv) If \(j - i > 1\), go to step (ii).
(v) Deliver \(i\).

When \(2^n < N \leq 2^{n+1}\), \(i\) is obtained after \(n + 1\) comparisons. This number of comparisons is evidently much less than the number required when using purely sequential search.

Walker’s aliasing method

Walker (1977) described an optimal sampling method for discrete distributions, which yields the sampled value with only one comparison. The idea underlying Walker’s method can be easily understood by resorting to graphical arguments (Salvat, 1987). For this purpose, let us represent the PDF (1.41) as a histogram constructed with \(N\) bars of width \(1/N\) and heights \(Np_i\) (see fig. 1.3). Now, the histogram bars can be cut off at convenient heights and the resulting pieces can be arranged to fill up the square of unit side in such a way that each vertical line crosses, at most, two different pieces. This arrangement can be performed systematically by selecting the lowest and the highest bars in the histogram, say the \(\ell\)th and the \(j\)th, respectively, and by cutting the highest bar off to complete the lowest one, which subsequently is kept unaltered. In order to
keep track of the performed transformation, we label the added piece with the “alias”
value $K_{\ell} = j$, giving its original position in the histogram, and introduce the “cutoff”
value $F_{\ell}$ defined as the height of the lower piece in the $\ell$th bar of the resulting square.
This lower piece keeps the label $\ell$. Evidently, iteration of this process eventually leads
to the complete square (after $N - 1$ steps). Notice that the point probabilities $p_i$ can
be reconstructed from the alias and cutoff values. We have

$$Np_i = F_i + \sum_{j \neq i} (1 - F_j)\delta(i, K_j),$$

(1.46)

where $\delta(i, j)$ denotes the Kronecker delta ($= 1$ if $i = j$ and $= 0$ otherwise). Walker’s
method for random sampling of $x$ proceeds as follows: We sample two independent
random numbers, say $\xi_1$ and $\xi_2$, and define the random point $(\xi_1, \xi_2)$, which is uniformly
distributed in the square. If $(\xi_1, \xi_2)$ lies over a piece labelled with the index $i$, we take
$x = i$ as the selected value. Obviously, the probability of obtaining $i$ as a result of the
sampling equals the fractional area of the pieces labelled with $i$, which coincides with
$p_i$.

![Figure 1.3](image)

Figure 1.3: Graphical representation of the inverse transform method (top) and Walker’s
aliasing method (bottom) for random sampling from a discrete distribution. In this example,
the random variable can take the values $i = 1, 2, 3$ and $4$ with relative probabilities $1, 2, 5$
and $8$, respectively.

As formulated above, Walker’s algorithm requires the generation of two random
numbers for each sampled value of $x$. With the aid of the following trick, the $x$-value
1.2. Random sampling methods

can be generated from a single random number. Continuing with our graphical picture, assume that the $N$ bars in the square are aligned consecutively to form a segment of length $N$ (bottom of fig. 1.3). To sample $x$, we can generate a single random value $\xi N$, which is uniformly distributed in $(0,N)$ and determines one of the segment pieces. The result of the sampling is the label of the selected piece. Explicitly, the sampling algorithm proceeds as follows:

(i) Generate a random number $\xi$ and set $R = \xi N + 1$.
(ii) Set $i = \lfloor R \rfloor$ and $r = R - i$.
(iii) If $r > F_i$, deliver $x = K_i$.
(iv) Deliver $x = i$.

We see that the sampling of $x$ involves only the generation of a random number and one comparison (irrespective of the number $N$ of possible outcomes). The price we pay for this simplification reduces to doubling the number of memory locations that are needed: the two arrays $K_i$ and $F_i$ are used instead of the single array $p_i$ (or $P_i$).

Unfortunately, the calculation of alias and cutoff values is fairly involved and this limits the applicability of Walker’s algorithm to distributions that remain constant during the course of the simulation.

1.2.4 Rejection methods

The inverse transform method for random sampling is based on a one-to-one correspondence between $x$ and $\xi$ values, which is expressed in terms of a single-valued function. There is another kind of sampling method, due to von Neumann, that consists of sampling a random variable from a certain distribution [different from $p(x)$] and subjecting it to a random test to determine whether it will be accepted for use or rejected. These rejection methods lead to very general techniques for sampling from any PDF.

The rejection algorithms can be understood in terms of simple graphical arguments (fig. 1.4). Consider that, by means of the inverse transform method or any other available sampling method, random values of $x$ are generated from a PDF $\pi(x)$. For each sampled value of $x$ we sample a random value $y$ uniformly distributed in the interval $(0,C\pi(x))$, where $C$ is a positive constant. Evidently, the points $(x,y)$, generated in this way, are uniformly distributed in the region $A$ of the plane limited by the $x$-axis ($y = 0$) and the curve $y = C\pi(x)$. Conversely, if (by some means) we generate random points $(x,y)$ uniformly distributed in $A$, their $x$-coordinate is a random variable distributed according to $\pi(x)$ (irrespective of the value of $C$). Now, consider that the distribution $\pi(x)$ is such that $C\pi(x) \geq p(x)$ for some $C > 0$ and that we generate random points $(x,y)$ uniformly distributed in the region $A$ as described above. If we reject the points with $y > p(x)$, the accepted ones (with $y \leq p(x)$) are uniformly distributed in the region between the $x$-axis and the curve $y = p(x)$ and hence, their $x$-coordinate is distributed according to $p(x)$.

A rejection method is thus completely specified by representing the PDF $p(x)$ as

$$p(x) = C\pi(x)r(x),$$  \hspace{1cm} (1.47)
where $\pi(x)$ is a PDF that can be easily sampled e.g. by the inverse transform method, $C$ is a positive constant and the function $r(x)$ satisfies the conditions $0 < r(x) \leq 1$. The rejection algorithm for sampling from $p(x)$ proceeds as follows:

(i) Generate a random value $x$ from $\pi(x)$.
(ii) Generate a random number $\xi$.
(iii) If $\xi > r(x)$, go to step (i).
(iv) Deliver $x$.

From the geometrical arguments given above, it is clear that the algorithm does yield $x$ values distributed according to $p(x)$. The following is a more formal proof: Step (i) produces $x$-values in the interval $(x, x + dx)$ with probability $\pi(x) \, dx$, these values are accepted with probability $r(x) = p(x)/[C\pi(x)]$ and, therefore, (apart from a normalization constant) the probability of delivering a value in $(x, x + dx)$ is equal to $p(x) \, dx$ as required. It is important to realize that, as regards Monte Carlo, the normalization of the simulated PDF is guaranteed by the mere fact that the algorithm delivers some value of $x$.

The efficiency of the algorithm, i.e. the probability of accepting a generated $x$-value, is

$$
\epsilon = \int_x^\phi r(x)\pi(x) \, dx = \frac{1}{C}.
$$

Graphically, the efficiency equals the ratio of the areas under the curves $y = p(x)$ and $y = C\pi(x)$, which are 1 and $C$, respectively. For a given $\pi(x)$, since $r(x) \leq 1$, the constant $C$ must satisfy the condition $C\pi(x) \geq p(x)$ for all $x$. The minimum value of $C$, with the requirement that $C\pi(x) = p(x)$ for some $x$, gives the optimum efficiency.

The PDF $\pi(x)$ in eq. (1.47) should be selected in such a way that the resulting sampling algorithm is as fast as possible. In particular, random sampling from $\pi(x)$

![Figure 1.4: Random sampling from a distribution $p(x)$ using a rejection method.](image-url)
must be performed rapidly, by the inverse transform method or by the composition method (see below). High efficiency is also desirable, but not decisive. One hundred percent efficiency is obtained only with \( \pi(x) = p(x) \) (but random sampling from this PDF is just the problem we want to solve); any other PDF gives a lower efficiency. The usefulness of the rejection method lies in the fact that a certain loss of efficiency can be largely compensated with the ease of sampling \( x \) from \( \pi(x) \) instead of \( p(x) \). A disadvantage of this method is that it requires the generation of several random numbers \( \xi \) to sample each \( x \)-value.

### 1.2.5 Two-dimensional variables. Composition methods

Let us consider a two-dimensional random variable \((x, y)\) with joint probability distribution \( p(x, y) \). Introducing the marginal PDF \( q(y) \) and the conditional PDF \( p(x|y) \) [see eqs. (1.18) and (1.20)],

\[
q(y) \equiv \int p(x, y) \, dx, \quad p(x|y) = \frac{p(x, y)}{q(y)},
\]

the two-variate distribution can be expressed as

\[
 p(x, y) = q(y) p(x|y).
\]  

(1.49)

It is now evident that to generate random points \((x, y)\) from \( p(x, y) \) we can first sample \( y \) from \( q(y) \) and then \( x \) from \( p(x|y) \). Hence, two-dimensional random variables can be generated by using single-variable sampling methods. This is also true for multivariate distributions, because an \( n \)-dimensional PDF can always be expressed as the product of a single-variable marginal distribution and an \((n - 1)\)-dimensional conditional PDF.

From the definition of the marginal PDF of \( x \),

\[
q(x) \equiv \int p(x, y) \, dy = \int q(y) \, p(x|y) \, dy,
\]

(1.50)

it is clear that if we sample \( y \) from \( q(y) \) and, then, \( x \) from \( p(x|y) \), the generated values of \( x \) are distributed according to \( q(x) \). This idea is the basis of the composition methods, which are applicable when \( p(x) \), the distribution to be simulated, is a probability mixture of several PDFs. More specifically, we consider that \( p(x) \) can be expressed as

\[
p(x) = \int w(y) \, p_y(x) \, dy,
\]

(1.51)

where \( w(y) \) is a continuous distribution and \( p_y(x) \) is a family of one-parameter PDFs, where \( y \) is the parameter identifying a unique distribution. Notice that if the parameter \( y \) takes only integer values \( y = i \) with point probabilities \( w_i \), we would write

\[
p(x) = \sum_i w_i \, p_i(x).
\]

(1.52)
The composition method for random sampling from the PDF $p(x)$ is as follows. First, a value of $y$ (or $i$) is drawn from the PDF $w(y)$ and then $x$ is sampled from the PDF $p_j(x)$ for that chosen $y$.

This technique may be applied to generate random values from complex distributions obtained by combining simpler distributions that are themselves easily generated, by the inverse transform method or by rejection methods.

Devising fast, exact methods for random sampling from a given PDF is an interesting technical challenge. The ultimate criterion for the quality of a sampling algorithm is its speed in actual simulations: the best algorithm is the fastest. However, programming simplicity and elegance may justify the use of slower algorithms. For simple analytical distributions that have an analytical inverse cumulative distribution function, the inverse transform method is usually satisfactory. This is the case for a few elementary distributions (e.g. the uniform and exponential distributions considered above). The inverse transform method is also adequate for discrete distributions and for continuous PDFs given in numerical form. By combining the inverse transform, rejection and composition methods we can devise sampling algorithms for virtually any (single- or multivariate) PDF.

**Example 1. Sampling from the normal distribution**

Frequently, we need to generate random values from the normal (or Gaussian) distribution

$$p_G(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right). \tag{1.53}$$

Since the cumulative distribution function cannot be inverted analytically, the inverse transform method is not appropriate. The easiest (but not the fastest) method to sample from the normal distribution consists of generating two independent random variables at a time, as follows. Let $x_1$ and $x_2$ be two independent normal variables. They determine a random point in the plane with PDF

$$p_{2G}(x_1, x_2) = p_G(x_1) p_G(x_2) = \frac{1}{2\pi} \exp\left[-\left(x_1^2 + x_2^2\right)/2\right].$$

Introducing the polar coordinates $r$ and $\phi$,

$$x_1 = r \cos \phi, \quad x_2 = r \sin \phi,$$

the PDF can be expressed as

$$p_{2G}(x_1, x_2) \, dx_1 \, dx_2 = \frac{1}{2\pi} \exp\left(-r^2/2\right) r \, d\phi \, dr = \left[\exp\left(-r^2/2\right) r \, dr\right] \left[\frac{1}{2\pi} \, d\phi\right].$$

We see that $r$ and $\phi$ are independent random variables. The angle $\phi$ is distributed uniformly on $(0, 2\pi)$ and can be sampled as $\phi = 2\pi \xi$. The PDF of $r$ is $\exp(-r^2/2)$ for $r$ and
the corresponding cumulative distribution function is $\mathcal{P}(r) = 1 - \exp(-r^2/2)$. Therefore, $r$ can be generated by the inverse transform method as

$$ r = \sqrt{-2\ln(1 - \xi)} = \sqrt{-2\ln \xi}. $$

The two independent normal random variables are given by

$$ x_1 = \sqrt{-2\ln \xi_1} \cos(2\pi \xi_2), $$

$$ x_2 = \sqrt{-2\ln \xi_1} \sin(2\pi \xi_2), \quad (1.54) $$

where $\xi_1$ and $\xi_2$ are two independent random numbers. This procedure is known as the Box-Müller method. It has the advantages of being exact and easy to program (it can be coded as a single FORTRAN statement).

The mean and variance of the normal variable are $\langle x \rangle = 0$ and $\text{var}(x) = 1$. The linear transformation

$$ X = m + \sigma x \quad (\sigma > 0) \quad (1.55) $$

defines a new random variable. From the properties (1.14) and (1.29), we have

$$ \langle X \rangle = m \quad \text{and} \quad \text{var}(X) = \sigma^2. \quad (1.56) $$

The PDF of $X$ is

$$ p(X) = p_G(x) \frac{dx}{dX} = \frac{1}{\sigma \sqrt{2\pi}} \exp \left[ -\frac{(X - m)^2}{2\sigma^2} \right], \quad (1.57) $$

i.e. $X$ is normally distributed with mean $m$ and variance $\sigma^2$. Hence, to generate $X$ we only have to sample $x$ using the Box-Müller method and apply the transformation (1.55).

**Example 2. Uniform distribution on the unit sphere**

In radiation transport, the direction of motion of a particle is described by a unit vector $\hat{d}$. Given a certain frame of reference, the direction $\hat{d}$ can be specified by giving either its direction cosines $(u, v, w)$ (i.e. the projections of $\hat{d}$ on the directions of the coordinate axes) or the polar angle $\theta$ and the azimuthal angle $\phi$, defined as in fig. 1.5,

$$ \hat{d} = (u, v, w) = (\sin \theta \cos \phi, \sin \theta \sin \phi, \cos \theta). \quad (1.58) $$

Notice that $\theta \in (0, \pi)$ and $\phi \in (0, 2\pi)$.

A direction vector can be regarded as a point on the surface of the unit sphere. Consider an isotropic source of particles, i.e. such that the initial direction $(\theta, \phi)$ of emitted particles is a random point uniformly distributed on the surface of the sphere. The PDF is

$$ p(\theta, \phi) \, d\theta \, d\phi = \frac{1}{4\pi} \sin \theta \, d\theta \, d\phi = \frac{1}{2\pi} \sin \theta \left[ \frac{1}{2\pi} \, d\phi \right]. \quad (1.59) $$
That is, $\theta$ and $\phi$ are independent random variables with PDFs $p_\theta(\theta) = \sin \theta / 2$ and $p_\phi(\phi) = 1/(2\pi)$, respectively. Therefore, the initial direction of a particle from an isotropic source can be generated by applying the inverse transform method to these PDFs,

$$\theta = \arccos(1 - 2\xi_1), \quad \phi = 2\pi\xi_2.$$  \hfill (1.60)

In some cases, it is convenient to replace the polar angle $\theta$ by the variable

$$\mu = (1 - \cos \theta)/2,$$  \hfill (1.61)

which varies from 0 ($\theta = 0$) to 1 ($\theta = \pi$). In the case of an isotropic distribution, the PDF of $\mu$ is

$$p_\mu(\mu) = p_\theta(\theta) \left(\frac{d\mu}{d\theta}\right)^{-1} = 1.$$  \hfill (1.62)

That is, a set of random points $(\mu, \phi)$ uniformly distributed on the rectangle $(0, 1) \times (0, 2\pi)$ corresponds to a set of random directions $(\theta, \phi)$ uniformly distributed on the unit sphere.

## 1.3 Monte Carlo integration

As pointed out by James (1980), at least in a formal sense, all Monte Carlo calculations are equivalent to integrations. This equivalence permits a formal theoretical foundation for Monte Carlo techniques. An important aspect of simulation is the evaluation of the statistical uncertainties of the calculated quantities. We shall derive the basic formulae by considering the simplest Monte Carlo calculation, namely, the evaluation of a unidimensional integral. Evidently, the results are also valid for multidimensional integrals.
1.3. Monte Carlo integration

Consider the integral

\[ I = \int_a^b F(x) \, dx, \]

which we recast in the form of an expectation value,

\[ I = \int f(x) \, p(x) \, dx \equiv \langle f \rangle, \]

by introducing an arbitrary PDF \( p(x) \) and setting \( f(x) = F(x)/p(x) \) [it is assumed that \( p(x) > 0 \) in \((a, b)\) and \( p(x) = 0 \) outside this interval]. The Monte Carlo evaluation of the integral \( I \) is very simple: generate a large number \( N \) of random points \( x_i \) from the PDF \( p(x) \) and accumulate the sum of values \( f(x_i) \) in a counter. At the end of the calculation, the expected value of \( f \) is estimated as

\[ \bar{f} = \frac{1}{N} \sum_{i=1}^{N} f(x_i). \]

The law of large numbers says that, as \( N \) becomes very large,

\[ \bar{f} \to I \quad \text{(in probability)}. \]

In statistical terminology, this means that \( \bar{f} \), the Monte Carlo result, is a consistent estimator of the integral (1.63). This is valid for any function \( f(x) \) that is finite and piecewise continuous, i.e. with a finite number of discontinuities.

The law of large numbers (1.66) can be restated as

\[ \langle f \rangle = \lim_{N \to \infty} \frac{1}{N} \sum_{i=1}^{N} f(x_i). \]

By applying this law to the integral that defines the variance of \( f(x) \) [cf. eq. (1.16)]

\[ \text{var} \{ f(x) \} = \int f^2(x) \, p(x) \, dx - \langle f \rangle^2, \]

we obtain

\[ \text{var} \{ f(x) \} = \lim_{N \to \infty} \left\{ \frac{1}{N} \sum_{i=1}^{N} [f(x_i)]^2 - \left[ \frac{1}{N} \sum_{i=1}^{N} f(x_i) \right]^2 \right\}. \]

The expression in curly brackets is a consistent estimator of the variance of \( f(x) \). It is advisable (see below) to accumulate the squared function values \([f(x_i)]^2\) in a counter and, at the end of the simulation, estimate \( \text{var} \{ f(x) \} \) according to eq. (1.69).

It is clear that different Monte Carlo runs [with different, independent sequences of \( N \) random numbers \( x_i \) from \( p(x) \)] will yield different estimates \( \bar{f} \). This implies that the outcome of our Monte Carlo code is affected by statistical uncertainties, similar to those found in laboratory experiments, which need to be properly evaluated to determine the “accuracy” of the Monte Carlo result. For this purpose, we may consider \( \bar{f} \) as a random
variable, the PDF of which is, in principle, unknown. Its mean and variance are given by
\[
\langle \mathcal{F} \rangle = \left\langle \frac{1}{N} \sum_{i=1}^{N} f(x_i) \right\rangle = \frac{1}{N} \sum_{i=1}^{N} \langle f \rangle = \langle f \rangle
\]  
(1.70)
and
\[
\text{var}(\mathcal{F}) = \text{var} \left[ \frac{1}{N} \sum_{i=1}^{N} f(x_i) \right] = \frac{1}{N^2} \sum_{i=1}^{N} \text{var}(f(x)) = \frac{1}{N} \text{var}(f(x)),
\]
(1.71)
where use has been made of properties of the expectation and variance operators. The standard deviation (or standard error) of \( \mathcal{F} \),
\[
\sigma_f \equiv \sqrt{\text{var}(\mathcal{F})} = \sqrt{\frac{\text{var}(f(x))}{N}},
\]
gives a measure of the statistical uncertainty of the Monte Carlo estimate \( \mathcal{F} \). The result (1.72) has an important practical implication: in order to reduce the statistical uncertainty by a factor of 10, we have to increase the sample size \( N \) by a factor of 100. Evidently, this sets a limit to the accuracy that can be attained with the available computer power.

We can now invoke the central limit theorem (see e.g. James, 1980), which establishes that, in the limit \( N \to \infty \), the PDF of \( \mathcal{F} \) is a normal (Gaussian) distribution with mean \( \langle f \rangle \) and standard deviation \( \sigma_f \),
\[
p(\mathcal{F}) = \frac{1}{\sigma_f \sqrt{2\pi}} \exp \left( -\frac{(\mathcal{F} - \langle f \rangle)^2}{2\sigma_f^2} \right),
\]
(1.73)
It follows that, for sufficiently large values of \( N \), for which the theorem is applicable, the interval \( \mathcal{F} \pm n\sigma_f \) contains the exact value \( \langle f \rangle \) with a probability of 68.3\% if \( n = 1 \), 95.4\% if \( n = 2 \) and 99.7\% if \( n = 3 \) (3\( \sigma \) rule).

The central limit theorem is a very powerful tool, since it predicts that the generated values of \( \mathcal{F} \) follow a specific distribution, but it applies only asymptotically. The minimum number \( N \) of sampled values needed to apply the theorem with confidence depends on the problem under consideration. If, in the case of our problem, the third central moment of \( f \),
\[
\mu_3 \equiv \int [f(x) - \langle f \rangle]^3 p(x) \, dx,
\]
exists, the theorem is essentially satisfied when
\[
|\mu_3| \ll \sigma_f^3 \sqrt{N}.
\]
(1.75)
In general, it is advisable to study the distribution of the estimator to ascertain the applicability of the central limit theorem. In most Monte Carlo calculations, however, statistical errors are estimated by simply assuming that the theorem is satisfied, irrespective of the sample size. We shall adopt this practice and report Monte Carlo results in the form \( \mathcal{F} \pm 3\sigma_f \). In simulations of radiation transport, this is empirically validated
by the fact that simulated continuous distributions do “look” continuous (i.e. the “error bars” define a smooth band).

Each possible \( p(x) \) defines a Monte Carlo algorithm to calculate the integral \( I \), eq. (1.63). The simplest algorithm (crude Monte Carlo) is obtained by using the uniform distribution \( p(x) = 1/(b-a) \). Evidently, \( p(x) \) determines not only the density of sampled points \( x_i \), but also the magnitude of the variance \( \text{var}\{f(x)\} \), eq. (1.68),

\[
\text{var}\{f(x)\} = \int_a^b p(x) \left[ \frac{F(x)}{p(x)} \right]^2 \, dx - I^2 = \int_a^b F(x) \left[ \frac{F(x)}{p(x)} - 1 \right] \, dx. \tag{1.76}
\]

As a measure of the effectiveness of a Monte Carlo algorithm, it is common to use the efficiency \( \epsilon \), which is defined by

\[
\epsilon = 1/[\sigma_f^2 T], \tag{1.77}
\]

where \( T \) is the computing time (or any other measure of the calculation effort) needed to get the simulation result. Since \( \sigma_f^2 \) and \( T \) are roughly proportional to \( N^{-1} \) and \( N \), respectively, \( \epsilon \) is a constant (i.e. it is independent of \( N \), on average.

The so-called variance-reduction methods are techniques that aim to optimize the efficiency of the simulation through an adequate choice of the PDF \( p(x) \). Improving the efficiency of the algorithms is an important, and delicate, part of the art of Monte Carlo simulation. The interested reader is addressed to the specialized bibliography (e.g. Rubinstein, 1981). Although of common use, the term “variance reduction” is somewhat misleading, since a reduction in variance does not necessarily lead to improved efficiency. To make this clear, consider that a Monte Carlo algorithm, based on a certain PDF \( p(x) \), has a variance that is less than that of crude Monte Carlo (i.e. with the uniform distribution); if the generation of \( x \)-values from \( p(x) \) takes a longer time than for the uniform distribution, the “variance-reduced” algorithm may be less efficient than crude Monte Carlo. Hence, one should avoid using PDFs that are too difficult to sample.

### 1.4 Simulation of radiation transport

In this section, we describe the essentials of Monte Carlo simulation of radiation transport. For the sake of simplicity, we limit our considerations to the detailed simulation method, where all the interaction events experienced by a particle are simulated in chronological succession, and we disregard the production of secondary particles, so that only one kind of particle is transported.

#### 1.4.1 Scattering model and probability distribution functions

Consider a particle with energy \( E \) (kinetic energy, in the case of electrons and positrons) moving in a given medium. We limit our considerations to homogeneous “random scattering” media, such as gases, liquids and amorphous solids, where the “molecules”
are distributed at random with uniform density. The composition of the medium is specified by its stoichiometric formula, i.e. atomic number $Z_i$ and number of atoms per molecule $n_i$ of all the elements present. The stoichiometric indices $n_i$ need not have integer values. In the case of alloys, for instance, they may be set equal to the percentage in number of each element and then a “molecule” is a group of 100 atoms with the appropriate proportion of each element. The “molecular weight” is $A_M = \Sigma n_i A_i$, where $A_i$ is the atomic weight of the $i$-th element. The number of molecules per unit volume is given by

$$N = N_A \frac{\rho}{A_M},$$

(1.78)

where $N_A$ is Avogadro’s number and $\rho$ is the mass density of the material.

In each interaction, the particle may lose energy $W$ and/or change its direction of movement. The angular deflection is determined by the polar scattering angle $\theta$, i.e. the angle between the directions of the particle before and after the interaction, and the azimuthal angle $\phi$. Let us assume that the particle can interact with the medium through two independent mechanisms, denoted as “A” and “B” (for instance, elastic and inelastic scattering, in the case of low-energy electrons). The scattering model is completely specified by the molecular differential cross sections (DCS)

$$\frac{d^2 \sigma_A}{dWd\Omega}(E; W, \theta) \quad \text{and} \quad \frac{d^2 \sigma_B}{dWd\Omega}(E; W, \theta),$$

(1.79)

where $d\Omega$ is a solid angle element in the direction $(\theta, \phi)$. We have made the parameteric dependence of the DCSs on the particle energy $E$ explicit. Considering that the molecules in the medium are oriented at random, the DCS is independent of the azimuthal scattering angle, i.e. the angular distribution of scattered particles is axially symmetrical around the direction of incidence. The total cross sections (per molecule) are

$$\sigma_{A,B}(E) = \int_0^E dW \int_0^{\pi} 2\pi \sin \theta d\theta \frac{d^2 \sigma_{A,B}}{dWd\Omega}(E; W, \theta).$$

(1.80)

The PDFs of the energy loss and the polar scattering angle in individual scattering events are

$$p_{A,B}(E; W, \theta) = \frac{2\pi \sin \theta}{\sigma_{A,B}(E)} \frac{d^2 \sigma_{A,B}}{dWd\Omega}(E; W, \theta).$$

(1.81)

Notice that $p_A(E; W, \theta)dWd\theta$ gives the (normalized) probability that, in a scattering event of type A, the particle loses energy in the interval $(W, W + dW)$ and is deflected into directions with polar angle (relative to the initial direction) in the interval $(\theta, \theta + d\theta)$. The azimuthal scattering angle in each collision is uniformly distributed in the interval $(0, 2\pi)$, i.e.

$$p(\phi) = \frac{1}{2\pi}.$$  

(1.82)

The total interaction cross section is

$$\sigma_T(E) = \sigma_A(E) + \sigma_B(E).$$

(1.83)
1.4. Simulation of radiation transport

When the particle interacts with the medium, the kind of interaction that occurs is a discrete random variable, that takes the values "A" and "B" with probabilities

\[ p_A = \frac{\sigma_A}{\sigma_T} \quad \text{and} \quad p_B = \frac{\sigma_B}{\sigma_T}. \]  

It is worth recalling that this kind of single scattering model is only valid when diffraction effects resulting from coherent scattering from several centres (e.g. Bragg diffraction, channelling of charged particles) are negligible. This means that the simulation is applicable only to amorphous media and, with some care, to polycrystalline solids.

To get an intuitive picture of the scattering process, we can imagine each molecule as a sphere of radius \( r_s \) such that the cross-sectional area \( \pi r_s^2 \) equals the total cross section \( \sigma_T \). Now, assume that a particle impinges normally on a very thin material foil of thickness \( ds \). What the particle sees in front of it is a uniform distribution of \( \mathcal{N} \) ds spheres per unit surface. An interaction takes place when the particle strikes one of these spheres. Therefore, the probability of interaction within the foil equals the fractional area covered by the spheres, \( \mathcal{N} \sigma_T \) ds. In other words, \( \mathcal{N} \sigma_T \) is the interaction probability per unit path length. Its inverse,

\[ \lambda_T \equiv (\mathcal{N} \sigma_T)^{-1}, \]  

is the (total) mean free path between interactions.

Let us now consider a particle that moves within an unbound medium. The PDF \( p(s) \) of the path length \( s \) of the particle from its current position to the site of the next interaction may be obtained as follows. The probability that the particle travels a path length \( s \) without interacting is

\[ \mathcal{F}(s) = \int_s^\infty p(s') \, ds'. \]  

The probability \( p(s) \) ds of having the next interaction when the travelled length is in the interval \( (s, s + ds) \) equals the product of \( \mathcal{F}(s) \) (the probability of arrival at \( s \) without interacting) and \( \lambda_T^{-1} \) ds (the probability of interacting within \( ds \)). It then follows that

\[ p(s) = \lambda_T^{-1} \int_s^\infty p(s') \, ds'. \]  

The solution of this integral equation, with the boundary condition \( p(\infty) = 0 \), is the familiar exponential distribution

\[ p(s) = \lambda_T^{-1} \exp \left( -s/\lambda_T \right). \]  

Notice that the mean free path \( \lambda_T \) coincides with the average path length between collisions:

\[ \langle s \rangle = \int_0^\infty s \, p(s) \, ds = \lambda_T. \]  

The differential inverse mean free path for the interaction process \( A \) is defined as

\[ \frac{d^2 \lambda_A^{-1}}{dWd\Omega}(E; W, \theta) = \mathcal{N} \frac{d^2 \sigma_A}{dWd\Omega}(E; W, \theta). \]
Evidently, the integral of the differential inverse mean free path gives the inverse mean free path for the process,

$$\lambda^{-1}_A = \int dW \int 2\pi \sin \theta \, d\theta \, \frac{d^2 \lambda^{-1}_A}{dW} (E; W, \theta) = N \sigma_A. \tag{1.91}$$

In the literature, the product $N \sigma_A$ is frequently called the macroscopic cross section, although this name is not appropriate for a quantity that has the dimensions of inverse length. Notice that the total inverse mean free path is the sum of the inverse mean free paths of the different active interaction mechanisms,

$$\lambda^{-1}_T = \lambda^{-1}_A + \lambda^{-1}_B. \tag{1.92}$$

### 1.4.2 Generation of random tracks

Each particle track starts off at a given position, with initial direction and energy in accordance with the characteristics of the source. The “state” of a particle immediately after an interaction (or after entering the sample or starting its trajectory) is defined by its position coordinates $r = (x, y, z)$, energy $E$ and direction cosines of the direction of flight, i.e. the components of the unit vector $\mathbf{d} = (u, v, w)$, as seen from the laboratory reference frame. Each simulated track is thus characterized by a series of states $r_n, E_n, \mathbf{d}_n$, where $r_n$ is the position of the $n$-th scattering event and $E_n$ and $\mathbf{d}_n$ are the energy and direction cosines of the direction of movement just after that event.

The generation of random tracks proceeds as follows. Let us assume that a track has already been simulated up to a state $r_n, E_n, \mathbf{d}_n$. The length $s$ of the free path to the next collision, the involved scattering mechanism, the change of direction and the energy loss in this collision are random variables that are sampled from the corresponding PDFs, using the methods described in section 1.2. Hereafter, $\xi$ stands for a random number uniformly distributed in the interval $(0, 1)$.

The length of the free flight is distributed according to the PDF given by eq. (1.88). Random values of $s$ are generated by using the sampling formula [see eq. (1.36)]

$$s = -\lambda_T \ln \xi. \tag{1.93}$$

The following interaction occurs at the position

$$r_{n+1} = r_n + s \mathbf{d}_n. \tag{1.94}$$

The type of this interaction ("A" or "B") is selected from the point probabilities given by eq. (1.84) using the inverse transform method (section 1.2.2). The energy loss $W$ and the polar scattering angle $\theta$ are sampled from the distribution $p_{AB}(E; W; \theta)$, eq. (1.81), by using a suitable sampling technique. The azimuthal scattering angle is generated, according to the uniform distribution in $(0, 2\pi)$, as $\phi = 2\pi \xi$.

After sampling the values of $W$, $\theta$ and $\phi$, the energy of the particle is reduced, $E_{n+1} = E_n - W$, and the direction of movement after the interaction $\mathbf{d}_{n+1} = (u', v', w')$
is obtained by performing a rotation of \( \hat{\mathbf{d}}_n = (u, v, w) \) (see fig. 1.6). The rotation matrix \( R(\theta, \phi) \) is determined by the polar and azimuthal scattering angles. To explicitly obtain the direction vector \( \mathbf{d}_{n+1} = R(\theta, \phi) \hat{\mathbf{d}}_n \) after the interaction, we first note that, if the initial direction is along the \( z \)-axis, the direction after the collision is

\[
\begin{pmatrix}
\sin \theta \cos \phi \\
\sin \theta \sin \phi \\
\cos \theta
\end{pmatrix}
= R_z(\phi) R_y(\theta) \hat{\mathbf{z}},
\]

where \( \hat{\mathbf{z}} = (0, 0, 1) \) and

\[
R_y(\theta) = \begin{pmatrix}
\cos \theta & 0 & \sin \theta \\
0 & 1 & 0 \\
-\sin \theta & 0 & \cos \theta
\end{pmatrix}
\text{ and } R_z(\phi) = \begin{pmatrix}
\cos \phi & -\sin \phi & 0 \\
\sin \phi & \cos \phi & 0 \\
0 & 0 & 1
\end{pmatrix}
\]

are rotation matrices corresponding to active rotations of angles \( \theta \) and \( \phi \) about the \( y \) - and \( z \)-axes, respectively. On the other hand, if \( \vartheta \) and \( \varphi \) are the polar and azimuthal angles of the initial direction

\[
\hat{\mathbf{d}}_n = (\sin \vartheta \cos \varphi, \sin \vartheta \sin \varphi, \cos \vartheta),
\]

the rotation \( R_y(-\vartheta)R_z(-\varphi) \) transforms the vector \( \hat{\mathbf{d}}_n \) into \( \hat{\mathbf{z}} \). It is then clear that the final direction vector \( \mathbf{d}_{n+1} \) can be obtained by performing the following sequence of rotations of the initial direction vector: 1) \( R_y(-\vartheta)R_z(-\varphi) \), which transforms \( \mathbf{d}_n \) into \( \hat{\mathbf{z}} \); 2) \( R_z(\varphi) R_y(\theta) \), which rotates \( \hat{\mathbf{z}} \) according to the sampled polar and azimuthal scattering angles; and 3) \( R_z(\varphi) R_y(-\vartheta) \), which inverts the rotation of the first step. Hence

\[
R(\theta, \phi) = R_z(\varphi) R_y(\vartheta) R_z(\phi) R_y(\theta) R_y(-\vartheta) R_z(-\varphi).
\]
The final direction vector is

\[ \hat{d}_{n+1} = R(\theta, \phi) \hat{d}_n = R_z(\phi) R_y(\theta) \begin{pmatrix} \sin \theta \cos \phi \\ \sin \theta \sin \phi \\ \cos \theta \end{pmatrix} \]  

(1.99)

and its direction cosines are

\[ u' = u \cos \theta + \frac{\sin \theta}{\sqrt{1-w^2}} [uw \cos \phi - v \sin \phi], \]

\[ v' = v \cos \theta + \frac{\sin \theta}{\sqrt{1-w^2}} [vw \cos \phi + u \sin \phi], \]

(1.100)

\[ w' = w \cos \theta - \sqrt{1-w^2} \sin \theta \cos \phi. \]

These equations are indeterminate when \( w \simeq \pm 1 \), i.e. when the initial direction is nearly parallel or antiparallel to the z-axis; in this case we can simply set

\[ u = \pm \sin \theta \cos \phi, \quad v = \pm \sin \theta \sin \phi, \quad w = \pm \cos \theta. \]  

(1.101)

Moreover, eqs. (1.100) are not very stable numerically and the normalization of \( \hat{d}_{n+1} \) tends to drift from 1 after repeated usage. This must be remedied by periodically renormalizing \( \hat{d}_{n+1} \). The change of direction expressed by eqs. (1.100) and (1.101) is performed by the subroutine DIRECT (see the PENEOPE source listing).

The simulation of the track then proceeds by repeating these steps. A track is finished either when it leaves the material system or when the energy becomes smaller than a given energy \( E_{\text{abs}} \), which is the energy where particles are assumed to be effectively stopped and absorbed in the medium.

### 1.4.3 Particle transport as a Markov process

The foregoing concepts, definitions and simulation scheme rest on the assumption that particle transport can be modelled as a Markov process\(^2\), i.e. “future values of a random variable (interaction event) are statistically determined by present events and depend only on the event immediately preceeding”. Owing to the Markovian character of the transport, we can stop the generation of a particle history at an arbitrary state (any point of the track) and resume the simulation from this state without introducing any bias in the results.

In mixed simulations of electron/positron transport, it is necessary to limit the length \( s \) of each “free jump” so that it does not exceed a given value \( s_{\text{max}} \). To accomplish this, we still sample the free path length \( s \) to the next interaction from the exponential PDF

\(^2\)The quoted definition is from the Webster’s Encyclopedic Unabridged Dictionary of the English Language (Portland House, New York, 1989).
1.4. Simulation of radiation transport

(1.88), but when \( s > s_{\text{max}} \) we only let the particle advance a distance \( s_{\text{max}} \) along the direction of motion. At the end of the truncated free jump we do nothing (i.e. the particle keeps its energy and direction of motion unaltered); however, for programming convenience, we shall say that the particle suffers a delta interaction (actually, a “non-interaction”). When the sampled value of \( s \) is less than \( s_{\text{max}} \), a real interaction is simulated. After the interaction (either real or delta), we sample a new free path \( s \), move the particle a distance \( s' = \min(s, s_{\text{max}}) \), etc. From the Markovian character of the transport, it is clear that the insertion of delta interactions keeps the simulation unbiased. If you do not see it so clearly, here comes a direct proof. First we note that the probability that a free jump ends with a delta interaction is

\[
p_\delta = \int_{s_{\text{max}}}^{\infty} p(s) \, ds = \exp(-s_{\text{max}}/\lambda_T),
\]

(1.102)

To obtain the probability \( p(s) \) of having the first real interaction at a distance in the interval \( (s, s + ds) \), we write \( s = ns_{\text{max}} + s' \) with \( n = [s/s_{\text{max}}] \) and, hence, \( s' < s_{\text{max}} \). The sought probability is then equal to the probability of having \( n \) successive delta interactions followed by a real interaction at a distance in \( (s', s' + ds) \) from the last, \( n \)-th, delta interaction,

\[
p(s) \, ds = p_\delta \lambda_T^{-1} \exp(-s'/\lambda_T) \, ds = \lambda_T^{-1} \exp(-s/\lambda_T) \, ds,
\]

(1.103)

which is the correct value [cf. eq. (1.88)].

Up to this point, we have considered transport in a single homogeneous medium. In practical cases, however, the material structure where radiation is transported may consist of various regions with different compositions. We assume that the interfaces between contiguous media are sharp (i.e. there is no diffusion of chemical species across them) and passive (which amounts to neglecting e.g. surface plasmon excitation and transition radiation). In the simulation code, when a particle arrives at an interface, it is stopped there and the simulation is resumed with the interaction properties of the new medium. Obviously, this procedure is consistent with the Markovian property of the transport process.

Consider two homogeneous media, 1 and 2 (with corresponding mean free paths \( \lambda_{T,1} \) and \( \lambda_{T,2} \)), separated by an interface, which is crossed by particles that move from the first medium to the second. The average path length between the last real interaction in medium 1 and the first real interaction in medium 2 is \( \lambda_{T,1} + \lambda_{T,2} \), as can be easily verified by simulation. This result seemed paradoxical to some authors and induced confusion in the past. In fact, there is nothing odd here as you may easily verify (again by simulation) as follows. Assume particles being transported within a single homogeneous medium with an imaginary plane that acts as a “virtual” interface, splitting the medium into two halves. In the simulation, the particles do not see this interface, i.e. they do not stop when crossing. Every time a particle crosses the plane, we score the length \( s_{\text{plane}} \) of the track segment between the two real interactions immediately before and after the crossing. It is found that the average value of \( s_{\text{plane}} \) is \( 2\lambda_T \), in spite of the fact that the free path length between consecutive collisions was sampled from an exponential PDF.
with the mean free path $\lambda_T$ [yes, the scored values $s_{\text{plane}}$ were generated from this PDF!]. The explanation of this result is that, as a consequence of the Markovian character, the average path length from the plane (an arbitrary fixed point in the track) back to the last collision (or up to the next collision) is $\lambda_T$.

1.5 Statistical averages and uncertainties

For the sake of being more specific, let us consider the simulation of a high-energy electron beam impinging on the surface of a semi-infinite water phantom. Each primary electron originates a shower of electrons and photons, which are individually tracked down to the corresponding absorption energy. Any quantity of interest $Q$ is evaluated as the average score of a large number $N$ of simulated random showers. Formally, $Q$ can be expressed as an integral of the form (1.64),

$$Q = \int q \, p(q) \, dq,$$

(1.104)

where the PDF $p(q)$ is usually unknown. The simulation of individual showers provides a practical method to sample $q$ from the “natural” PDF $p(q)$: from each generated shower we get a random value $q_i$ distributed according to $p(q)$. The only difference to the case of Monte Carlo integration considered above is that now the PDF $p(q)$ describes a cascade of random interaction events, each with its characteristic PDF. The Monte Carlo estimate of $Q$ is

$$\overline{Q} = \frac{1}{N} \sum_{i=1}^{N} q_i.$$

(1.105)

Thus, for instance, the average energy $E_{\text{dep}}$ deposited within the water phantom per incident electron is obtained as

$$E_{\text{dep}} = \frac{1}{N} \sum_{i=1}^{N} e_i,$$

(1.106)

where $e_i$ is the energy deposited by all the particles of the $i$-th shower. The statistical uncertainty (standard deviation) of the Monte Carlo estimate [eq. (1.72)] is

$$\sigma_Q = \sqrt{\frac{\text{var}(q)}{N}} = \sqrt{\frac{1}{N} \left[ \frac{1}{N} \sum_{i=1}^{N} q_i^2 - \overline{Q}^2 \right]}.$$

(1.107)

As mentioned above, we shall usually express the simulation result in the form $\overline{Q} \pm 3\sigma_Q$, so that the interval $(\overline{Q} - 3\sigma_Q, \overline{Q} + 3\sigma_Q)$ contains the true value $Q$ with 99.7% probability. Notice that to evaluate the standard deviation (1.107) we must score the squared contributions $q_i^2$. In certain cases, the contributions $q_i$ can only take the values 0 and 1, and the standard error can be determined without scoring the squares,

$$\sigma_Q = \sqrt{\frac{1}{N} Q(1 - Q)}.$$

(1.108)
Simulation/scoring can also be used to compute continuous distributions. The simplest method is to “discretize” the distributions, by treating them as histograms, and to determine the “heights” of the different bars. To make the arguments clear, let us consider the depth-dose distribution \( D(z) \), defined as the average energy deposited per unit depth and per incident electron within the water phantom. \( D(z) \) \( dz \) is the average energy deposited at depths between \( z \) and \( z + dz \) per incident electron, and the integral of \( D(z) \) from 0 to \( \infty \) is the average deposited energy \( E_{dep} \) (again, per incident electron). Since part of the energy is reflected back from the water phantom (through backscattered radiation), \( E_{dep} \) is less than the kinetic energy \( E_{inc} \) of the incident electrons. We are interested in determining \( D(z) \) in a limited depth interval, say from \( z = 0 \) to \( z = z_{\text{max}} \). The calculation proceeds as follows. First of all, we have to select a partition of the interval \((0, z_{\text{max}})\) into \( M \) different depth bins \((z_{k-1}, z_k)\), with \( 0 = z_0 < z_1 < \ldots < z_M = z_{\text{max}} \). Let \( e_{ij,k} \) denote the amount of energy deposited into the \( k \)-th bin by the \( j \)-th particle of the \( i \)-th shower (each incident electron may produce multiple secondary particles). The average energy deposited into the \( k \)-th bin (per incident electron) is obtained as

\[
E_k = \frac{1}{N} \sum_{i=1}^{N} e_{i,k} \quad \text{with} \quad e_{i,k} \equiv \sum_{j} e_{ij,k}, \tag{1.109}
\]

and is affected by a statistical uncertainty

\[
\sigma_{E_k} = \sqrt{\frac{1}{N} \sum_{i=1}^{N} \left( \frac{e_{i,k}^2}{N} - E_k^2 \right)}. \tag{1.110}
\]

The Monte Carlo depth-dose distribution \( D_{MC}(z) \) is a stepwise constant function,

\[
D_{MC}(z) = D_k \pm 3\sigma_{D_k} \quad \text{for} \quad z_{k-1} < z < z_k \tag{1.111}
\]

with

\[
D_k \equiv \frac{1}{z_k - z_{k-1}} E_k, \quad \sigma_{D_k} \equiv \frac{1}{z_k - z_{k-1}} \sigma_{E_k}. \tag{1.112}
\]

Notice that the bin average and standard deviation have to be divided by the bin width to obtain the final Monte Carlo distribution. Defined in this way, \( D_{MC}(z) \) is an unbiased estimator of the average dose in each bin. The limitation here is that we are approximating the continuous distribution \( D(z) \) as a histogram with finite bar widths. In principle, we could obtain a closer approximation by using narrower bins. However, care has to be taken in selecting the bin widths since statistical uncertainties may completely hide the information in narrow bins.

A few words regarding programming details are in order. To evaluate the average deposited energy and its standard deviation for each bin, eqs. (1.109) and (1.110), we must score the shower contributions \( e_{i,k} \) and their squares \( e_{i,k}^2 \). There are cases in which a senseless literal application of this recipe may take a large fraction of the simulation time. Consider, for instance, the simulation of the 3D dose distribution in the phantom, which may involve several thousand volume bins. For each bin, the energies \( e_{ij,k} \) deposited by the individual particles of a shower must be accumulated in a partial counter to obtain
the shower contribution $e_{i,k}$ and, after completion of the whole shower, the value $e_{i,k}$
and its square must be added to the accumulated counters. As only a small fraction
of the bins receive energy from a single shower, it is not practical to treat all bin
counters on an equal footing. The fastest method is to transfer partial scores to the
accumulated counters only when the partial counter is going to receive a contribution
from a new shower. This can be easily implemented in a computer program as follows.
For each quantity of interest, say $Q$, we define three real counters, $Q$, $Q2$ and $QP$, and
an integer label $LQ$; all these quantities are initially set to zero. The partial scores $q_{ij}$ of
the particles of a shower are accumulated in the partial counter $QP$, whereas the global
shower contribution $q_i$ and its square are accumulated in $Q$ and $Q2$, respectively. Each
shower is assigned a label, for instance its order number $i$, which is stored in $LQ$ the
first time that the shower contributes to $QP$. In the course of the simulation, the value
of $QP$ is transferred to the global counters $Q$ and $Q2$ only when it is necessary to store a
contribution $q_{ij}$ from a new shower. Explicitly, the FORTRAN code for scoring $Q$ is

```fortran
IF(i .NE. LQ) THEN
  Q=Q+QP
  Q2=Q2+QP**2
  QP=q_{ij}
  LQ=i
ELSE
  QP=QP+q_{ij}
ENDIF
```

At the end of the simulation, the residual contents of $QP$ must be transferred to the
global counters.

For some quantities (e.g., the mean number of scattering events per track, the depth-
dose function, . . .) almost all the simulated tracks contribute to the score and the
inherent statistical uncertainties of the simulation results are comparatively small. Other
quantities (e.g., angle and energy distributions of the particles transmitted through a
thick foil) have considerable statistical uncertainties (i.e., large variances) because only
a small fraction of the simulated tracks contribute to the partial scores.

### 1.6 Variance reduction

In principle, the statistical error of a quantity may be somewhat reduced (without in-
creasing the computer simulation time) by using variance-reduction techniques. Unfor-
nately, these optimization techniques are extremely problem-dependent, and general
recipes to minimize the variance cannot be given. On the other hand, the importance
of variance reduction should not be overvalued. In many cases, analogue\(^3\) simulation
does the work in a reasonable time. Spending manhours by complicating the program,

to get a modest reduction in computing time may not be a good investment. It is

---

\(^3\)We use the term “analogue” to refer to detailed, condensed or mixed simulations that do not incorporate variance-reduction procedures.
1.6. Variance reduction

also important to realize that an efficient variance-reduction method usually lowers the statistical error of a given quantity $Q$ at the expense of increasing the uncertainties of other quantities. Thus, variance-reduction techniques are not recommended when a global description of the transport process is sought. Here we give a brief description of those techniques which, with a modest programming effort, can be useful in improving the solution of some ill-conditioned problems. For the sake of generality, we consider that secondary particles can be generated in the interactions with the medium. A nice, and practically oriented, review of variance-reduction methods in radiation transport has been given by Bielajew and Rogers (1988).

1.6.1 Interaction forcing

Sometimes, a high variance results from an extremely low interaction probability. Consider, for instance, the simulation of the energy spectrum of bremsstrahlung photons emitted by medium energy ($\sim 100$ keV) electrons in a thin foil of a certain material. As radiative events are much less probable than elastic and inelastic scattering, the uncertainty of the simulated photon spectrum will be relatively large. In such cases, an efficient variance-reduction method is to artificially increase the interaction probability of the process $A$ of interest. Our practical implementation of interaction forcing consists of replacing the mean free path $\lambda_A$ of the real process by a shorter one, $\lambda_{Af}$, i.e. we force $A$ interactions to occur more frequently than for the real process. We consider that the PDF for the energy loss, the angular deflections (and the directions of emitted secondary particles, if any) in the forced interactions is the same as for the real interactions. To sample the length of the free jump to the next interaction, we use the exponential distribution with the reduced mean free path $\lambda_{Af}$. This is equivalent to increasing the interaction probability per unit path length of the process $A$ by a factor

$$\mathcal{F} = \frac{\lambda_A}{\lambda_{Af}} > 1.$$  \hspace{1cm} (1.113)

To keep the simulation unbiased, we must correct for the introduced distortion as follows:

(i) A weight $w_p^{(1)} = 1$ is associated with each primary particle. Secondary particles produced in forced interactions have an associated weight $w_p^{(2)} = w_p^{(1)}/\mathcal{F}$; the weights of successive generations of forced secondaries are $w_p^{(k)} = w_p^{(k-1)}/\mathcal{F}$. Secondary particles generated in non-forced interactions (i.e. of types other than $A$) are given a weight equal to that of their parent particle.

(ii) A weight $w_E^{(k)} = w_p^{(k)}/\mathcal{F}$ is given to the deposited energy (and to any other alteration of the medium such as e.g. charge deposition) that results from forced interactions of a particle with weight $w_p^{(k)}$. For non-forced interactions $w_E^{(k)} = w_p^{(k)}$.

(iii) Forced interactions are simulated to determine the energy loss and possible emission of secondary radiation, but the state variables of the interacting particle are altered only with probability $1/\mathcal{F}$. That is, the energy $E$ and direction of movement $\mathbf{d}$ of the projectile are varied only when the value $\xi$ of a random number falls below $1/\mathcal{F}$, otherwise $E$ and $\mathbf{d}$ are kept unchanged.
Of course, interaction forcing should be applied only to interactions that are dynamically allowed, i.e. for particles with energy above the corresponding "reaction" threshold.

Let \( w_{i1} \) and \( q_{i1} \) denote the weight and the contribution to the score of the \( i \)-th primary, and let \( w_{ij} \) and \( q_{ij} \) \((j > 1)\) represent the weights and contributions of the \( j \)-th secondary particles generated by the \( i \)-th primary. The Monte Carlo estimate of \( Q \) obtained from the \( N \) simulated histories is

\[
\bar{Q} = \frac{1}{N} \sum_{i,j} w_{ij} q_{ij}.
\]  
(1.114)

Evidently, the estimates \( \bar{Q} \) obtained with interaction forcing and from an analogue simulation are equal (in the statistical sense, i.e. in the limit \( N \to \infty \), their difference tends to zero). The standard deviation is given by

\[
\sigma_Q = \sqrt{\frac{1}{N} \left[ \frac{1}{N} \sum_i \left( \sum_j w_{ij} q_{ij} \right)^2 - \bar{Q}^2 \right]}.
\]  
(1.115)

Quantities directly related to the forced interactions will have a reduced statistical error, due to the increase in number of these interactions. However, for a given simulation time, other quantities may exhibit standard deviations larger than those of the analogue simulation, because of the time spent in simulating the forced interactions.

### 1.6.2 Splitting and Russian roulette

These two techniques, which are normally used in conjunction, are effective in problems where interest is focused on a localized spatial region. Typical examples are the calculation of dose functions in deep regions of irradiated objects and, in the case of collimated radiation beams, the evaluation of radial doses far from the beam axis. The basic idea of splitting and Russian roulette methods is to favour the flux of radiation towards the region of interest and inhibit the radiation that leaves that region. These techniques are also useful in other problems where only a partial description of the transport process is required. The "region of interest" may then be a limited volume in the space of state variables \((r, E, \mathbf{d})\). Thus, in studies of radiation backscattering, the region of interest may be selected as the spatial region of the sample close to the irradiated surface and the set of particle directions that point towards this surface.

As in the case of interaction forcing, variance reduction is accomplished by modifying the weights of the particles. It is assumed that primary particles start moving with unit weight and each secondary particle produced by a primary one is assigned an initial weight equal to that of the primary. Splitting consists of transforming a particle, with weight \( w_0 \) and in a certain state, into a number \( S > 1 \) of identical particles with weights \( w = w_0/S \) in the same initial state. Splitting should be applied when the particle "approaches" the region of interest. The Russian roulette technique is, in a way, the reverse process: when a particle tends to move away from the region of interest it is
1.6. Variance reduction

“killed” with a certain probability, $K < 1$, and, if it survives, its weight is increased by a factor $1/(1 - K)$. Here, killing means that the particle is just discarded (and does not contribute to the scores anymore). Evidently, splitting and killing leave the simulation unbiased. The mean and standard deviation of the calculated quantities are given by eqs. (1.114) and (1.115). The effectiveness of these methods relies on the adopted values of the parameters $S$ and $K$, and on the strategy used to decide when splitting and killing are to be applied. These details can only be dictated by the user’s experience.

1.6.3 Other methods

Very frequently, an effective “reduction of variance” may be obtained by simply avoiding unnecessary calculations. This is usually true for simulation codes that incorporate “general-purpose” geometry packages. In the case of simple (e.g. planar, spherical, cylindrical) geometries the program may be substantially simplified and this may speed up the simulation appreciably. In general, the clever use of possible symmetries of the problem under consideration may lead to spectacular variance reductions. As a last example, we can quote the so-called “range rejection” method, which simply consists of absorbing a particle when it (and its possible secondaries) cannot leave (or reach) the regions of interest. Range rejection is useful e.g. when computing the total energy deposition of electrons or positrons in a given spatial region. When the residual range of a particle is less than the distance to the nearest limiting surface of the region of interest, the particle will deposit all its energy inside or outside the considered region (depending of its current position) and the simulation of the track can be stopped. Range rejection is not adequate for photon transport simulation, since the concept of photon range is not well defined (or, to be more precise, photon path length fluctuations are very large).
Chapter 2

Photon interactions

In this chapter, we consider the interactions of unpolarized photons of energy $E$ with atoms of atomic number $Z$. We limit our considerations to the energy range from 100 eV up to 1 GeV, where the dominant interaction processes are coherent (Rayleigh) scattering, incoherent (Compton) scattering, the photoelectric effect and electron-positron pair production. Other interactions, such as photonuclear absorption, occur with much smaller probability and can be disregarded for most practical purposes (see e.g. Hubbell et al., 1980).

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure2_1.png}
\caption{Basic interactions of photons with matter.}
\end{figure}
As long as the response of an atom is not appreciably distorted by molecular binding, the single-atom theory can be extended to molecules by using the additivity approximation, i.e. the molecular cross section for a process is approximated by the sum of the atomic cross sections of all the atoms in the molecule. The additivity approximation can also be applied to dense media whenever interference effects between waves scattered by different centres (which, for instance, give rise to Bragg diffraction in crystals) are small. We assume that these conditions are always satisfied.

The ability of Monte Carlo simulation methods to describe photon transport in complex geometries has been established from research during the last five decades (Hayward and Hubbell, 1954; Zerby, 1963; Berger and Seltzer, 1972; Chan and Doi, 1983; Ljungberg and Strand, 1989). The most accurate DCSs available are given in numerical form and, therefore, advanced Monte Carlo codes make use of extensive databases. To reduce the amount of required numerical information, in PENELope we use a combination of analytical DCSs and numerical tables. The adopted DCSs are defined by simple, but physically sound analytical forms. The corresponding total cross sections are obtained by a single numerical quadrature that is performed very quickly, even on a personal computer, using the SUMGA external function described in appendix B. Moreover, the random sampling from these DCSs can be done analytically and, hence, exactly. Only coherent scattering requires a simple preparatory numerical step.

It may be argued that using analytical approximate DCSs, instead of more accurate tabulated DCSs implies a certain loss of accuracy. To minimize this loss, PENELope renormalizes the analytical DCSs so as to reproduce partial attenuation coefficients that are read from the input material data file. As a consequence, the free between events and the kind of interaction are sampled using total cross sections that are nominally exact; approximations are introduced only in the description of individual interaction events.

In the following, \( \kappa \) stands for the photon energy in units of the electron rest energy, i.e.

\[
\kappa \equiv \frac{E}{m_e c^2},
\]  

(2.1)

## 2.1 Coherent (Rayleigh) scattering

Coherent or Rayleigh scattering is the process by which photons are scattered by bound atomic electrons without excitation of the target atom, i.e. the energies of the incident and scattered photons are the same. The scattering is qualified as “coherent” because it arises from the interference between secondary electromagnetic waves coming from different parts of the atomic charge distribution.

The atomic DCS per unit solid angle for coherent scattering is given approximately by (see e.g. Born, 1969)

\[
\frac{d\sigma_{Ra}}{d\Omega} = \frac{d\sigma_{\pi}}{d\Omega} \left[ F(q, Z) \right]^2,
\]  

(2.2)
2.1. Coherent (Rayleigh) scattering

where
\[
\frac{d\sigma_T(\theta)}{d\Omega} = r_e^2 \frac{1 + \cos^2 \theta}{2}
\]
(2.3)
is the classical Thompson DCS for scattering by a free electron at rest, \(\theta\) is the polar scattering angle (see fig. 2.1) and \(F(q, Z)\) is the atomic form factor. The quantity \(r_e\) is the classical electron radius and \(q\) is the magnitude of the momentum transfer given by
\[
q = 2(E/c)\sin(\theta/2) = (E/c)[2(1 - \cos \theta)]^{1/2}.
\]
(2.4)
In the literature on x-ray crystallography, the dimensionless variable
\[
x = \frac{q \, 10^{-8}\, \text{cm}}{4\pi \hbar} = 20.6074 \frac{q}{m_e c}
\]
(2.5)
is normally used instead of \(q\).

The atomic form factor can be expressed as the Fourier transform of the atomic electron density \(\rho(r)\) which, for a spherically symmetrical atom, simplifies to
\[
F(q, Z) = 4\pi \int_0^\infty \rho(r) \sin(qr/\hbar) r^2 \, dr.
\]
(2.6)
\(F(q, Z)\) is a monotonically decreasing function of \(q\) that varies from \(F(0, Z) = Z\) to \(F(\infty, Z) = 0\). The most accurate form factors are those obtained from Hartree-Fock or configuration-interaction atomic-structure calculations; here we adopt the non-relativistic atomic form factors tabulated by Hubbell et al. (1975). Although relativistic form factors are available (Doyle and Turner, 1968), Hubbell has pointed out that the non-relativistic form factors yield results in closer agreement with experiment (Cullen et al., 1997).

In the calculations, we use the following analytical approximation
\[
F(q, Z) = \begin{cases} 
   f(x, Z) \equiv Z \frac{1 + a_1 x^2 + a_2 x^3 + a_3 x^4}{(1 + a_4 x^2 + a_5 x^4)^2}, & \text{if } Z > 10 \text{ and } f(x, Z) < 2, \\
   \max \{ f(x, Z), F_K(q, Z) \} & \text{else}
\end{cases}
\]
(2.7)
where
\[
F_K(q, Z) \equiv \frac{\sin(2b \arctan Q)}{bQ (1 + Q^2)^{5/2}},
\]
(2.8)
with
\[
Q = \frac{q}{2m_e c \alpha}, \quad b = \sqrt{1 - a^2}, \quad a \equiv \alpha(Z - 5/16),
\]
(2.9)
where \(\alpha\) is the fine-structure constant. The function \(F_K(q, Z)\) is the contribution to the atomic form factor due to the two K-shell electrons (see e.g. Baró et al., 1994a). The parameters of expression \(f(x, Z)\) for \(Z = 1\) to 92, which have been determined by Baró et al. (1994a) by numerically fitting the atomic form factors tabulated by Hubbell et al. (1975), are included in the block data subprogram PENDAT. The average relative
Figure 2.2: Atomic form factors for carbon and lead. Crosses are values from the tables of Hubbell et al. (1975), continuous curves represent the analytical approximation given by eq. (2.7).

The total coherent scattering cross section per atom is

$$
\sigma_{Ra} = \frac{d\sigma_{Ra}}{d\Omega} = \pi r^2 \int_{-1}^{1} \left( 1 + \cos^2 \theta \right) \left[ F(q, Z) \right]^2 d(\cos \theta).
$$

(2.10)

Introducing \( q \), eq. (2.4), as a new integration variable, the asymptotic behaviour of the total cross section for small and large photon energies is made clear. For low photon energies, the form factor in the integrand does not depart appreciably from the value \( F(0, Z) = Z \), i.e. coherent scattering reduces to pure Thompson scattering. Consequently, we have

$$
\sigma_{Ra} \simeq \frac{8}{3} \pi r^2 Z^2.
$$

(2.11)

In the high-energy limit, we get

$$
\sigma_{Ra} \propto E^{-2}.
$$

(2.12)

In practice, this limiting behaviour is attained for energies of the order of \( Z/2 \) MeV.

Strictly speaking, expression (2.2) is adequate only for photons with energy well above the K absorption edge. The low-energy behaviour given by eq. (2.11) is substantially altered when anomalous scattering factors are introduced (see e.g. Cullen et al., 1989; Kane et al., 1986). These factors lead to a general decrease of the coherent scattering cross section near the absorption edges and at low energies. Nevertheless, at the energies where anomalous scattering effects become significant, coherent scattering is much less probable than photoelectric absorption (see fig. 2.10 below), and the approximation given by eq. (2.2) is usually sufficient for simulation purposes.
2.1.1 Simulation of coherent scattering events

The PDF of the angular deflection, $\cos \theta$, can be written as [see eqs. (2.2) and (2.3); normalization is irrelevant here]

$$p_{Ra}(\cos \theta) = \frac{1 + \cos^2 \theta}{2} [F(x, Z)]^2,$$  

(2.13)

where $x$, which is defined by eqs. (2.4) and (2.5), can take values in the interval from 0 to

$$x_{\text{max}} = 20.6074 \times 2\kappa.$$  

(2.14)

This PDF can be factorized in the form

$$p_{Ra}(\cos \theta) = g(\cos \theta) \pi(x^2)$$  

(2.15)

with

$$g(\cos \theta) \equiv \frac{1 + \cos^2 \theta}{2} \quad \text{and} \quad \pi(x^2) \equiv [F(x, Z)]^2.$$  

(2.16)

Notice that, for a compound, $[F(x, Z)]^2$ has to be replaced by the sum of squared form factors of the atoms in the molecule.

The function $\pi(x^2)$ can be considered as the (unnormalized) PDF of the variable $x^2$. Random values of $x^2$ distributed according to this PDF can be generated by the inverse transform method (section 1.2.2), i.e. from the sampling equation

$$\int_0^{x^2} \pi(x'^2) \, dx'^2 = \xi \int_0^{x_{\text{max}}^2} \pi(x'^2) \, dx'^2.$$  

(2.17)

It is convenient to introduce the function

$$\Pi(x^2) = \int_0^{x^2} \pi(x'^2) \, dx'^2,$$  

(2.18)

which increases monotonically with $x^2$ and saturates for high $x^2$-values to a constant finite value. Then, the sampling equation (2.17) can be written in the form

$$\Pi(x^2) = \xi \Pi(x_{\text{max}}^2),$$  

(2.19)

which is easy to solve numerically. To this end, we only need to have a table of values of the function $\Pi(x^2)$ stored in memory. For a given photon energy, $\Pi(x_{\text{max}}^2)$ can be evaluated by interpolation in this table. Linear log-log interpolation (extrapolation) in a table with about 210 points logarithmically distributed in the interval $$(10^{-4}, 10^4)$$ yields results which are accurate to within 0.01% (notice that in the interval from 0 to $10^{-4}$, $F(x, Z) \simeq Z$ and, hence, $\Pi(x^2)$ is proportional to $x^2$, i.e. extrapolation for $x^2 < 10^{-4}$ is exact). The value

$$x^2 = \Pi^{-1}\left(\xi \Pi(x_{\text{max}}^2)\right),$$  

(2.20)

can then be obtained by inverse linear interpolation (or extrapolation) with a binary search.
The angular deflection $\cos \theta$ can now be sampled by the rejection method (section 1.2.4), since the function $g(\cos \theta)$ is a valid rejection function (i.e., it is positive and less than or equal to unity). The algorithm for sampling $\cos \theta$ proceeds as follows:

(i) Compute $\Pi(x^2_{\text{max}})$.

(ii) Generate a random number $\xi$ and determine $x^2$ using eq. (2.20). Set

$$\cos \theta = 1 - \frac{1}{2} \frac{x^2}{(20.6074 \kappa)^2}. \quad (2.21)$$

(iii) Generate a new random number $\xi$.

(iv) If $\xi > g(\cos \theta)$, go to step (ii).

(v) Deliver $\cos \theta$.

Although numerical interpolation is necessary, it is performed on a single function that is independent of the photon energy and the errors introduced are negligible. It is worth noting that the sampling algorithm is essentially independent of the adopted form factor, and directly applicable to molecules. The advantage of using the analytical form factor, eq. (2.7), instead of a numerical database is that $\Pi(x^2)$ can be easily calculated to the desired accuracy, using the SUMGA integration function (appendix B).

The efficiency of the sampling method (i.e., the fraction of generated values of $\cos \theta$ that is accepted) increases with photon energy. At low energies, it equals $2/3$ (exactly) for all elements. For $E = 100$ keV, the efficiencies for hydrogen and uranium are 100% and 86%, respectively.

2.2 Photoelectric effect

In the photoelectric effect, a photon of energy $E$ is absorbed by the target atom, which makes a transition to an excited state. The photon beams found in radiation transport studies have relatively low photon densities and, as a consequence, only single-photon absorption is observed\(^1\). To represent the atomic states, we can adopt an independent-electron model, such as the Dirac-Hartree-Fock-Slater self-consistent model (see e.g. Pratt et al., 1973), in which each electron occupies a single-particle orbital, with well-defined ionization energy. The set of orbitals with the same principal and total angular momentum quantum numbers and the same parity constitute a shell. Each shell $i$ can accommodate a finite number of electrons, with characteristic ionization energy $U_i$. Notice that the shell ionization energies are positive, the quantity $-U_i$ represents the “binding” energy of each individual electron. Fig. 2.3 (left diagram) shows the various notations used to designate the innermost atomic electron shells (i.e., those with the

\(^1\)In intense low-energy photon beams, such as those from high-power lasers, simultaneous absorption of several photons is possible.
2.2. Photoelectric effect

Figure 2.3: Various notations for inner atomic electron shells (left) and allowed radiative transitions (right) to these shells. Transitions different from the ones indicated in the diagram (e.g. K-M4) are also possible, but their transition probabilities are extremely small.

Figure 2.4: Ionization energies of the innermost shells of free atoms, as given by Lederer and Shirley (1978).
largest ionization energies) as well as their ordering in energy and allowed occupancies. In our simulations, we use the experimental ionization energies given by Lederer and Shirley (1978), which pertain to free, neutral atoms. Ionization energies of K-, L- and M-shells are displayed in fig. 2.4.

Considering the interaction with the photon field as a first-order perturbation (which is appropriate for fields with low photon densities) it follows that only one-electron transitions are allowed. That is, in the photoelectric effect, the photon is absorbed by an individual electron in the “active” shell $i$, which leaves the parent atom with kinetic energy $E_\text{c} = E - U_i$. Evidently, photoionization of a given shell is only possible when the photon energy exceeds the corresponding ionization energy; this gives rise to the characteristic absorption edges in the photoelectric cross section (see fig. 2.5).

The photoelectric cross sections used in PENEOPE are obtained by interpolation in a numerical table that was extracted from the LLNL Evaluated Photon Data Library (EPDL; Cullen et al., 1997). This library contains photoelectric cross sections for all shells of the elements $Z = 1-100$ and photon energies from 1 eV to 1000 GeV, derived from Scofield’s theoretical calculations of shell cross sections (Saloman et al., 1988) and Hubbell’s total cross sections (Hubbell et al., 1980; Berger and Hubbell, 1987). The PENEOPE database for photoelectric absorption (a subset of the EPDL) consists of tables of the total atomic cross section $\sigma_{\text{ph}}(E)$ and the cross sections for the K- and L-shells, $\sigma_{\text{ph},i}(E)$ ($i = \text{K, L1, L2 and L3}$) for the elements $Z = 1-92$, which span the energy range from 100 eV to 1000 GeV. These tables are estimated to be accurate to within a few percent for photon energies above 1 keV (Cullen et al., 1997). At lower energies, uncertainties in the data are much larger: 10–20% for 0.5 keV < $E$ < 1 keV and 100–200% for 0.1 keV < $E$ < 0.5 keV. Notice that the cross sections in the EPDL are based on free-atom theoretical calculations and, therefore, near-edge absorption structures produced by molecular or crystalline ordering (e.g. extended x-ray absorption fine-structure) are ignored.

For compound materials (and also for mixtures) the molecular cross section $\sigma_{\text{ph}}(E)$ is evaluated by means of the additivity approximation, that is, as the sum of the atomic cross sections of the elements involved. In the energy range between successive absorption edges, the photoelectric cross section is a continuous function of the photon energy (see fig. 2.5). In PENEOPE, the molecular cross section is defined by means of a table of numerical values $\sigma_{\text{ph}}(E_i)$ for a logarithmic grid of energies $E_i$, which is stored in memory. Photon mean free paths are determined by linear log-log interpolation in this table. Knowledge of the atomic cross sections is needed, only when a photoabsorption event has effectively occurred, to select the element that has been ionized (whose probability is proportional to the atomic cross section).

### 2.2.1 Simulation of photoelectron emission

Let us consider that a photon with energy $E$ is absorbed by an atom of the element $Z$. The “active” shell $i$ that is ionized is considered as a discrete random variable with
2.2. Photoelectric effect

![Graph: Atomic photoelectric cross sections for carbon, iron, and uranium as functions of the photon energy E.]

**Figure 2.5:** Atomic photoelectric cross sections for carbon, iron, and uranium as functions of the photon energy $E$.

\[ p_i = \frac{\sigma_{ph,i}(Z, E)}{\sigma_{ph}(Z, E)}, \]  \hspace{1cm} \text{(2.22)}

where $\sigma_{ph,i}(Z, E)$ is the cross section for ionization of shell $i$ and $\sigma_{ph}(Z, E)$ is the total photoelectric cross section of the atom. PENELOPE incorporates a detailed description of photoabsorption in K- and L-shells (including the subsequent atomic relaxation). The ionization probabilities of these inner shells are determined from the corresponding partial cross sections. The probability of ionization in an outer shell is obtained as

\[ p_{outer} = 1 - (p_K + p_{L1} + p_{L2} + p_{L3}). \] \hspace{1cm} \text{(2.23)}

When the ionization occurs in an inner K- or L-shell, the initial energy of the photoelectron is set equal to $E_e = E - U_i$; the residual atom, with a vacancy in the shell, subsequently relaxes to its ground state by emitting x-rays and Auger electrons. If the ionization occurs in an outer shell, we assume that the photoelectron leaves the target atom with kinetic energy equal to the energy deposited by the photon, $E_e = E$, and we disregard the emission of subsidiary fluorescent radiation (see section 2.6).

**Initial direction of photoelectrons**

The direction of emission of the photoelectron, relative to that of the absorbed photon, is defined by the polar and azimuthal angles $\theta_e$ (fig. 2.1) and $\phi_e$. We consider that the incident photon is not polarized and, hence, the angular distribution of photoelectrons is independent of $\phi_e$, which is uniformly distributed in the interval $(0, 2\pi)$. The polar angle $\theta_e$ is sampled from the K-shell cross section derived by Sauter (1931) using K-shell
hydrogenic electron wave functions. The Sauter DCS (per electron) can be written as
\[
\frac{d\sigma_{\text{ph}}}{d\Omega} = \alpha^4 r_e^2 \left( \frac{Z}{\kappa} \right)^5 \frac{\beta^3}{\gamma(1 - \beta \cos \theta_e)^4} \left[ 1 + \frac{1}{2} \gamma(\gamma - 1)(\gamma - 2)(1 - \beta \cos \theta_e) \right],
\]
where \(\alpha\) is the fine-structure constant, \(r_e\) is the classical electron radius, and
\[
\gamma = 1 + E_e/(m_e c^2), \quad \beta = \sqrt{E_e(E_e + 2m_e c^2)} / E_e + m_e c^2.
\]
Strictly speaking, the DCS (2.24) is adequate only for ionization of the K-shell by high-energy photons. Nevertheless, in many practical simulations no appreciable errors are introduced when Sauter’s distribution is used to describe any photoionization event, irrespective of the atomic shell and the photon energy. The main reason is that the emitted photoelectron immediately starts to interact with the medium, and its direction of movement is strongly altered after travelling a path length much shorter than the photon mean free path. On the other hand, when the photon energy exceeds the K-edge, most of the ionizations occur in the K-shell and then the Sauter distribution represents a good approximation.

Introducing the variable \(\nu = 1 - \cos \theta_e\), the angular distribution of photoelectrons can be expressed in the form
\[
p(\nu) = (2 - \nu) \left[ \frac{1}{A + \nu} + \frac{1}{2} \beta \gamma(\gamma - 1)(\gamma - 2) \right] \frac{\nu}{(A + \nu)^3}, \quad A = \frac{1}{\beta} - 1,
\]
apart from a normalization constant. Random sampling of \(\nu\) from this distribution can be performed analytically. To this end, \(p(\nu)\) can be factorized in the form
\[
p(\nu) = g(\nu) \pi(\nu)
\]
with
\[
g(\nu) = (2 - \nu) \left[ \frac{1}{A + \nu} + \frac{1}{2} \beta \gamma(\gamma - 1)(\gamma - 2) \right]
\]
and
\[
\pi(\nu) = \frac{A(A + 2)^2}{2} \frac{\nu}{(A + \nu)^3}.
\]

The variable \(\nu\) takes values in the interval \((0,2)\), where the function \(g(\nu)\) is definite positive and attains its maximum value at \(\nu = 0\), while the function \(\pi(\nu)\) is positive and normalized to unity. Random values from the probability distribution \(\pi(\nu)\) are generated by means of the sampling formula (inverse transform method, see section 1.2.2)
\[
\int_0^\nu \pi(\nu') \, d\nu' = \xi,
\]
which can be solved analytically to give
\[
\nu = \frac{2A}{(A + 2)^2 - 4\xi} \left[ 2\xi + (A + 2)\xi^{1/2} \right].
\]
Therefore, random sampling from Sauter’s distribution can be performed by the rejection method (see section 1.2.4) as follows:
23. Incoherent (Compton) scattering

(i) Generate $\nu$ from $\pi(\nu)$ by using eq. (2.31).

(ii) Generate a random number $\xi$.

(iii) If $\xi g(0) > g(\nu)$, go to step (i).

(iv) Deliver $\cos \theta_e = 1 - \nu$.

The efficiency of this algorithm is $\sim 0.33$ at low energies and increases slowly with $E_e$; for $E_e = 1$ MeV, the efficiency is 0.4. As photoelectric absorption occurs at most once in each photon history, this small sampling efficiency does not slow down the simulation significantly.

2.3 Incoherent (Compton) scattering

In Compton scattering, a photon of energy $E$ interacts with an atomic electron, which absorbs it and re-emits a secondary (Compton) photon of energy $E'$ in the direction $\Omega = (\theta, \phi)$ relative to the direction of the original photon. In PENelope, Compton scattering events are described by means of the cross section obtained from the relativistic impulse approximation (Ribberfors, 1983). Contributions from different atomic electron shells are considered separately. After a Compton interaction with the $i$-th shell, the active target electron is ejected to a free state with kinetic energy $E_e = E - E' - U_i > 0$, where $U_i$ is the ionization energy of the considered shell, and the residual atom is left in an excited state with a vacancy in the $i$-th shell.

In the case of scattering by free electrons at rest, the conservation of energy and momentum implies the following relation between the energy $E'$ of the scattered (Compton) photon and the scattering angle $\theta$ [cf. eq. (A.19)]

$$E' = \frac{E}{1 + \kappa(1 - \cos \theta)} \equiv E_C,$$  

where $\kappa = E/m_e c^2$, as before. The DCS for Compton scattering by a free electron at rest is given by the familiar Klein-Nishina formula,

$$\frac{d\sigma_{KN}}{d\Omega} = \frac{r_e^2}{2} \left( \frac{E_C}{E} \right)^2 \left( \frac{E}{E_C} + \frac{E}{E_C} - \sin^2 \theta \right).$$  

Although this simple DCS was generally used in old Monte Carlo transport codes, it represents only a rough approximation for the Compton interactions of photons with atoms. In reality, atomic electrons are not at rest, but move with a certain momentum distribution, which gives rise to the so-called Doppler broadening of the Compton line. Moreover, transitions of bound electrons are allowed only if the energy transfer $E - E'$ is larger than the ionization energy $U_i$ of the active shell (binding effect).

The impulse approximation accounts for Doppler broadening and binding effects in a natural, and relatively simple, way. The DCS is obtained by considering that
electrons in the $i$-th shell move with a momentum distribution $\rho_i(\mathbf{p})$. For an electron in an orbital $\psi_i(\mathbf{r})$, $\rho_i(\mathbf{p}) \equiv |\psi_i(\mathbf{p})|^2$, where $\psi_i(\mathbf{p})$ is the wave function in the momentum representation. The DCS for Compton scattering by an electron with momentum $\mathbf{p}$ is derived from the Klein-Nishina formula by applying a Lorentz transformation with velocity $\mathbf{v}$ equal to that of the moving target electron. The impulse approximation to the Compton DCS (per electron) of the considered shell is obtained by averaging over the momentum distribution $\rho_i(\mathbf{p})$.

After some manipulations, the Compton DCS of an electron in the $i$-th shell can be expressed as [eq. (21) in Brusa et al., 1996]

$$\frac{d^2\sigma_{\text{Co},i}}{dE'd\Omega} = \frac{r_e^2}{2} \left( \frac{E_C}{E} \right)^2 \left( \frac{E}{E_C} + \frac{E'}{E_C} - \sin^2 \theta \right) F(p_z) J_i(p_z) \frac{dp_z}{dE'},$$

(2.34)

where $r_e$ is the classical electron radius. $E_C$ is the energy of the Compton line, defined by eq. (2.32), i.e. the energy of photons scattered in the direction $\theta$ by free electrons at rest. The momentum transfer vector is given by $\mathbf{q} \equiv \hbar \mathbf{k} - \hbar \mathbf{k}'$, where $\hbar \mathbf{k}$ and $\hbar \mathbf{k}'$ are the momenta of the incident and scattered photons; its magnitude is

$$q = \frac{1}{c} \sqrt{E'^2 + E'^2 - 2EE' \cos \theta}.$$  

(2.35)

The quantity $p_z$ is the projection of the initial momentum $\mathbf{p}$ of the electron on the direction of the scattering vector $\hbar \mathbf{k}' - \hbar \mathbf{k} = -\mathbf{q}$; it is given by

$$p_z = -\frac{\mathbf{p} \cdot \mathbf{q}}{q} = \frac{EE'(1 - \cos \theta) - m_e c^2 (E - E')}{c^2 q}$$

(2.36)

or, equivalently,

$$\frac{p_z}{m_e c} = \frac{E(E' - E_C)}{E_C c q}.$$  

(2.37)

Notice that $p_z = 0$ for $E' = E_C$. Moreover,

$$\frac{dp_z}{dE'} = \frac{m_e c}{E_C q} \left( \frac{E}{E_C} + \frac{E \cos \theta - E'}{E_C c q} \frac{p_z}{m_e c} \right).$$  

(2.38)

The function $J_i(p_z)$ in eq. (2.34) is the one-electron Compton profile of the active shell, which is defined as

$$J_i(p_z) \equiv \int \rho_i(\mathbf{p}) dp_x dp_y,$$

(2.39)

where $\rho_i(\mathbf{p})$ is the electron momentum distribution. That is, $J_i(p_z) \, dp_z$ gives the probability that the component of the electron momentum in the $z$-direction is in the interval $(p_z, p_z + dp_z)$. Notice that the normalization

$$\int_{-\infty}^{\infty} J_i(p_z) \, dp_z = 1.$$  

(2.40)

---

2The expression (2.36) contains an approximation, the exact relation is obtained by replacing the electron rest energy $m_e c^2$ in the numerator by the electron initial total energy, $\sqrt{(m_e c^2)^2 + (cE)^2}$. 

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2.3. Incoherent (Compton) scattering

Figure 2.6: Atomic Compton profiles \( p_z > 0 \) for aluminium, copper and gold. The continuous curves are numerical Hartree-Fock profiles tabulated by Biggs et al. (1975). The dashed curves represent the analytical profiles defined by eq. (2.57). (Adapted from Brusa et al., 1996.)

is assumed. In the Hartree-Fock approximation for closed-shell configurations, the momentum distribution of the electrons in an atomic shell, obtained by adding the contributions of the orbitals in that shell, is isotropic. For an isotropic distribution, expression (2.39) simplifies to

\[
J_i(p_z) = 2\pi \int_{|p_z|}^{\infty} p \rho_i(p) \, dp.
\]  

(2.41)

The atomic Compton profile is given by

\[
J(p_z) = \sum_i f_i J_i(p_z),
\]

(2.42)

where \( f_i \) is the number of electrons in the \( i \)-th shell and \( J_i(p_z) \) is the one-electron profile of this shell. The functions \( J(p_z) \) and \( J_i(p_z) \) are both bell-shaped and symmetrical about \( p_z = 0 \) (see fig. 2.6). Extensive tables of Hartree-Fock Compton profiles for the elements have been published by Biggs et al. (1975). These numerical profiles are adequate for bound electron shells. In the case of conductors, the one-electron Compton profile for conduction electrons may be estimated by assuming that these form a free-electron gas with \( \rho_e \) electrons per unit volume. The one-electron profile for this system is (see e.g. Cooper, 1971)

\[
J_i^{\text{Fre}}(p_z) = \frac{3}{4p_F} \left( 1 - \frac{p_z^2}{p_F^2} \right) \Theta(p_F - |p_z|), \quad J_i^{\text{Fre}}(0) = \frac{3}{4p_F}.
\]  

(2.43)

where \( p_F \equiv \hbar(3\pi^2 \rho_e)^{1/3} \) is the Fermi momentum. For scattering in a compound material,
the molecular Compton profile is obtained as the sum of atomic profiles of the atoms in a molecule (additivity rule).

\[ F(p_z) \approx 1 + \frac{e q_C}{E} \left( 1 + \frac{E_C (E_C - E \cos \theta)}{(e q_C)^2} \right) \frac{p_z}{m_e c}, \]  

(2.44)

where \( q_C \) is the momentum transfer associated with the energy \( E' = E_C \) of the Compton line,

\[ q_C \equiv \frac{1}{c} \sqrt{E^2 + E_C^2 - 2 E E_C \cos \theta}. \]  

(2.45)

Expression (2.44) is accurate only for small \( |p_z| \) values. For large \( |p_z| \), \( J_1(p_z) \) tends to zero and the factor \( F(p_z) \) has no effect on the DCS. We use the values given by expression (2.44) only for \( |p_z| < 0.2 m_e c \) and take \( F(\pm |p_z|) = F(\pm 0.2 m_e c) \) for \( |p_z| > 0.2 m_e c \). Owing to the approximations introduced, negative values of \( F \) may be obtained for large \( |p_z| \); in this case, we must set \( F = 0 \).

We can now introduce the effect of electron binding: Compton excitations are allowed only if the target electron is promoted to a free state, i.e. if the energy transfer \( E - E' \)
is larger than the ionization energy \( U_i \) of the active shell. Therefore the atomic DCS, including Doppler broadening and binding effects, is given by

\[
\frac{d^2\sigma_{\text{Co}}}{dE'd\Omega} = \frac{r_e^2}{2} \left( \frac{E_C}{E} \right)^2 \left( \frac{E_C}{E} + \frac{E}{E_C} - \sin^2 \theta \right) \times F(p_z) \left( \sum_i f_i J_i(p_z) \Theta(E - E' - U_i) \right) \frac{dp_z}{dE'},
\]

(2.46)

where \( \Theta(x) \) (= 1 if \( x > 0 \), = 0 otherwise) is the Heaviside step function. In the calculations we use the ionization energies \( U_i \) given by Lederer and Shirley (1978), fig. 2.4. The DCS for scattering of 10 keV photons by aluminium atoms is displayed in fig. 2.7, for \( \theta = 60 \) and 180 degrees, as a function of the fractional energy of the emerging photon. The DCS for a given scattering angle has a maximum at \( E' = E_C \); its shape resembles that of the atomic Compton profile, except for the occurrence of edges at \( E' = E - U_i \).

In the case of scattering by free electrons at rest we have \( U_i = 0 \) (no binding) and \( J_i(p_z) = \delta(p_z) \) (no Doppler broadening). Moreover, from eq. (2.37) \( E' = E_C \), so that photons scattered through an angle \( \theta \) have energy \( E_C \). Integration of the DCS, eq. (2.46), over \( E' \) then yields the familiar Klein-Nishina cross section,

\[
\frac{d\sigma_{\text{Co}}^{\text{KN}}}{d\Omega} = Z r_e^2 \frac{E_C}{E} \left( \frac{E_C}{E} + \frac{E}{E_C} - \sin^2 \theta \right),
\]

(2.47)

for the \( Z \) atomic electrons [cf. eq. (2.33)]. For energies of the order of a few MeV and larger, Doppler broadening and binding effects are relatively small and the free-electron theory yields results practically equivalent to those of the impulse approximation.

The angular distribution of scattered photons is given by the directional DCS,

\[
\frac{d\sigma_{\text{Co}}}{d\Omega} = \int d^2\sigma_{\text{Co}} \frac{dE'}{d\Omega} dE' = \frac{r_e^2}{2} \left( \frac{E_C}{E} \right)^2 \left( \frac{E_C}{E} + \frac{E}{E_C} - \sin^2 \theta \right) \times \sum_i f_i \Theta(E - U_i) \int_{-\infty}^{p_{i,max}} F(p_z)J_i(p_z) \frac{dp_z}{dE'},
\]

(2.48)

where \( p_{i,max} \) is the highest \( p_z \)-value for which an electron in the \( i \)-th shell can be excited. It is obtained from eq. (2.36) by setting \( E' = E - U_i \),

\[
p_{i,max}(E,\theta) = \frac{E(E - U_i)(1 - \cos \theta) - m_e c^2 U_i}{c \sqrt{2E(E - U_i)(1 - \cos \theta) + U_i^2}}.
\]

(2.49)

Except for energies just above the shell ionization threshold, the function \( F(p_z) \) in the integral can be replaced by unity, since \( p_z J_i(p_z) \) is an odd function and its integral is close to zero, i.e.

\[
\int_{-\infty}^{p_{i,max}} F(p_z)J_i(p_z) \frac{dp_z}{dE'} \simeq n_i(p_{i,max}),
\]

(2.50)
where
\[ n_i(p_z) = \int_{-\infty}^{p_z} J_i(p_z') \, dp_z'. \]  
(2.51)

Notice that \( n_i(p_z) \) is a monotonously increasing function of \( p_z \), which varies from 0 at \( p_z = -\infty \) to unity at \( p_z = \infty \); the quantity \( n_i(p_{i,\text{max}}) \) represents the fraction of electrons in the \( i \)-th shell that can be effectively excited in a Compton interaction. We can then write
\[ \frac{d\sigma_{\text{co}}}{d\Omega} \propto r_e^2 \left( \frac{E_C}{E} \right)^2 \left( \frac{E_C}{E} + \frac{E}{E_C} - \sin^2 \theta \right) S(E, \theta). \]  
(2.52)

The function
\[ S(E, \theta) = \sum_i f_i \Theta(E - U_i) n_i(p_{i,\text{max}}) \]  
(2.53)

can be identified with the incoherent scattering function in the impulse approximation (see e.g. Ribberfors and Berggren, 1982). The total cross section can then be obtained as
\[ \sigma_{\text{co}} = 2\pi \int \frac{d\sigma_{\text{co}}}{d\Omega} \, d(\cos \theta). \]  
(2.54)

For comparison purposes, and also to calculate the energy deposition, it is useful to consider the cross section differential in only the energy of the scattered photon,
\[ \frac{d\sigma_{\text{co}}}{dE'} = \int \frac{d^2\sigma_{\text{co}}}{dE' d\Omega} \, d\Omega. \]  
(2.55)

In the case of scattering by free electrons at rest, \( E' = E_C \) and the Klein-Nishina formula (2.47) gives the following expression for the energy DCS,
\[ \frac{d\sigma_{\text{co}}^{\text{KN}}}{dE'} = 2\pi \frac{d\sigma_{\text{co}}^{\text{KN}}}{d\Omega} \, d(\cos \theta) \frac{d\Omega}{dE_C} \]
\[ = \frac{\pi r_e^2}{E} \kappa^{-3} \left( \frac{E^2}{E'^2} + \frac{(\kappa^2 - 2\kappa - 2)E}{E'} + (2\kappa + 1) \frac{\kappa^2 E'}{E} \right). \]  
(2.56)

Fig. 2.8 displays the energy DCSs obtained from this formula and from the impulse approximation for scattering of high-energy \((E > U_i)\) photons by aluminium and gold atoms. These results show clearly the differences between the physics of the impulse approximation and the cruder free-electron approximation. The most conspicuous feature of the impulse approximation DCS is the absence of a threshold energy, which is a direct manifestation of the Doppler broadening. For relatively small energy transfers \((E' \sim E)\) the Klein-Nishina DCS increases with the energy of the scattered photon, whereas the energy DCS obtained from the impulse approximation vanishes at \( E' = E \) due to the effect of binding, which also causes the characteristic edge structure, similar to that of the photoelectric cross section (see fig. 2.8).
2.3. Incoherent (Compton) scattering

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure2_8}\hspace{1cm}
\caption{Energy DCSs for Compton scattering of 50 and 500 keV photons by aluminium and gold atoms. The continuous curves represent the DCS (2.55), computed using the analytical Compton profiles (2.57). The dashed curves are obtained from the Klein-Nishina formula (2.56), i.e. assuming that the atomic electrons are free and at rest.}
\end{figure}

2.3.1 Analytical Compton profiles

In order to minimize the required numerical information and to simplify the random sampling, we use approximate one-electron profiles of the form

\[ J_i^A(p_z) = J_{i,0} \frac{nd_2}{2} \left( d_1 + d_2 J_{i,0} |p_z| \right)^{n-1} \exp \left[ d_1^n - \left( d_1 + d_2 J_{i,0} |p_z| \right)^n \right] \]

(2.57)

with

\[ n = 2, \quad d_1 = \frac{(n-1)^{1/n}}{n}, \quad d_2 = \frac{2^n}{n} d_1^{1-n} = \sqrt{2}. \]

The quantity \( J_{i,0} \equiv J_i(0) \) is the value of the profile at \( p_z = 0 \) obtained from the Hartree-Fock orbital (Biggs et al., 1975). \( J_i(0) \) is tabulated in the file PDATCONF.TAB for all shells of the elements \( Z = 1 \) to 92. Notice that \( J_i^A(p_z) \) is normalized according to eq.
(2.40). With the profiles (2.57),
\[
\begin{align*}
n^A_i(p_z) & \equiv \int_{-\infty}^{p_z} J^A_i(p'_z) \, dp'_z = \begin{cases} 
\frac{1}{2} \exp \left[ d_1^2 - \left( d_1 - d_2 J_{i,0} p_z \right)^2 \right] & \text{if } p_z < 0, \\
1 - \frac{1}{2} \exp \left[ d_1^2 - \left( d_1 + d_2 J_{i,0} p_z \right)^2 \right] & \text{if } p_z > 0.
\end{cases}
\end{align*}
\]

Thus, the incoherent scattering function (2.53) can be expressed analytically and the integral (2.54) evaluated very quickly with the aid of function \texttt{SUMP5} (appendix B). On the other hand, the sampling equation \( n^A_i(p_z) \equiv \xi n^A_i(p_{i,\text{max}}) \) (see section 1.2.2) can be solved analytically,
\[
p_z = \begin{cases} 
\frac{d_1}{d_2 J_{i,0}} \left[ d_1 - \left( d_1^2 - \ln 2A \right)^{1/2} \right] & \text{if } A < \frac{1}{2}, \\
\frac{d_1}{d_2 J_{i,0}} \left[ \left( d_1^2 - \ln 2(1 - A) \right)^{1/2} - d_1 \right] & \text{if } A > \frac{1}{2},
\end{cases}
\]

where \( A \equiv \xi n^A_i(p_{i,\text{max}}) \). Atomic Compton profiles obtained from the approximation given by eq. (2.57) are accurate for small \( p_z \) and oscillate about the Hartree-Fock values for intermediate momenta (see fig. 2.6). The relative differences are normally less than \( 5\% \), except for large momenta for which \( J(p_z) \) is very small. Similar differences are found between the DCS computed from Hartree-Fock and analytical Compton profiles (see fig. 2.7). For most applications (e.g. studies of detector response, dosimetry, radiotherapy, etc.), the effect of these differences on the simulation results is not important. The impulse approximation with the analytical one-electron profiles (2.57) then provides a conveniently simple method to introduce Doppler broadening and binding effects in the simulation of Compton scattering.

### 2.3.2 Simulation of incoherent scattering events

Compton events are simulated on the basis of the DCS given by eq. (2.46) with the analytical Compton profiles (2.57). The sampling algorithm adopted here is due to Brusa et al. (1996). It is similar to the one described by Namito et al. (1994), but has a higher efficiency.

The PDF of the polar deflection \( \cos \theta \) and the energy \( E' \) of the scattered photon is given by (apart from normalization constants, which are irrelevant here)

\[
P_{C_0}(\cos \theta, E') = \left( \frac{E_C}{E} \right)^2 \left( \frac{E_C}{E} + \frac{E}{E_C} - \sin^2 \theta \right) \\
\times F(p_z) \left( \sum_i f_i J_i(p_z) \Theta(E - E' - U_i) \right) \frac{dp_z}{dE'}.
\]

Integration of expression (2.60) over \( E' \), using the approximation (2.50), yields the PDF of the polar deflection

\[
P_\theta(\cos \theta) = \left( \frac{E_C}{E} \right)^2 \left( \frac{E_C}{E} + \frac{E}{E_C} - \sin^2 \theta \right) S(E, \theta),
\]
23. Incoherent (Compton) scattering

where $S(E, \theta)$ is the incoherent scattering function, eq. (2.53).

Random values of $\cos \theta$ from the PDF (2.61) can be generated by using the following algorithm (Baró et al., 1994a). Let us introduce the quantity

$$\tau \equiv \frac{E_C}{E} = \frac{1}{1 + \kappa(1 - \cos \theta)}.$$  \hspace{1cm} (2.62)

The minimum and maximum values of $\tau$ are

$$\tau_{\text{min}} = \frac{1}{1 + 2\kappa} \quad \text{and} \quad \tau_{\text{max}} = 1,$$  \hspace{1cm} (2.63)

which correspond to backward ($\theta = \pi$) and forward ($\theta = 0$) scattering, respectively. The PDF of this variable is (again ignoring normalization constants)

$$P_\tau(\tau) = P_\theta(\cos \theta) \frac{d(\cos \theta)}{d\tau} = \left( \frac{1}{\tau^2} + \frac{\kappa^2 - 2\kappa - 2}{\tau} + (2\kappa + 1) + \kappa^2 \tau \right) S(E, \theta).$$  \hspace{1cm} (2.64)

This distribution can be rewritten in the form (Nelson et al., 1985)

$$P_\tau(\tau) = [a_1 P_1(\tau) + a_2 P_2(\tau)] T(\cos \theta),$$  \hspace{1cm} (2.65)

where

$$a_1 = \ln(1 + 2\kappa), \quad a_2 = \frac{2\kappa(1 + \kappa)}{(1 + 2\kappa)^2},$$  \hspace{1cm} (2.66)

$$P_1(\tau) = \frac{1}{\ln(1 + 2\kappa)} \cdot \frac{1}{\tau}, \quad P_2(\tau) = \frac{(1 + 2\kappa)^2}{2\kappa(1 + \kappa)} \tau$$  \hspace{1cm} (2.67)

and

$$T(\cos \theta) = \left\{ 1 - \frac{(1 - \tau)[(2\kappa + 1)\tau - 1]}{\kappa^2 \tau(1 + \tau^2)} \right\} \frac{S(E, \theta)}{S(E, \theta = \pi)},$$  \hspace{1cm} (2.68)

The function in braces is positive, it equals 1 at the end points of the interval $(\tau_{\text{min}}, 1)$, and is less than unity inside this interval. Moreover, the ratio of incoherent scattering functions is also less than unity for any value of $\theta < \pi$. Hence, the function $T(\cos \theta)$ is a valid rejection function. The functions $P_i(\tau) (i = 1, 2)$ are normalized PDFs in the interval $(\tau_{\text{min}}, 1)$, which can be easily sampled by using the inverse transform method. The generation of random values of $\tau$ according to the PDF given by eq. (2.64) can then be performed by combining the composition and rejection methods (section 1.2). The algorithm to sample $\cos \theta$ proceeds as follows:

(i) Sample a value of the integer $i (=1, 2)$ according to the point probabilities

$$\pi(1) = \frac{a_1}{a_1 + a_2} \quad \text{and} \quad \pi(2) = \frac{a_2}{a_1 + a_2},$$  \hspace{1cm} (2.69)
(ii) Sample $\tau$ from $P_i(\tau)$ using the sampling formulae

$$
\tau = \begin{cases} 
\tau_{i,\text{min}} & \text{if } i = 1, \\
[\tau_{i,\text{min}}^2 + \xi (1 - \tau_{i,\text{min}}^2)]^{1/2} & \text{if } i = 2,
\end{cases}
$$

(2.70)

which can be easily derived by the inverse transform method (section 1.2.2).

(iii) Determine $\cos \theta$ using eq. (2.62),

$$
\cos \theta = 1 - \frac{1 - \tau}{\kappa \tau},
$$

(2.71)

and compute the quantities $p_{i,\text{max}}(E, \theta)$, eq. (2.49), and

$$
S(E, \theta) = \sum_i f_i \Theta(E - U_i) n_i^A(p_{i,\text{max}}).
$$

(2.72)

(iv) Generate a new random number $\xi$.

(v) If $\xi > T(\cos \theta)$, go to step (i).

(vi) Deliver $\cos \theta$.

The efficiency of this algorithm, i.e. the probability of accepting a generated $\cos \theta$-value, increases monotonically with photon energy and is nearly independent of $Z$; typical values are 35%, 80% and 95% for $E = 1$ keV, 1 MeV and 10 MeV, respectively.

Once the direction of the emerging photon has been set, the active electron shell $i$ is selected with relative probability equal to $Z_i \Theta(E - U_i) n_i^A(p_{i,\text{max}}(E, \theta))$. A random value of $p_z$ is generated from the analytical Compton profile (2.57) using the sampling formula (2.59). If $p_z$ is less than $-m_e c$, it is rejected and a new shell and a $p_z$-value are sampled$^3$. Finally, the factor $F(p_z)$ in the PDF (2.46) is accounted for by means of a rejection procedure. It should be noted that the approximation $F \simeq 1$ is valid only when the DCS is integrated over $E'$; otherwise the complete expression (2.44) must be used. Let $F_{\text{max}}$ denote the maximum value of $F(p_z)$, which occurs at $p_z = 0.2m_e c$ or $-0.2m_e c$; a random number $\xi$ is generated and the value $p_z$ is accepted if $\xi F_{\text{max}} < F(p_z)$, otherwise the process of selecting a shell and a $p_z$-value is reinitiated. The energy $E'$ of the emerging photon is then calculated from eq. (2.36), which gives

$$
E' = E \frac{\tau}{1 - t \tau^2} \left[ (1 - t \tau \cos \theta) + \text{sign}(p_z) \sqrt{(1 - t \tau \cos \theta)^2 - (1 - t \tau^2)(1 - t)} \right],
$$

(2.73)

where

$$
t \equiv (p_z/m_e c)^2 \quad \text{and} \quad \text{sign}(p_z) \equiv p_z/|p_z|,
$$

(2.74)

For photons with energy larger than 5 MeV, for which Doppler broadening is negligible, we set $E' = E_C$ (which amounts to assuming that $p_z = 0$). In this case, the active

$^3$Notice that, due to the approximation introduced in eq. (2.36), a value $p_z < -m_e c$ would yield a negative energy for the scattered photon.
electron shell $i$ is sampled with relative probability $Z_i$ and binding effects are accounted for by simply rejecting $E'$-values such that $E - E' < U_i$.

The azimuthal scattering angle $\phi$ of the photon is sampled uniformly in the interval $(0, 2\pi)$. We assume that the Compton electron is emitted with energy $E_c = E - E' - U_i$ in the direction of the momentum transfer vector $\mathbf{q} = \hbar \mathbf{k} - \hbar \mathbf{k'}$, with polar angle $\theta_e$ and azimuthal angle $\phi_e = \phi + \pi$, relative to the direction of the incident photon. $\cos \theta_e$ is given by

$$\cos \theta_e = \frac{E - E' \cos \theta}{\sqrt{E^2 + E'^2 - 2EE' \cos \theta}}$$  \hspace{1cm} (2.75)

When $E' = E_C$, this expression simplifies to

$$\cos \theta_e = \frac{E + m_e c^2}{E} \left( \frac{E - E_C}{2m_e c^2 + E - E_C} \right)^{1/2}$$  \hspace{1cm} (2.76)

which coincides with the result (A.20). Since the active electron shell is known, characteristic x-rays and electrons emitted in the de-excitation of the ionized atom can also be followed. This is important, for instance, to account for escape peaks in scintillation or solid state detectors.

<table>
<thead>
<tr>
<th>$E$ (eV)</th>
<th>Al</th>
<th>Ag</th>
<th>Au</th>
</tr>
</thead>
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<tr>
<td>$10^3$</td>
<td>16.6</td>
<td>11.9</td>
<td>13.4</td>
</tr>
<tr>
<td>$10^4$</td>
<td>11.0</td>
<td>11.4</td>
<td>11.5</td>
</tr>
<tr>
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</tr>
<tr>
<td>$10^6$</td>
<td>8.2</td>
<td>8.2</td>
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</tr>
<tr>
<td>$10^7$</td>
<td>7.5</td>
<td>7.5</td>
<td>7.5</td>
</tr>
</tbody>
</table>

As a measure of the efficiency of the sampling algorithm, we may consider the average number $n_e$ of random numbers $\xi$ required to simulate an incoherent scattering event. $n_e$ is practically independent of the atomic number and decreases with photon energy (see table 2.1). The increase of $n_e$ at low energies stems from the loss of efficiency of the algorithm used to sample $\cos \theta$. Although the simulation of incoherent events becomes more laborious as the photon energy decreases, this has only a small influence on the speed of practical photon transport simulations since low-energy photons interact predominantly via photoelectric absorption (see fig. 2.10 below).

### 2.4 Electron-positron pair production

Electron-positron pairs can be created by absorption of a photon in the vicinity of a massive particle, a nucleus or an electron, which absorbs energy and momentum so that
Chapter 2. Photon interactions

these two quantities are conserved. The threshold energy for pair production in the field of a nucleus (assumed of infinite mass) is \(2m_e c^2\). When pair production occurs in the field of an electron, the target electron recoils after the event with appreciable kinetic energy; the process is known as “triplet production” because it causes three visible tracks when observed, e.g., in a cloud chamber. If the target electron is at rest, triplet production is only possible for photons with energy larger than \(4m_e c^2\).

For the simulation of pair production events in the field of an atom of atomic number \(Z\), we shall use the following semiempirical model (Baró et al., 1994a). Our starting point is the high-energy DCS for arbitrary screening, which was derived by Bethe and Heitler from the Born approximation (Motz et al., 1969; Tsai, 1974). The Bethe-Heitler DCS for a photon of energy \(E\) to create an electron-positron pair, in which the electron has a kinetic energy \(E_\text{e} = \epsilon E - m_e c^2\), can be expressed as (Tsai, 1974)

\[
\frac{d\sigma^{(BH)}}{d\epsilon} = r_0^2 \alpha Z[Z + \eta] \left\{ \left[\epsilon^2 + (1 - \epsilon)^2\right] \left(\Phi_1 - 4f_c\right) + \frac{2}{3}\epsilon(1 - \epsilon)(\Phi_2 - 4f_c) \right\}.
\] (2.77)

Notice that the “reduced energy” \(\epsilon = (E_\text{e} + m_e c^2)/E\) is the fraction of the photon energy that is taken away by the electron. The screening functions \(\Phi_1\) and \(\Phi_2\) are given by integrals that involve the atomic form factor and, therefore, must be computed numerically when a realistic form factor is adopted (e.g., the analytical one described in section 2.1). To obtain approximate analytical expressions for these functions, we shall assume that the Coulomb field of the nucleus is exponentially screened by the atomic electrons (Schiff, 1968; Tsai, 1974), i.e., the electrostatic potential of the atom is assumed to be (Wentzel model)

\[
\varphi_W(r) = \frac{Ze}{r} \exp(-r/R),
\] (2.78)

with the screening radius \(R\) considered as an adjustable parameter (see below). The corresponding atomic electron density is obtained from Poisson’s equation,

\[
\rho_W(r) = \frac{1}{4\pi e} \nabla^2 \varphi(r) = \frac{1}{4\pi e} \frac{1}{r} \frac{d^2}{dr^2} \left[ r \varphi(r) \right] = \frac{Z}{4\pi R^2} \exp(-r/R),
\] (2.79)

and the atomic form factor is

\[
F_W(q, Z) = 4\pi \int_0^\infty \rho_W(r) \frac{\sin(qr/\hbar)}{qr/\hbar} r^2 dr = \frac{Z}{1 + (Rq/\hbar)^2}.
\] (2.80)

The screening functions for this particular form factor take the following analytical expressions (Tsai, 1974)

\[
\begin{align*}
\Phi_1 &= 2 - 2\ln(1 + b^2) - 4b \arctan(b^{-1}) + 4\ln(Rm_e c/\hbar) \\
\Phi_2 &= \frac{4}{3} - 2\ln(1 + b^2) + 2b^2 \left[4 - 4\arctan(b^{-1}) - 3\ln(1 + b^{-2})\right] \\
&\quad + 4\ln(Rm_e c/\hbar),
\end{align*}
\] (2.81)

where

\[
b = \frac{Rm_e c}{\hbar} \frac{1}{2\kappa \epsilon(1 - \epsilon)}.
\] (2.82)
24. Electron-positron pair production

The quantity \( \eta \) in eq. (2.77) accounts for pair production in the field of the atomic electrons (triplet production), which is considered in detail by Hubbell et al. (1980) and Tsai (1974). In order to simplify the calculations, the dependence of the triplet cross section on the electron reduced energy, \( \epsilon \), is assumed to be the same as that of the pair cross section. The function \( f_c \) in (2.77) is the high-energy Coulomb correction of Davies, Bethe and Maximon (1954) given by

\[
f_c(Z) = a^2 \left[ (1 + a^2)^{-1} + 0.202059 - 0.03693a^2 + 0.00835a^4 
- 0.00201a^6 + 0.00049a^8 - 0.00012a^{10} + 0.00003a^{12} \right],
\]

(2.83)

with \( a = \alpha Z \). The total atomic cross section for pair (and triplet) production is obtained as

\[
\sigma_{\text{pp}}^{(\text{BH})} = \int_{\epsilon_{\text{min}}}^{\epsilon_{\text{max}}} \frac{d\sigma_{\text{pp}}^{(\text{BH})}}{d\epsilon} d\epsilon,
\]

(2.84)

where

\[
\epsilon_{\text{min}} = m_e e^2 / E = \kappa^{-1} \quad \text{and} \quad \epsilon_{\text{max}} = 1 - m_e e^2 / E = 1 - \kappa^{-1}.
\]

(2.85)

Extensive tables of pair production total cross sections, evaluated by combining different theoretical approximations, have been published by Hubbell et al. (1980). These tables give the separate contributions of pair production in the field of the nucleus and in that of the atomic electrons for \( Z = 1 \) to 100 and for photon energies from threshold up to \( 10^6 \) MeV. Following Salvat and Fernández-Varea (1992), the screening radius \( R \) has been determined by requiring that eq. (2.77) with \( \eta = 0 \) exactly reproduces the total cross sections given by Hubbell et al. (1980) for pair production in the nuclear field by \( 10^6 \) MeV photons (after exclusion of radiative corrections, which only amount to \( \sim 1\% \) of the total cross section). The screening radii for \( Z = 1-92 \) obtained in this way are given in table 2.2.

Actually, the triplet contribution, \( \eta \), varies with the photon energy. It increases monotonically from zero at \( E \approx 4m_e c^2 \) and reaches a saturation value, \( \eta_{\infty} \), at high energies. It can be obtained, for all elements and energies up to \( 10^6 \) MeV, as

\[
\eta(E) = Z\sigma_{\text{triplet}}^{\text{HGO}}(E) / \sigma_{\text{pair}}^{\text{HGO}}(E),
\]

(2.86)

where \( \sigma_{\text{pair}}^{\text{HGO}} \) and \( \sigma_{\text{triplet}}^{\text{HGO}} \) are the total pair and triplet production cross sections given by Hubbell et al. (1980). At \( 10^6 \) MeV, the high-energy limit is reached, i.e.

\[
\eta_{\infty} \simeq Z\sigma_{\text{triplet}}^{\text{HGO}}(10^6 \text{ MeV}) / \sigma_{\text{pair}}^{\text{HGO}}(10^6 \text{ MeV}).
\]

(2.87)

The values of \( \eta_{\infty} \) for the elements \( Z = 1-92 \) are given in table 2.2. The average dependence of \( \eta \) on the photon energy is approximated by the following empirical expression

\[
\eta = [1 - \exp(-v)]\eta_{\infty},
\]

(2.88)
Table 2.2: Reduced screening radius, $Rm_e c / \hbar$, and high-energy triplet contribution, $\eta_{\infty}$, for electron-positron pair production obtained from the tables of Hubbell et al. (1980) as described in the text. Notice that $\hbar / m_e c = 3.8616 \times 10^{-13}$ m is the Compton wavelength of the electron.

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<th>$Z$</th>
<th>$Rm_e c / \hbar$</th>
<th>$\eta_{\infty}$</th>
<th>$Z$</th>
<th>$Rm_e c / \hbar$</th>
<th>$\eta_{\infty}$</th>
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where

\[ v = (0.2840 - 0.1909a) \ln(4/\kappa) + (0.1095 + 0.2206a) \ln^2(4/\kappa) \\
+ (0.02888 - 0.04269a) \ln^3(4/\kappa) + (0.002527 + 0.002623a) \ln^4(4/\kappa). \]  

(2.89)

Then, the single quantity \( \eta_\infty \) characterizes the triplet production for each element.

The approximation given by eq. (2.77) with the fitted value of the screening radius, fails at low energies where it systematically underestimates the total cross section (it can even become negative). To compensate for this fact we introduce an empirical correcting term \( F_0(\kappa, Z) \), which acts in a way similar to the Coulomb correction. To facilitate the random sampling, the Bethe-Heitler DCS, eq. (2.77), including this low-energy correction and a high-energy radiative correction, is written in the form

\[
\frac{d\sigma_{pp}}{d\epsilon} = r_e^2 \alpha Z \left[ Z + \eta \right] C_\epsilon \frac{2}{3} \left[ 2 \left( \frac{1}{2 \epsilon} \right) \phi_1(\epsilon) + \phi_2(\epsilon) \right],
\]

(2.90)

where

\[
\phi_1(\epsilon) = g_1(\beta) + g_0(\kappa)
\]

\[
\phi_2(\epsilon) = g_2(\beta) + g_0(\kappa)
\]

(2.91)

with

\[
g_1(\beta) = \frac{1}{2} (3\Phi_1 - \Phi_2) - 4 \ln(Rm_e c/\hbar) = \frac{7}{3} - 2 \ln(1 + b^2) - 6b \arctan(b^{-1}) \\
- b^2 \left[ 4 - 4b \arctan(b^{-1}) - 3 \ln(1 + b^{-2}) \right],
\]

\[
g_2(\beta) = \frac{1}{4} (3\Phi_1 + \Phi_2) - 4 \ln(Rm_e c/\hbar) = \frac{11}{6} - 2 \ln(1 + b^2) - 3b \arctan(b^{-1}) \\
+ \frac{1}{2} b^2 \left[ 4 - 4b \arctan(b^{-1}) - 3 \ln(1 + b^{-2}) \right],
\]

\[
g_0(\kappa) = 4 \ln(Rm_e c/\hbar) - 4 f_\infty(Z) + F_0(\kappa, Z).
\]

(2.92)

\( C_\epsilon = 1.0093 \) is the high-energy limit of Mork and Olsen’s radiative correction (Hubbell et al., 1980).

The correcting factor \( F_0(\kappa, Z) \) has been determined by requiring that the total cross section for pair production obtained from the expression given in eq. (2.90) (with \( \eta = 0 \)) coincides with the total cross sections for pair production in the field of the nucleus tabulated by Hubbell et al. (1980). By inspection and numerical fitting, we have obtained the following analytical approximation

\[
F_0(\kappa, Z) = (-0.1774 - 12.10a + 11.18a^2)(2/\kappa)^{1/2} \\
+ (8.523 + 73.26a - 44.41a^2)(2/\kappa) \\
- (13.52 + 121.1a - 96.41a^2)(2/\kappa)^{3/2} \\
+ (8.946 + 62.05a - 63.41a^2)(2/\kappa)^2.
\]

(2.93)
The functions $\phi_1$ and $\phi_2$ are now positive except for $\epsilon$-values very near the endpoints of the allowed interval, given by eq. (2.85), for high atomic number elements. To avoid inconsistencies, these functions are set equal to zero when they take negative values.

The relative differences between the total atomic cross sections obtained from the DCS given by eq. (2.90) and the total cross sections tabulated by Hubbell et al. (1980) are appreciable near the threshold [actually, (2.90) shifts the threshold for pair production to values slightly larger than $2m_e c^2$], but decrease rapidly with increasing photon energy. At $E = 3 \text{ MeV}$, the differences reduce to 4\% and do not exceed 2\% for energies larger than 6 MeV, for almost all the elements. Although these differences are not important, they may be larger than the uncertainties in the cross sections given by Hubbell et al. (1980). To avoid systematic errors, the mean free paths for pair production used in PENLOPE are obtained by interpolation in a table generated with the XCOM program (Berger and Hubbell, 1987). The Bethe-Heitler DCS is only used to sample the kinetic energies of the produced pair.

It is also worth noting that the Bethe-Heitler theory predicts that the pair-production DCS, considered as a function of the electron reduced energy $\epsilon$, is symmetrical about $\epsilon = 1/2$ (see fig. 2.9). This dependence on $\epsilon$ is reasonably accurate only for photon energies larger than $\sim 5 \text{ MeV}$. For lower photon energies, the effect of the electrostatic field of the atom (which slows down the electron and accelerates the positron) becomes increasingly important, with the result that the actual DCS becomes asymmetrical and the mean value of $\epsilon$ becomes less than 1/2 (see e.g. Motz et al., 1969). At these relatively low energies, however, pair production is not dominant and, moreover, the produced particles have ranges that are much less than the mean free path of the absorbed photon. Therefore, no appreciable simulation errors are incurred by using the Bethe-Heitler DCS, eq. (2.90), for energies down to the threshold.

### 2.4.1 Simulation of pair production events

The Bethe-Heitler DCS, eq. (2.90), only depends on the kinetic energy $E_\gamma = \epsilon E - m_e c^2$ of the produced electron, so that $E_\gamma$ can be directly sampled from eq. (2.90): the kinetic energy of the positron is obtained as $E_+ = E - E_\gamma - 2m_e c^2$. Notice that, although the Bethe-Heitler total atomic cross section accounts for pair and triplet production, all the events are simulated as if they were pairs. This approximation is justified by the fact that, in triplet production, the recoiling electron has a range that is much smaller than the mean free path of the incident photon.

The electron reduced energy $\epsilon$ is distributed in the interval $(\kappa^{-1}, 1 - \kappa^{-1})$, see eq. (2.85), according to the PDF given by eq. (2.90) (normalization is again irrelevant)

$$p_{pp}(\epsilon) = 2 \left( \frac{1}{2} - \epsilon \right)^2 \phi_1(\epsilon) + \phi_2(\epsilon),$$

(2.94)

which is symmetrical about the point $\epsilon = 1/2$. Fig. 2.9 shows this PDF for lead and various photon energies. The following algorithm for sampling $\epsilon$ is based on the fact...
that the functions \( \phi_1(\epsilon) \) and \( \phi_2(\epsilon) \) are non-negative and attain their maximum values at \( \epsilon = 1/2 \).

**Figure 2.9:** Pair production DCS in lead as a function of the electron reduced energy, \( \epsilon = (E_\gamma + m_e c^2)/E \). (Adapted from Baró et al., 1994a.)

Except for a normalization constant, the PDF (2.94) can be written in the form

\[
p_{pp}(\epsilon) = u_1 U_1(\epsilon) \pi_1(\epsilon) + u_2 U_2(\epsilon) \pi_2(\epsilon)
\]  

(2.95)

with

\[
u_1 = \frac{2}{3} \left( \frac{1}{2} - \frac{1}{\kappa} \right)^2 \phi_1(1/2), \quad u_2 = \phi_2(1/2),
\]  

(2.96)

\[
\pi_1(\epsilon) = \frac{3}{2} \left( \frac{1}{2} - \frac{1}{\kappa} \right)^{-3} \left( \frac{1}{2} - \epsilon \right)^2, \quad \pi_2(\epsilon) = \frac{1}{2} \left( \frac{1}{2} - \frac{1}{\kappa} \right)^{-1}
\]  

(2.97)

and

\[
U_1(\epsilon) = \phi_1(\epsilon)/\phi_1(1/2), \quad U_2(\epsilon) = \phi_2(\epsilon)/\phi_2(1/2).
\]  

(2.98)

The functions \( \pi_i(\epsilon) \) are normalized PDFs in the interval \((\kappa^{-1}, 1 - \kappa^{-1})\), from which random values of \( \epsilon \) can be easily sampled by using the inverse transform method. In this interval, the functions \( U_i(\epsilon) \) are positive and less than unity, i.e. they are valid rejection functions. The generation of random values of \( \epsilon \) from the distribution (2.95) can now be performed by combining the composition and rejection methods (see section 1.2) according to the following algorithm:

(i) Sample a value of the integer \( i (=1, 2) \) according to the point probabilities

\[
p(1) = \frac{u_1}{u_1 + u_2} \quad \text{and} \quad p(2) = \frac{u_2}{u_1 + u_2}.
\]  

(2.99)
(ii) Sample \( \epsilon \) from \( \pi_i(\epsilon) \) using the sampling formulae (inverse transform method, see section 1.2.2)

\[
\epsilon = \begin{cases} 
\frac{1}{2} + \left( \frac{1}{2} - \frac{1}{\kappa} \right) (2\xi - 1)^{1/3} & \text{if } i = 1, \\
\frac{1}{\kappa} \left( \frac{1}{2} - \frac{1}{\kappa} \right) 2\xi & \text{if } i = 2.
\end{cases}
\] (2.100)

(iii) Generate a new random number \( \xi \).

(iv) If \( \xi > U_i(\epsilon) \), go to step (i).

(v) Deliver \( \epsilon \).

Notice that the quantity \( 2\xi - 1 \) may be negative and, therefore, taking its cube root will lead to a computer error; provision of this fact must be made when programming the algorithm. The efficiency of the algorithm is greater than 70\% for energies near the threshold, and increases with increasing photon energies. For \( E = 1 \) GeV it is of the order of 95\% for all the elements in the periodic table.

**Angular distribution of the produced particles**

Actually, the complete DCS for pair production is a function of the directions of the pair of particles. As the final state involves three bodies (the nucleus and the produced pair), the directions of the produced particles cannot be obtained from only their kinetic energies. The polar angles of the directions of movement of the electron and positron (\( \theta_- \) and \( \theta_+ \), fig. 2.1) relative to the direction of the incident photon are sampled from the leading term of the expression obtained from high-energy theory (Heitler, 1954; Motz et al., 1969)

\[
p(\cos \theta_\pm) = a (1 - \beta_\pm \cos \theta_\pm)^{-2},
\] (2.101)

where \( a \) is a normalization constant and

\[
\beta_\pm = \sqrt{E_\pm(E_\pm + 2m_e c^2)}
\] (2.102)

is the particle velocity in units of the speed of light. Random values of \( \cos \theta_\pm \) are obtained by using the inverse transform method (see section 1.2.2), which leads to the sampling formula

\[
\cos \theta_\pm = \frac{2\xi - 1 + \beta_\pm}{(2\xi - 1)\beta_\pm + 1}.
\] (2.103)

As the directions of the produced particles and the incident photon are not necessarily coplanar, the azimuthal angles \( \phi_- \) and \( \phi_+ \) of the electron and the positron are sampled independently and uniformly in the interval \((0, 2\pi)\).

It is worth stressing the fact that the produced charged particles have ranges that are much smaller than the mean free path of the photons. Moreover, the charged particles immediately enter a multiple elastic scattering process which randomizes their
2.5. Attenuation coefficients

directions of movement. As a consequence, there should be little difference between simulation results obtained with the present method and with exact random sampling from a more accurate DCS, differential in the energies and directions of the generated particles.

**Compound materials**

Let us consider a compound $X_x Y_y$, in which the molecules consist of $x$ atoms of the element $X$ and $y$ atoms of the element $Y$. The number of electrons per molecule is $Z_M = xZ(X) + yZ(Y)$ and the molecular weight is $A_M = xA_w(X) + yA_w(Y)$, where $Z(X)$ and $A_w(X)$ stand for the atomic number and atomic weight of element $X$.

In the simulation of pair-production events, we could use the molecular DCSs obtained from the additivity rule. The simulation of each event would then consist of 1) sampling the atom which participates in the interaction and 2) generating a random value of the electron reduced energy $\epsilon$ from the corresponding atomic DCS. To save computer time, PENELLOPE generates $\epsilon$ by considering an “equivalent” single element material of the same mass density $\rho$ as the actual medium, atomic number $Z_{eq}$ and atomic weight $A_{eq}$ given by

$$Z_{eq}A_M = Z_MA_{eq} = xZ(X)A_w(X) + yZ(Y)A_w(Y), \quad (2.104)$$

i.e. its atomic number (weight) is the mass-average ($Z$-average) of the atomic numbers (weights) of the constituent atoms. The reduced energy is sampled from the DCS of the element with the atomic number closest to $Z_{eq}$. Usually, this approximation does not alter the simulation results appreciably and permits a considerable simplification of the program and a reduction of the simulation time.

**2.5 Attenuation coefficients**

The photon inverse mean free path for a given mechanism is known as the partial attenuation coefficient of that mechanism. Thus, the partial attenuation coefficient for photoelectric absorption is

$$\mu_{ph} = \mathcal{N}\sigma_{ph}, \quad (2.105)$$

where $\mathcal{N} = N_A \rho / A_M$ is the number of atoms or molecules per unit volume and $\sigma_{ph}$ is the atomic or molecular photoelectric cross section. The photoelectric mass attenuation coefficient is defined as $\mu_{ph}/\rho$ and, therefore, is independent of the density of the material. Analogous definitions apply for the other interaction processes. The total mass attenuation coefficient is obtained as

$$\frac{\mu}{\rho} = \frac{N_A}{A_M} (\sigma_{pa} + \sigma_{co} + \sigma_{ph} + \sigma_{pp}). \quad (2.106)$$

As mentioned above, PENELLOPE uses tables of total cross sections for photoelectric absorption and pair production obtained from the database EPDL (Cullen et al., 1997)
and the program XCOM (Berger and Hubbell, 1987), respectively. Photoelectric cross sections for energies different from those in the tables are calculated by linear log-log interpolation. Total cross sections for pair production are evaluated by cubic spline log-log interpolation of the function \((1 - 2m_ec^2/E)^{-3}\sigma_{pp}\), which varies slowly with the photon energy.

\[E (\text{eV})\]
\[\mu/\rho (\text{cm}^2/\text{g})\]

**Figure 2.10**: Partial and total mass attenuation coefficients of water and lead as functions of the photon energy.

Mean free paths for coherent and incoherent scattering are computed from the DCSs described in sections 2.1 and 2.3. The resulting values are virtually identical to those given by the XCOM program for \(E\) greater than \(\sim 50\) keV. At lower energies, our mean free paths for Compton scattering deviate from those given by XCOM; these were calculated from a different theoretical model (Hubbell et al., 1975), which neglects Doppler broadening (see e.g. Brusa et al., 1996). The evaluation of the total atomic cross section for these processes [see eqs. (2.10) and (2.54)] involves a numerical quadrature, which is performed by using the function SUMGA (appendix B). Notice that for high-energy photons, the integrand in the coherent scattering cross section, eq. (2.10), is sharply peaked at \(\theta = 0\). In such a case, the numerical integration method is not effective. For energies larger than \(\sim Z/2\) MeV, we take advantage of the asymptotic behaviour shown by eq. (2.12) to avoid time-consuming integration. Partial and total attenuation coefficients for water and lead, as representatives of low- and high-Z materials, are displayed in fig. 2.10.
2.6 Atomic relaxation

Atoms are primarily ionized by photon interactions and by electron or positron impact. There is a fundamental difference between the ionizing effects of photons and of charged particles. A photon is only able to directly ionize a few atoms. In the case of photoabsorption, when the photon energy is larger than the K-shell binding energy, about 80% of photoabsorptions occur in the K-shell, i.e. the resulting ion with a vacancy in the K-shell is highly excited. Incoherent scattering is not as highly preferential, but still the probability that an inner shell is ionized is nearly proportional to the number of electrons in the shell. Conversely, fast electrons and positrons (and other charged particles) ionize many atoms along their paths; the ionizations occur preferentially in the less tightly bound atomic shells, or the conduction band in the case of metals (see section 3.2), so that most of the produced ions are only weakly excited.

![Diagram showing relative probabilities for radiative and non-radiative (Auger) transitions that fill a vacancy in the K-shell of atoms.]

**Figure 2.11:** Relative probabilities for radiative and non-radiative (Auger) transitions that fill a vacancy in the K-shell of atoms.

Excited ions with a vacancy in an inner shell relax to their ground state through a sequence of radiative and non-radiative transitions. In a radiative transition, the vacancy is filled by an electron from an outer shell and an x-ray with characteristic energy is emitted. In a non-radiative transition, the vacancy is filled by an outer electron and the excess energy is released through emission of an electron from a shell that is farther out (Auger effect). Each non-radiative transition generates an additional vacancy that, in turn, migrates “outwards”. The production of vacancies in inner shells and their subsequent relaxation must be simulated in detail, since the energetic x-rays and/or electrons emitted during the process may transport energy quite a distance from the excited ion.

PENELOPE simulates the emission of characteristic radiation and Auger electrons that
result from vacancies produced in K-shells and L-subshells by photoelectric absorption, Compton scattering and electron/positron impact (see chapter 3). The relaxation of these vacancies is followed until the K- and L-shells are filled up, i.e. until the vacancies have migrated to M and outer shells. Vacancies in these outer shells originate much less energetic secondary radiation, whose main effect is to spread out the excitation energy of the ion within the surrounding material. To get a reliable description of the dose distribution, and other macroscopic transport characteristics, we only have to follow secondary radiation that is able to propagate to distances of the order of, say, 1% of the penetration distance (or range) of the primary radiation. Radiation with lower energy does not need to be followed, since its only effect is to blur the “primary” dose distribution on a small length scale.

To simplify the description of the ionization processes of outer shells (i.e. photoelectric absorption, Compton scattering and electron/positron impact), we simply assume that, when ionization occurs in M or outer shells, a secondary (delta) electron is emitted from the parent ion with a kinetic energy $E_s$ equal to the energy deposited by the primary particle,

$$E_{\text{dep}} = \begin{cases} E - E' & \text{in Compton scattering,} \\ E & \text{in photoelectric absorption,} \\ W & \text{in electron/positron impact (see chapter 3).} \end{cases}$$

That is, the whole excitation energy of the ion is taken up by the ejected electron and no fluorescent radiation is simulated. In reality, the emitted electrons have energies less than the values (2.107) and can be followed by characteristic x-rays, which have mean free paths that are usually much larger than the Bethe range of photoelectrons. By giving an artificially increased initial energy to the electron we allow it to transport energy farther from the ion so as to partially compensate for the neglect of other radiation emitted during the de-excitation cascade.

In the case of ionization of an inner shell $i$, i.e. a K-shell or an L-shell, we consider that the electron is ejected with kinetic energy

$$E_s = E_{\text{dep}} - U_i,$$

where $U_i$ is the ionization energy of the active shell, and that the target atom is left with a vacancy in shell $i$. As mentioned above, we consider only characteristic x-rays and Auger electrons emitted in the first stages of the relaxation process. These secondary radiations are assumed to be emitted isotropically from the excited atom. We use the following notation to designate the possible transitions

- Radiative: S0-S1 (an electron from the S1 shell fills the vacancy in the S0 shell, leaving a hole in the S1 shell). The considered radiative transitions (for elements with $Z > 18$ with the M-shell filled) are shown in fig. 2.3.
- Non-radiative: S0-S1-S2 (an electron from the S1 shell fills the vacancy in the S0 shell, and the released energy is taken away by an electron in the S2 shell; this process leaves two vacancies, in the S1 and S2 shells).
2.6. Atomic relaxation

Non-radiative transitions of the type L-i-LJ-Xq, which involve an electron transition between two L-subshells and the ejection of an electron from an outer shell Xq are known as L-shell Coster-Kronig transitions.

The information furnished to PENELOE for each element consists of a table of possible transitions, transition probabilities and energies of the emitted x-rays or electrons for ionized atoms with a single vacancy in the K-shell or in an L-subshell. These data are entered through the material definition file. The transition probabilities are extracted from the LLNL Evaluated Atomic Data Library (Perkins et al., 1991). Fig. 2.11 displays transition probabilities for the transitions that fill a vacancy in the K shell as functions of the atomic number Z; the curve labelled “Auger” corresponds to the totality of non-radiative transitions. We see that for low-Z elements, the relaxation proceeds mostly through non-radiative transitions. It is worth noting that the ratio of probabilities of the radiative transitions K-S2 and K-S3 (where S stands for L, M or N) is approximately 1/2, as obtained from the dipole approximation (see e.g. Bransden and Joachain, 1983); radiative transitions K-S1 are strictly forbidden (to first order) within the dipole approximation.

The energy of the x-ray emitted in the radiative transition S0-S1 is assumed to be

$$E_x = U_{S0} - U_{S1}, \quad (2.109)$$

where $U_{Si}$ is the binding energy of an electron in the shell Si of the neutral atom, which is taken from the PENELOE database. Similarly, the energy of the electron emitted in the non-radiative transition S0-S1-S2 is set equal to

$$E_e = U_{S0} - U_{S1} - U_{S2}. \quad (2.110)$$

These emission energies correspond to assuming that the presence of the vacancy (or vacancies) does not alter the ionization energies of the active electron shells, which is an approximation. It should be noted that these formulae are also used to determine the energies of the emitted radiation at any stage of the de-excitation cascade, which means that we neglect the possible relaxation of the ion (see e.g. Sevier, 1972). Therefore, our approach will not produce L$_a$ and L$_{a'}$ x-ray satellite lines; these arise from the filling of a vacancy in a doubly-ionized L-shell (generated e.g. by a Coster-Kronig transition), which releases an energy that is slightly different from the energy liberated when the shell contains only a single vacancy. It is also worth recalling that the adopted transition probabilities are approximate. For K shells they are expected to be accurate to within one per cent or so, but for other shells they are subject to much larger uncertainties. Even the L-shell fluorescence yield (the sum of radiative transition probabilities for an L-shell vacancy) is uncertain by about 20% (see e.g. Hubbell, 1989; Perkins et al., 1991).

The simulation of the relaxation cascade is performed by subroutine RELAX. The transition that fills the initial vacancy is randomly selected according to the adopted transition probabilities, by using Walker’s aliasing method (section 1.2.3). This transition leaves the ion with one or two vacancies. If the energy of the emitted characteristic x-ray or Auger electron is larger than the corresponding absorption energy, the state
variables of the particle are stored in the secondary stack (which contains the initial states of all particles produced during the current shower that have not yet been simulated). The generation of the cascade continues by repeating the process for each remaining vacancy. It ends either when the K-shell and L-subshells have been filled up or when there is not enough energy to produce “active” radiation (with energy larger than the absorption energy). The excitation energy of the residual ion is assumed to be deposited locally.

It is important to bear in mind that we are disregarding the emission and transport of soft x-rays and slow electrons. This sets a lower limit to the photon energies for which PENEOPE is applicable. In principle, simulation results are expected to be reliable only for photons with energies larger than the ionization energy of the M1 subshell of the heaviest element present (125 eV for copper, 720 eV for silver, 3.4 keV for gold and 5.5 keV for uranium).
Chapter 3

Electron and positron interactions

In this chapter we consider the interactions of fast electrons and positrons of kinetic energy $E$ with matter. For the sake of simplicity, we start by assuming that the particles move in a single-element medium of atomic number $Z$ and density $\rho$, with $N$ atoms per unit volume. The extension to compounds, and mixtures, is normally done on the basis of the additivity approximation, i.e. the molecular DCS is approximated as the incoherent sum of the atomic DCSs of all the atoms in a molecule.

**Figure 3.1:** Basic interactions of electrons and positrons with matter.
The possible interactions of electrons and positrons with the medium are elastic scattering, inelastic collisions and bremsstrahlung emission; positrons can also undergo annihilation, either in flight or at rest. The atomic DCSs adopted in PENELLOPE are defined either as analytical functions or by means of numerical tables, or as a combination of both. These DCSs, which are sufficiently accurate for most practical simulation purposes, allow fast and accurate random sampling of the individual interactions. It is worth pointing out that multiple scattering distributions are quite insensitive to the fine details of the single scattering DCSs. If the adopted DCSs have a physically reasonable shape, only the values of a few integrals of the DCS have a direct influence on the simulation results (Liljequist, 1987; Fernández-Varea et al., 1993b). As a consequence, a general-purpose simulation procedure can be made fairly simple by using approximate DCSs with the proviso that they exactly reproduce the correct values of the relevant integrals. The DCSs described below represent a compromise between reliability and simplicity; they are simple enough to allow the use of fast sampling methods and, at the same time, they are flexible enough to account for the relevant features of the interactions.

Owing to the large number of interactions suffered by a fast electron or positron before coming to rest, detailed simulation is unfeasible at high energies. In PENELLOPE we overcome this practical difficulty by using a mixed simulation procedure (see chapter 4) instead of the habitual condensed simulation schemes adopted in other high-energy simulation codes —e.g. ETRAN (Berger and Seltzer, 1988), ITSS (Halbleib et al., 1992), EGS4 (Nelson et al., 1985), GEANT (Brun et al., 1986). The formulation of mixed simulation is complicated by the fact that the sampling of hard interactions is done from restricted DCSs, with cutoffs that vary with the particle energy during the evolution of a track. This limits the complexity of the DCSs that can be efficiently used in a simulation code.

3.1 Elastic collisions

In this section we consider the theoretical description of elastic collisions of electrons and positrons with isolated neutral atoms of atomic number Z at rest. By definition, elastic interactions are those in which the initial and final quantum states of the target atom are the same, normally the ground state. The angular deflections of electron trajectories in matter are mainly (but not completely) due to elastic scattering. Notice that there is a certain energy transfer from the projectile to the target, which causes the recoil of the latter (see section A.1.1). Because of the large mass of the target ($\sim 3500Zm_e$), the average energy lost by the projectile is a very small fraction of its initial energy (a few meV for scattering of 30 keV electron by Al atoms) and is usually neglected, which is equivalent to assuming that the target has an infinite mass and does not recoil.

For a wide energy range (say from a few hundred eV to $\sim 1$ GeV), elastic interactions can be described as scattering of the projectile by the electrostatic field of the target (Mott and Massey, 1965). The charge distribution of the target atom consists of the
nucleus and the electron cloud. The density of atomic electrons $\rho(r)$ can be calculated by using available Hartree-Fock codes (e.g. the one of Desclaux, 1975). For atoms with closed shell configurations, the electron distribution is spherically symmetrical; for atoms with open shells, we assume that an average over directions is performed to give a spherical density $\rho(r)$. To account for the effect of the finite size of the nucleus on the elastic DCS (which is appreciable only for projectiles with energy $E$ larger than a few MeV), we can represent the nucleus as a uniformly charged sphere of radius

$$R_{\text{nuc}} = 1.05 \times 10^{-15} A_w^{1/3} \text{ m},$$

(3.1)

where $A_w$ is the atomic mass (in g/mol). The electrostatic potential of the target atom is

$$\varphi(r) = \varphi_{\text{nuc}}(r) - e 4\pi \left[ \frac{1}{r} \int_0^r \rho(r') r'^2 \, dr' + \int_r^\infty \rho(r') r' \, dr' \right],$$

(3.2)

where

$$\varphi_{\text{nuc}}(r) = \begin{cases} 
\frac{1}{2} \frac{Ze}{R_{\text{nuc}}} \left[ 3 - \left( \frac{r}{R_{\text{nuc}}} \right)^2 \right] & \text{if } r \leq R_{\text{nuc}}, \\
\frac{Ze}{r} & \text{if } r > R_{\text{nuc}} 
\end{cases}$$

(3.3)

is the potential of the nucleus.

Within the static-field approximation (Mott and Massey, 1965; Walker, 1971), the DCS for elastic scattering of electrons or positrons is obtained by solving the partial-wave expanded Dirac equation for the motion of the projectile in the field of the target atom. The interaction energy is given by

$$V(r) = z_0 e \varphi(r) + V_{\text{ex}}(r),$$

(3.4)

where $z_0$ is the charge of the projectile in units of $e$ (−1 for electrons, +1 for positrons). The term $V_{\text{ex}}(r)$, which applies only for electrons, represents a local approximation to the exchange interaction between the projectile and the atomic electrons (see e.g. Salvat, 1998). We shall limit our considerations to the case of spin unpolarized projectiles, i.e. their spin is randomly oriented. Then, the effect of elastic interactions can be described as a deflection of the projectile trajectory, characterized by the polar and azimuthal scattering angles $\theta$ and $\phi$. For a central field, the angular distribution of singly scattered electrons is axially symmetric about the direction of incidence, i.e. independent of $\phi$. The DCS (per unit solid angle) for elastic scattering of a projectile with kinetic energy $E$ into the solid angle element $d\Omega$ about the direction $(\theta, \phi)$ is given by (Walker, 1971)

$$\frac{d\sigma_{\text{el}}}{d\Omega} = |f(\theta)|^2 + |g(\theta)|^2,$$

(3.5)

where

$$f(\theta) = \frac{1}{2ik} \sum_{\ell=0}^\infty \left\{ (\ell + 1) \left[ \exp(2i\delta_{\ell+}) - 1 \right] + \ell \left[ \exp(2i\delta_{\ell-}) - 1 \right] \right\} P_\ell(\cos \theta),$$

$$g(\theta) = \frac{1}{2ik} \sum_{\ell=0}^\infty \left\{ \exp(2i\delta_{\ell-}) - \exp(2i\delta_{\ell+}) \right\} P_\ell^1(\cos \theta)$$

(3.6)
are the direct and spin-flip scattering amplitudes, respectively.

\[ k \equiv \frac{p}{\hbar} = \frac{1}{\hbar c} \left[ E(E + 2m_ec^2) \right]^{1/2} \]  

(3.7)

is the wave number of the projectile, \( P_l(\cos \theta) \) are Legendre polynomials, \( P_l^m(\cos \theta) \) are associated Legendre functions and \( \delta_{\ell \pm} \) are the phase shifts. These are determined from the asymptotic behaviour of the Dirac radial functions for large \( r \) (Walker, 1971). Thus, to determine each phase shift we must solve the radial Dirac equations for the potential \( V(r) \). The convergence of the partial-wave series (3.6) slows down when the energy of the projectile increases. This makes the calculation difficult for energies larger than a few MeV (in the case of scattering by gold atoms, about 10,000 phase shifts are required at \( E = 10 \) MeV). The partial-wave DCS, eq. (3.5), rigourously accounts for spin and other relativistic effects, as well as finite nuclear size effects. The elastic scattering database used in PENELLOPE has been calculated essentially by this method, using a computer code written by Salvat (2000).

Single elastic collisions are determined by the values of the polar and azimuthal scattering angles, \( \theta \) and \( \phi \), respectively. Owing to the assumed spherical symmetry of the scattering centres, single and multiple scattering angular distributions are axially symmetrical about the direction of incidence, i.e. they are independent of the azimuthal scattering angle \( \phi \). For simulation purposes, it is convenient to measure polar angular deflections produced by single scattering events in terms of the variable [see eq. (1.61)]

\[ \mu = (1 - \cos \theta)/2 \]  

(3.8)

instead of the scattering angle \( \theta \). Notice that \( \mu \) varies from 0 (forward scattering) to 1 (backward scattering). The DCS per unit angular deflection is

\[ \frac{d\sigma_{el}}{d\mu} = 4\pi \frac{d\sigma_{el}}{d\Omega}. \]  

(3.9)

Fig. 3.2 displays DCSs for elastic scattering of electrons and positrons of various energies by aluminium and gold atoms. These numerical results illustrate the variation of the DCS with the atomic number \( Z \), the charge of the projectile and the energy \( E \). Since the interaction \( V(r) \) is attractive for electrons and repulsive for positrons, the scattering is more intense for electrons (which can fall deeply into the potential well of the atom) than for positrons (which are repelled from the nucleus and cannot “see” the inner part of the atom). The DCS for low-energy electrons exhibits a diffraction-like structure, while the DCS for positrons decreases monotonously with the deflection \( \mu \). The Born approximation (see e.g. Mott and Massey, 1965) predicts a structureless DCS that decreases with \( \mu \) and is proportional to the squared charge of the projectile (i.e. the same for electrons and positrons). This approximation considers the scattering field as a perturbation (to first order) and, hence, it is valid only for weak fields (low-\( Z \) elements). The difference between the (partial wave) DCSs for electrons and positrons gives a clear indication of the applicability of the Born approximation.
$$P = \text{HOHFWURQV SRVLWURQV}$$

$$G V H P F$$

Figure 3.2: DCS for elastic scattering of electrons and positrons by aluminium and gold atoms as a function of the deflection \( \mu = (1 - \cos \theta)/2 \). Notice the change from logarithmic to linear scale at \( \mu = 0.05 \).
The total elastic cross section is given by
\[ \sigma_{el} = \int \frac{d\sigma_{el}}{d\Omega} d\Omega = \int \frac{d\sigma_{el}}{d\mu} d\mu. \]  
(3.10)

Notice that we can write
\[ \frac{d\sigma_{el}}{d\mu} = \sigma_{el} p_{el}(\mu), \]
(3.11)
where \( p_{el}(\mu) \) is the normalized PDF of \( \mu \) in a single collision. The mean free path between consecutive elastic events in a homogeneous single-element medium is
\[ \lambda_{el} = 1/(N\sigma_{el}), \]
(3.12)
where \( N \) is the number of atoms per unit volume.

Other important quantities (see section 4.1) are the transport cross sections\(^1\)
\[ \sigma_{el,\ell} \equiv \int [1 - P_{\ell}(\cos \theta)] \frac{d\sigma_{el}}{d\Omega} d\Omega. \]  
(3.13)
The \( \ell \)-th transport mean free path is defined by
\[ \lambda_{el,\ell} \equiv 1/(N\sigma_{el,\ell}). \]  
(3.14)
The first and second transport cross sections, \( \sigma_{el,1} \) and \( \sigma_{el,2} \), are given by
\[ \sigma_{el,1} = \int (1 - \cos \theta) \frac{d\sigma_{el}}{d\Omega} d\Omega = 2\sigma_{el} \int_0^1 \mu p_{el}(\mu) d\mu = 2\sigma_{el} \langle \mu \rangle \]  
(3.15)
and
\[ \sigma_{el,2} = \int \frac{3}{2} (1 - \cos^2 \theta) \frac{d\sigma_{el}}{d\Omega} d\Omega = 6\sigma_{el} \int_0^1 (\mu - \mu^2) p_{el}(\mu) d\mu = 6\sigma_{el} \left( \langle \mu \rangle - \langle \mu^2 \rangle \right), \]  
(3.16)
where \( \langle \cdots \rangle \) indicates the average value in a single collision. The quantities \( \lambda_{el,1} \) and \( \lambda_{el,2} \), eq. (3.14), determine the first and second moments of the multiple scattering distributions (see section 4.1). The inverse of the first transport mean free path,
\[ \lambda_{el,1}^{-1} = N\sigma_{el,1} = \frac{2}{\lambda_{el,1}} \langle \mu \rangle, \]  
(3.17)
gives a measure of the average angular deflection per unit path length. By analogy with the “stopping power”, which is defined as the mean energy loss per unit path length (see section 3.2.3), the quantity \( 2\lambda_{el,1}^{-1} \) is sometimes called the “scattering power”\(^2\).

\(^1\)The Legendre polynomials of lowest orders are
\[ P_0(x) = 1, \quad P_1(x) = x, \quad P_2(x) = \frac{1}{2}(3x^2 - 1). \]

\(^2\)At high energies, where the scattering is concentrated at very small angles, \( \langle \mu \rangle \simeq \langle \theta^2 \rangle/4 \) and
\[ \lambda_{el,1}^{-1} \simeq (\theta^2)/(2\lambda_{el}). \]
3.1. Elastic collisions

Fig. 3.3 shows elastic mean free paths and transport mean free paths for electrons in aluminium and gold. At low energies, the differences between the DCS of the two elements (see fig. 3.2) produce very visible differences between the transport mean free paths. When $E$ increases, the DCS becomes strongly peaked in the forward direction and $\langle \mu^2 \rangle$ becomes much smaller than $\langle \mu \rangle$. In the high-energy limit, $\sigma_{el,2} \simeq 3\sigma_{el,1}$ ($\lambda_{el,2} \simeq \lambda_{el,1}/3$). The total cross section, $\propto 1/(\rho \lambda_{el})$, decreases monotonously with $E$ to reach a constant value at high energies. This saturation is a relativistic effect: the total cross section measures the interaction probability, which is proportional to the time spent by the projectile within the region where the scattering field is appreciable. This time is determined by the speed of the projectile, which approaches $c$ from below when the projectile energy increases. In the non-relativistic theory, the speed $v_{n.r.} = (2E/m_e)^{1/2}$ increases without limit with $E$ and the calculated non-relativistic total cross section tends to zero at high energies.

Figure 3.3: Elastic mean free path, $\lambda_{el}$, and first and second transport mean free paths, $\lambda_{el,1}$ and $\lambda_{el,2}$, for electrons scattered in aluminium and gold as functions of the kinetic energy of the projectile.

3.1.1 The modified Wentzel (MW) model

Although it is possible to do Monte Carlo simulation of electron and positron transport using numerical partial-wave DCSs (Benedito et al., 2001), this procedure is too
Chapter 3. Electron and positron interactions

laborious to be adopted as the basis of a simulation code for general purposes (mostly because of the large volume of required numerical information). It is more convenient to use suitable analytical approximate DCSs that may differ in detail from the partial wave DCSs but lead to nearly the same multiple scattering distributions. In PENELOPE we use a model in which the DCS is expressed as

$$\frac{d\sigma_{el}^{(MW)}}{d\mu} = \sigma_{el} p_{MW}(\mu). \quad (3.18)$$

The single scattering distribution $p_{MW}(\mu)$ is defined by a simple analytical expression, with a physically plausible form, depending on two adjustable parameters. These parameters are determined in such a way that the values of $\langle \mu \rangle$ and $\langle \mu^2 \rangle$ obtained from $p_{MW}(\mu)$ are equal to those of the actual (partial wave) DCS:

$$\langle \mu \rangle_{MW} \equiv \int_0^1 \mu p_{MW}(\mu) \, d\mu = \langle \mu \rangle = \frac{1}{2} \frac{\sigma_{el1}}{\sigma_{el}}, \quad (3.19)$$

and

$$\langle \mu^2 \rangle_{MW} \equiv \int_0^1 \mu^2 p_{MW}(\mu) \, d\mu = \langle \mu^2 \rangle = \frac{1}{6} \frac{\sigma_{el1}}{\sigma_{el}} - \frac{1}{6} \frac{\sigma_{el2}}{\sigma_{el}}. \quad (3.20)$$

Thus, the MW model will give the same mean free path and the same first and second transport mean free paths as the partial wave DCS. As a consequence (see chapter 4), detailed simulations using this model will yield multiple scattering distributions that do not differ significantly from those obtained from the partial wave DCS, quite irrespectively of other details of the “artificial” distribution $p_{MW}(\mu)$.

To set the distribution $p_{MW}(\mu)$, we start from the Wentzel (1927) angular distribution,

$$p_{W,A_0}(\mu) \equiv \frac{A_0(1 + A_0)}{(\mu + A_0)^2}, \quad (A_0 > 0) \quad (3.21)$$

which describes the scattering by an exponentially screened Coulomb field within the Born approximation (see e.g. Mott and Massey, 1965), that is, it provides a physically plausible angular distribution, at least for light elements or high-energy projectiles. It is also worth mentioning that the multiple scattering theory of Molière (1947, 1948) can be derived by assuming that electrons scatter according to the Wentzel distribution (see Fernández-Varea et al., 1993b). The first moments of the Wentzel distribution are

$$\langle \mu \rangle_{W,A_0} = \int_0^1 \mu \frac{A_0(1 + A_0)}{(\mu + A_0)^2} \, d\mu = A_0 \left[ (1 + A_0) \ln \left( \frac{1 + A_0}{A_0} \right) - 1 \right] \quad (3.22)$$

and

$$\langle \mu^2 \rangle_{W,A_0} = \int_0^1 \mu^2 \frac{A_0(1 + A_0)}{(\mu + A_0)^2} \, d\mu = A_0 \left[ 1 - 2 \langle \mu \rangle_{W,A_0} \right]. \quad (3.23)$$

Let us define the value of the screening constant $A_0$ so that $\langle \mu \rangle_{W,A_0} = \langle \mu \rangle$. The value of $A_0$ can be easily calculated by solving eq. (3.22) numerically, e.g. by the Newton-Raphson method. Usually, we shall have $\langle \mu^2 \rangle_{W,A_0} \neq \langle \mu^2 \rangle$. At low energies, the Wentzel
distribution that gives the correct average deflection is too “narrow” \([\langle \mu^2 \rangle_{W,A_0} < \langle \mu^2 \rangle\) for both electrons and positrons and for all the elements\]. At high energies, the angular distribution is strongly peaked in the forward direction and the Wentzel distribution becomes too “wide”. This suggests using a modified Wentzel (MW) model obtained by combining a Wentzel distribution with a simple distribution, which takes different forms in these two cases,

- **Case I.** If \(\langle \mu^2 \rangle_{W,A_0} > \langle \mu^2 \rangle\) (the Wentzel distribution is too wide), we take \(p_{MW}(\mu)\) as a statistical admixture of the Wentzel distribution and a delta distribution (a zero-width, fixed scattering angle process)

\[
p_{MW,1}(\mu) = (1 - B) p_{W,A}(\mu) + B \delta(\mu - \langle \mu \rangle)
\]

with

\[
A = A_0 \quad \text{and} \quad B = \frac{\langle \mu^2 \rangle_{W,A} - \langle \mu^2 \rangle}{\langle \mu^2 \rangle_{W,A} - \langle \mu \rangle^2}.
\]

Notice that in this case we usually have \(\langle \mu \rangle \ll 1\), so that the delta distribution is at very small angles. Although we have introduced a discrete peak in the DCS, its effect is smeared out by the successive collisions and not visible in the multiple scattering angular distributions.

- **Case II.** If \(\langle \mu^2 \rangle_{W,A_0} < \langle \mu^2 \rangle\) (the Wentzel distribution is too narrow), we express \(p_{MW}(\mu)\) as a statistical admixture of a Wentzel distribution (with \(A\) not necessarily equal to \(A_0\)) and a triangle distribution in the interval \((1/2,1)\),

\[
p_{MW,II}(\mu) = (1 - B) p_{W,A}(\mu) + B \left(\mu - \langle \mu \rangle\right) \Theta\left(\mu - \langle \mu \rangle\right).
\]

The parameters \(A\) and \(B\) are obtained from the conditions (3.19) and (3.20), which give

\[
(1 - B) \langle \mu \rangle_{W,A} + B \left[\frac{5}{6} \langle \mu \rangle\right] = \langle \mu \rangle
\]

\[
(1 - B) \langle \mu^2 \rangle_{W,A} + B \left[\frac{17}{24} \langle \mu^2 \rangle\right] = \langle \mu^2 \rangle.
\]

From the first of these equations,

\[
B = \frac{\langle \mu \rangle - \langle \mu \rangle_{W,A}}{\langle 5/6 - \langle \mu \rangle_{W,A} \rangle}.
\]

Inserting this value in the second of eqs. (3.27), we obtain

\[
\left(\frac{17}{24} - \langle \mu^2 \rangle\right) \langle \mu \rangle_{W,A} - \left(\frac{5}{6} - \langle \mu \rangle\right) \langle \mu^2 \rangle_{W,A} = \frac{17}{24} \langle \mu \rangle - \frac{5}{6} \langle \mu^2 \rangle.
\]

For all situations of interest, this equation has a single root \(A\) in the interval \((0, A_0)\) and can be easily solved by means of a bipartition procedure. The value of \(B\) given by eq. (3.28) is then positive and less than unity, as required.

In fig. 3.4 we compare partial wave DCSs and MW model DCSs for elastic scattering of electrons of various energies by gold atoms. The considered energies correspond to
the case-II MW model [so that the distribution $p_{\text{MW}}(\mu)$ is continuous]. We see that
the MW model does imitate the partial wave DCSs, but the differences are significant.
Nevertheless, the important fact here is that both DCSs give exactly the same values of
$\sigma_{\text{el}}, \langle \mu \rangle$ and $\langle \mu^2 \rangle$.

The information needed to determine the parameters of the MW model reduces to the
characteristic functions $\sigma_{\text{el}}(E)$, $\sigma_{\text{el,1}}(E)$ and $\sigma_{\text{el,2}}(E)$. PENelope reads these functions
from a precalculated database for electrons and positrons, for the elements $Z = 1$–92
and for a grid of energies that is dense enough to permit accurate cubic spline log-
log interpolation. This elastic scattering database was generated by using the partial-wave code of Salvat (2000); the atomic electron densities were obtained from the Dirac-
Hartree-Fock code of Desclaux (1975) — which correspond to free atoms. Before starting
the simulation, PENelope evaluates a table of the parameters $A$ and $B$, and stores it in
the computer memory. Instead of $B$, PENelope tabulates the quantity $B' = +B$ (case
I) and $B' = -B$ (case II); this avoids the need to specify the case, which can be inferred
from the sign of $B'$. It is worth noting that $A$ and $B'$ are continuous functions of energy
and, therefore, can be rapidly evaluated, for any energy, by interpolation in the stored
table. In case I, $\langle \mu \rangle$ coincides with $\langle \mu \rangle_{\text{w,A}}$, which is determined by $A$, eq. (3.22). Fig.
3.5 displays the MW model parameters for aluminium and gold, as representative of
low- and high-$Z$ elements. Notice that at high energies, where the case I model applies,
the strength of the delta contribution increases rapidly with energy, indicating that the

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{Au_Z_79_e_au.png}
\caption{Partial wave and MW model DCSs for elastic scattering of electrons by gold atoms.}
\end{figure}
partial-wave DCS is much narrower than the Wentzel distribution.

The MW model is directly applicable to compounds (and mixtures) by using the appropriate values of the total cross section and the first and second transport cross sections. In PENEL%OPE, these are calculated from atomic total and transport cross sections by means of the additivity approximation (incoherent sum of scattered intensities). This amounts to neglecting chemical binding effects. A more accurate approach, which yields a good estimate of these effects, is provided by the following independent atom approximation (Walker, 1968; Yates, 1968). Assume that the interaction of the projectile with each atom is still given by the free-atom static potential (3.4). The molecular DCS may then be evaluated by adding the waves (not the currents) scattered from the various atoms in the molecule and averaging over molecular orientations. The resulting DCS is given by

$$\frac{d\sigma_{el}}{d\Omega} = \sum_{i,j} \frac{\sin(qa_{ij}/\hbar)}{qa_{ij}/\hbar} \left[ f_i(\theta)f_j^*(\theta) + g_i(\theta)g_j^*(\theta) \right],$$  

(3.30)

where $q = 2\hbar k \sin(\theta/2)$ is the momentum transfer, $a_{ij}$ is the distance between the atoms $i$ and $j$ and $f_i, g_i$ are the scattering amplitudes, eq. (3.6), for the atom $i$. It has been claimed that DCSs obtained from this formulation agree with experiments to within $\sim 2\%$ (Walker, 1968; Yates, 1968). DCSs for scattering of 100 eV and 2.5 keV electrons in water vapour, obtained from the simple additivity rule and computed from

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{fig3_5.png}
\caption{Parameters of the MW model for scattering of electrons and positrons by aluminium and gold atoms. The scale of the energy axes is logarithmic.}
\end{figure}
eq. (3.30), are compared in fig. 3.6. It is seen that, for energies above a few keV, chemical binding causes a slight distortion of the DCS at small angles, and a slight rippling for intermediate angles. Therefore, the use of the additivity approximation (i.e. neglecting chemical binding effects) in Monte Carlo simulation at these energies is justified.

![Figure 3.6: Differential cross sections for elastic scattering of electrons by water molecules, calculated as the coherent sum of scattered waves, eq. (3.30), and from the additivity approximation (incoherent sum).](image)

### 3.1.2 Simulation of single elastic events with the MW model

As mentioned above, the angular distribution in single elastic events is axially symmetrical about the direction of incidence. Hence, the azimuthal scattering angle $\phi$ is sampled uniformly in the interval $(0, 2\pi)$ using the sampling formula $\phi = 2\pi \xi$. In detailed simulations, $\mu$ is sampled in the whole interval $(0,1)$. However, we shall also make use of the MW model for mixed simulation (see chapter 4), in which only hard events, with deflection $\mu$ larger than a given cutoff value $\mu_c$, are sampled individually. In this section we describe analytical (i.e. exact) methods for random sampling of $\mu$ in the restricted interval $(\mu_c, 1)$.

- Case I. The cumulative distribution function of $p_{MW,1}(\mu)$ is

  $$
P_{MW,1}(\mu) \equiv \int_0^\mu p_{MW,1}(\mu') \, d\mu' = \begin{cases} 
  (1 - B)(1 + A)\frac{\mu}{A + \mu} & \text{if } 0 \leq \mu < \langle \mu \rangle, \\
  B + (1 - B)(1 + A)\frac{\mu}{A + \mu} & \text{if } \langle \mu \rangle \leq \mu \leq 1.
  \end{cases}
  \tag{3.31}
$$

Owing to the analytical simplicity of this function, the random sampling of $\mu$ can be performed by using the inverse transform method (section 1.2.2). The sampling equation
for \( \mu \) in \((0,1)\) reads

\[
\mu = \mathcal{P}_{\mu,W}^{-1}(\xi),
\]

(3.32)

where \( \mathcal{P}_{\mu,W}^{-1}(\xi) \) is the inverse of the cumulative distribution function, which is given by

\[
\mathcal{P}_{\mu,W}^{-1}(\xi) = \begin{cases} 
\frac{\xi A}{(1 - B)(1 + A) - \xi} & \text{if } 0 \leq \xi < \xi_0, \\
\langle \mu \rangle & \text{if } \xi_0 \leq \xi < \xi_0 + B, \\
\frac{(\xi - B) A}{(1 - B)(1 + A) - (\xi - B)} & \text{if } \xi_0 + B \leq \xi \leq 1,
\end{cases}
\]

(3.33)

with

\[
\xi_0 = (1 - B) \frac{(1 + A) \langle \mu \rangle}{A + \langle \mu \rangle}.
\]

(3.34)

To sample \( \mu \) in the restricted interval \((\mu_c,1)\), we can still use the inverse transform method, cf. (3.32), but with the random number \( \xi \) sampled uniformly in the interval \((\xi_c,1)\) with

\[
\xi_c = \mathcal{P}_{\mu,W}(\mu_c).
\]

(3.35)

- Case II. The cumulative distribution function is

\[
\mathcal{P}_{\mu,W}(\mu) \equiv \int_0^\mu \mathcal{P}_{\mu,W}(\mu') d\mu'
\]

\[
= \begin{cases} 
(1 - B) \frac{(1 + A) \mu}{A + \mu} & \text{if } 0 \leq \mu < \frac{1}{2}, \\
(1 - B) \frac{(1 + A) \mu}{A + \mu} + B 4 \left[ \mu^2 - \mu + \frac{1}{4} \right] & \text{if } \frac{1}{2} \leq \mu \leq 1.
\end{cases}
\]

(3.36)

In principle, to sample \( \mu \) in \((0,1)\), we can adopt the inverse transform method. The sampling equation

\[
\xi = \mathcal{P}_{\mu,W}(\mu)
\]

(3.37)

can be cast in the form of a cubic equation. This equation can be solved either by using the analytical solution formulas for the cubic equation, which are somewhat complicated, or numerically, e.g. by the Newton-Raphson method. We employ this last procedure to determine the cutoff deflection (see section 4.1) for mixed simulation. To sample \( \mu \) in the restricted interval \((\mu_c,1)\) we use the composition method, which is easier than solving eq. (3.37). Notice that the sampling from the (restricted) Wentzel and from the triangle distributions can be performed analytically by the inverse transform method.

### 3.2 Inelastic collisions

The dominant energy loss mechanisms for electrons and positrons with intermediate and low energies are inelastic collisions, i.e. interactions that produce electronic excitations
and ionizations in the medium. The quantum theory of inelastic collisions of charged particles with individual atoms and molecules was first formulated by Bethe (1930, 1932) on the basis of the first-order (plane-wave) Born approximation. The extension of the theory to inelastic collisions in condensed materials has been discussed by Fano (1963). The formal aspects of the quantum theory for condensed matter are quite complicated. Fortunately, the results are essentially equivalent to those from classical dielectric theory.

The effect of individual inelastic collisions on the projectile is completely specified by giving the energy loss $W$ and the polar and azimuthal scattering angles $\theta$ and $\phi$, respectively. For amorphous media with randomly oriented atoms (or molecules), the DCS for inelastic collisions is independent of the azimuthal scattering angle $\phi$. Instead of the polar scattering angle $\theta$, it is convenient to use the recoil energy $Q$ [see eqs. (A.29) and (A.30)], defined by

$$Q(Q + 2m_\text{e}c^2) = (cq)^2.$$  \hspace{1cm} (3.38)

The quantity $q$ is the magnitude of the momentum transfer $q \equiv p - p'$, where $p$ and $p'$ are the linear momenta of the projectile before and after the collision. Notice that $Q$ is the kinetic energy of an electron that moves with a linear momentum equal to $q$.

Let us first consider the inelastic interactions of electrons or positrons ($\epsilon_\text{e}^2 = 1$) with an isolated atom (or molecule) containing $Z$ electrons in its ground state. The DCS for collisions with energy loss $W$ and recoil energy $Q$, obtained from the first Born approximation, can be written in the form (Fano, 1963)

$$\frac{d^2\sigma_{\text{in}}}{dW\,dQ} = \frac{2\pi \epsilon_\text{e}^2 e^4}{m_\text{e}v^2} \left( \frac{2m_\text{e}c^2}{WQ(Q + 2m_\text{e}c^2)} + \frac{\beta^2 \sin^2 \theta_r W^2 m_\text{e}c^2}{2m_\text{e}c^2 - W^2} \right) \frac{dJ(Q, W)}{dW},$$ \hspace{1cm} (3.39)

where $v = \beta c$ is the velocity of the projectile. $\theta_r$ is the angle between the initial momentum of the projectile and the momentum transfer, which is given by eq. (A.42),

$$\cos^2 \theta_r = \frac{W^2/\beta^2}{Q(Q + 2m_\text{e}c^2)} \left( 1 + \frac{Q(Q + 2m_\text{e}c^2) - W^2}{2W(E + m_\text{e}c^2)} \right)^2.$$ \hspace{1cm} (3.40)

The result (3.39) is obtained in the Coulomb gauge (Fano, 1963); the two terms on the right-hand side are the contributions from interactions through the instantaneous (longitudinal) Coulomb field and through the exchange of virtual photons (transverse field), respectively. The factor $dJ(Q, W)/dW$ is the atomic generalized oscillator strength (GOS), which completely determines the effect of inelastic interactions on the projectile, within the Born approximation. Notice, however, that knowledge of the GOS does not suffice to describe the energy spectrum and angular distribution of secondary knock-on electrons (delta rays).

The GOS can be represented as a surface over the $(Q, W)$ plane, which is called the Bethe surface (see Inokuti, 1971; Inokuti et al., 1978). Unfortunately, the GOS is known in analytical form only for two simple systems, namely, the (non-relativistic) hydrogenic ions (see fig. 3.7) and the free-electron gas. Even in these cases, the analytical expressions of the GOSs are too complicated for simulation purposes. For ionization of inner shells, the GOS can be computed numerically from first principles (see e.g. Manson, 1972), but
using GOSs defined through extensive numerical tables is impractical for Monte Carlo simulation. Fortunately, the physics of inelastic collisions is largely determined by a few global features of the Bethe surface. Relatively simple GOS models can be devised that are consistent with these features and, therefore, lead to a fairly realistic description of inelastic interactions (see e.g. Salvat and Fernández-Varea, 1992).

![Figure 3.7](image)

**Figure 3.7:** The GOS for ionization of the hydrogen atom \((Z = 1)\) in the ground state. All energies are in units of the ionization energy \(U_i = 13.6\) eV. The GOS for ionization of (non-relativistic) hydrogenic ions is independent of \(Z\) if energies are expressed in units of the ionization energy.

As mentioned above, the “atomic” DCS for inelastic interactions in dense media can be obtained from a semiclassical treatment in which the medium is considered as a dielectric, characterized by a complex dielectric function \(\epsilon(k, \omega)\), which depends on the wave number \(k\) and the frequency \(\omega\). In the classical picture, the (external) electric field of the projectile polarizes the medium producing an induced electric field that causes the slowing down of the projectile. The dielectric function relates the Fourier components of the total (external+induced) and the external electric potentials. It is convenient to interpret the quantities \(q = \hbar k\) and \(W = \hbar \omega\) as the momentum and energy transfers and consider that the dielectric function depends on the variables \(Q\) [defined by eq. (3.38)] and \(W\). The DCSs obtained from the dielectric and quantum treatments are consistent (i.e. the former reduces to the latter for a low-density medium) if one assumes the identity

\[
\frac{df(Q,W)}{dW} = W \frac{Q + m_e c^2}{m_e c^2} \frac{2Z}{\pi \Omega_p^2} \text{Im} \left( \frac{-1}{\epsilon(Q,W)} \right), \tag{3.41}
\]
where $\Omega_p$ is the plasma energy of a free-electron gas with the electron density of the medium, given by

$$\Omega_p^2 = 4\pi N Z e^2/m_e.$$  \hfill (3.42)

Eq. (3.41) establishes the connection between the atomic GOS (a property of individual atoms) and the dielectric function (a macroscopic concept). The DCS for the condensed medium can be expressed in the form [cf. eq. (3.39)],

$$\frac{d^2\sigma_{\text{in}}}{dW \, dQ} = \frac{2\pi s^2 e^4}{m_e e^2} \frac{d\bar{f}(Q, W)}{dW} \left( \frac{2m_e e^2}{W Q (Q + 2m_e e^2)} \right)$$

$$+ \left\{ \frac{\theta^2 \sin^2 \theta_e W^2 m_e e^2}{Q (Q + 2m_e e^2) - W^2|W^2|} - D(Q, W) \right\},$$  \hfill (3.43)

where the term $D(Q, W)$, which is appreciable only for small $Q$, accounts for the so-called density-effect correction (Sternheimer, 1952). The origin of this term is the polarizability of the medium, which “screens” the distant transverse interaction causing a net reduction of its contribution to the stopping power. The density-effect correction $D(Q, W)$ is determined by the dielectric function that, in turn, is related to the GOS. Thus, the GOS contains all the information needed to compute the DCS for electron/positron inelastic interactions in condensed media.

In the limit of very large recoil energies, the binding and momentum distribution of the target electrons have a small effect on the interaction. Therefore, in the large-$Q$ region, the target electrons behave as if they were essentially free and at rest and, consequently, the GOS reduces to a ridge along the line $W = Q$, which was named the Bethe ridge by Inokuti (1971). In the case of hydrogenic ions in the ground state, fig. 3.7, the Bethe ridge becomes clearly visible at relatively small recoil energies, of the order of the ionization energy $U_i$. For smaller $Q$’s, the structure of the Bethe surface is characteristic of the material. In the limit $Q \to 0$, the GOS reduces to the optical oscillator strength (OOS),

$$\frac{df(W)}{dW} = \frac{df(Q = 0, W)}{dW},$$  \hfill (3.44)

which is closely related to the (dipole) photoelectric cross section for photons of energy $W$ (Fano, 1963). Experimental information on the OOS is provided by measurements of either photoelectric cross sections or dielectric functions (see e.g. Fernández-Varea et al., 1993a). The GOS satisfies the Bethe sum rule (Inokuti, 1971)

$$\int_0^\infty \frac{df(Q, W)}{dW} \, dW = Z \quad \text{for any } Q.$$  \hfill (3.45)

This sum rule, which is a result from non-relativistic theory (see e.g. Mott and Massey, 1965), is assumed to be generally satisfied. It leads to the interpretation of the GOS as the effective number of electrons per unit energy transfer that participate in interactions with given recoil energy $Q$. The mean excitation energy $I$, defined by (Fano, 1963; Inokuti, 1971)

$$Z \ln I = \int_0^\infty \ln W \frac{df(W)}{dW} \, dW,$$  \hfill (3.46)
plays a central role in the Bethe stopping power formula [eq. (3.105) below]. This quantity has been determined empirically for a large number of materials (see Berger and Seltzer, 1982, and references therein) from measurements of the stopping power of heavy charged particles and/or from experimental optical dielectric functions. In the following, we shall assume that the mean excitation energy of the stopping medium is known.

### 3.2.1 GOS model

The simulation of inelastic collisions of electrons and positrons in PENELOPE is performed on the basis of the following GOS model, which is tailored to allow fast random sampling of $W$ and $Q$. We assume that the GOS splits into contributions from the different atomic electron shells. Each atomic shell $k$ is characterized by the number $f_k$ of electrons in the shell and the ionization energy $U_k$. To model the contribution of a shell to the GOS, we refer to the example of the hydrogen atom (fig. 3.7) and observe that for $Q > U_k$ the GOS reduces to the Bethe ridge, whereas for $Q < U_k$ it is nearly constant with $Q$ and decreases rapidly with $W$; a large fraction of the OOS concentrates in a relatively narrow $W$-interval. Consideration of other well-known systems, such as inner shells of heavy atoms (Manson, 1972) and the free-electron gas (Lindhard and Winther, 1964), shows that these gross features of the GOS are universal. Liljequist (1983) proposed modelling the GOS of each shell by means of a "$\delta$-oscillator", which is an entity with a simple excitation spectrum given by (see fig. 3.8)

$$
\frac{df_k(Q, W)}{dW} = f_k [\delta(W - W_k)\Theta(W_k - Q) + \delta(W - Q)\Theta(Q - W_k)] .
$$

(3.47)

The first term represents resonant low-$Q$ (distant) interactions, which are described as a single resonance at the energy $W_k$. The second term corresponds to large-$Q$ (close) interactions, in which the target electrons react as if they were free and at rest ($W = Q$). Notice that the oscillator GOS satisfies the sum rule

$$
\int_0^\infty \frac{df_k(Q, W)}{dW} dW = f_k \quad \text{for any } Q
$$

(3.48)

and, consequently, the oscillator strength $f_k$ can be identified with the number of electrons represented by the oscillator. Here we shall adopt Liljequist’s model and describe the excitations of a shell (or a group of shells) by means of a single oscillator, with the resonance energy considered as an adjustable parameter. The oscillator model gives a Bethe ridge with zero width, i.e. the broadening caused by the momentum distribution of the target electrons is neglected. This is not a serious drawback for light projectiles (electrons and positrons), but it can introduce sizeable errors in the computed cross sections for slow heavy projectiles with $m \gg m_e$. Our oscillator model also disregards the fact that, for low-$Q$ interactions, there is a transfer of oscillator strength from inner to outer shells (see e.g. Shiles et al., 1980).

In mixed (class II) simulations, only hard collisions, with energy loss larger than a specified cutoff value $W_{cc}$, are simulated (see chapter 4). The effect of soft interactions
Figure 3.8: Oscillator model for the GOS of an inner shell with \( U_k = 2 \) keV. The continuous
curve represents the maximum allowed energy loss as a function of the recoil energy, \( W_m(Q) \),
for electrons/positrons with \( E = 10 \) keV. For distant interactions the possible recoil energies lie
in the interval from \( Q_\pm \) to \( W_k \). Recoil energies larger than \( W_k \) correspond to close interactions.
The largest allowed energy loss \( W_{\text{max}} \) is \( E/2 \) for electrons and \( E \) for positrons (see text).

(with \( W < W_{\text{oc}} \)) is described by means of a multiple scattering approximation, which
does not require detailed knowledge of the shell DCSs. K and L shells with ionization
energy \( U_i \) larger than \( \max(200 \text{ eV}, W_{\text{oc}}) \) will be referred to as “inner” shells. The partial
GOS of an inner shell will be represented by the oscillator

\[
\frac{df_i(Q,W)}{dW} = f_i \left[ \delta(W - W_i)\Theta(W_i - Q) + \delta(W - Q)\Theta(Q - W_i) \right],
\]

(3.49)

with resonance energy \( W_i = 1.65 \) \( U_i \). This value of \( W_i \) is derived from the hydrogenic
model (see below), which is expected to be approximately valid for the innermost shells.
It should be noted that the shell ionization cross sections obtained from this model are
only roughly approximate. Their use in a Monte Carlo code is permissible only because
ionization of inner shells is a low-probability process (see fig. 3.9 below) that has a weak
effect on the global transport properties. In cases where inner-shell ionization is directly
observed (e.g., in the simulation of x-ray emission by electron bombardment), a more
accurate description of the process should be used.

The largest contribution to the total cross section arises from low-\( W \) (soft) excitations. Therefore, the total cross section is mostly determined by the OOS of weakly
bound electrons, which is strongly dependent on the state of aggregation. In the case
of conductors and semiconductors, electrons in the outermost shells form the conduction band (cb). These electrons can move quite freely through the medium and, hence,
their binding energy is set to zero, \( U_{cb} = 0 \). Excitations of the conduction band will be
described by a single oscillator,
\[
\frac{df_{cb}(Q, W)}{dW} = f_{cb} \left[ \delta(W - W_{cb}) \Theta(W_{cb} - Q) + \delta(W - Q) \Theta(Q - W_{cb}) \right].
\]
(3.50)

The parameters \(W_{cb}\) and \(f_{cb}\) should be identified with the plasmon energy and the effective number of electrons (per atom or molecule) that participate in plasmon excitations; these quantities can be deduced e.g. from electron energy-loss spectra or from measured optical data. When this information is not available, we will simply fix the value of \(f_{cb}\) (as the number of electrons with ionization energies less than, say, 15 eV) and set the resonance energy \(W_{cb}\) equal to the plasmon energy of a free-electron gas with the same density as that of conduction electrons,
\[
W_{cb} = \sqrt{\frac{4\pi N f_{cb} \hbar^2 e^2}{Z m_e}} = \sqrt{\frac{f_{cb}}{Z}} \Omega_p.
\]
(3.51)

This is a fairly realistic model for free-electron-like metals (such as aluminium), because the resonance energy is set equal to the plasmon energy of a free-electron gas (see e.g. Kittel, 1976).

Electron shells other than K and L shells, or with \(U_j < \max(200 \text{ eV}, W_{ce})\), will be referred to as “outer” shells. In the case of conductors, shells that contribute to the conduction band are excluded from the set of outer shells. Each outer shell is described by an oscillator
\[
\frac{df_j(Q, W)}{dW} = f_j \left[ \delta(W - W_j) \Theta(W_j - Q) + \delta(W - Q) \Theta(Q - W_j) \right],
\]
(3.52)

with resonance energy \(W_j = gU_j\), where \(g\) is an empirical adjustment factor (the same for all outer shells) which will be determined below. We thus arrive at the following expression for the GOS,
\[
\frac{df(Q, W)}{dW} = \sum_k \frac{df_k(Q, W)}{dW} = \sum_k f_k \left[ \delta(W - W_k) \Theta(W_k - Q) + \delta(W - Q) \Theta(Q - W_k) \right],
\]
(3.53)

where the summation in \(k\) extends over all inner and outer electron shells (and the conduction band, in the case of conductors). This GOS model satisfies the Bethe sum rule (3.45),
\[
\int_0^\infty \frac{df(Q, W)}{dW} dW = \sum_k f_k = Z \quad \text{for any } Q.
\]
(3.54)

The corresponding OOS reduces to
\[
\frac{df(Q = 0, W)}{dW} = \sum_k f_k \delta(W - W_k),
\]
(3.55)
which has the same form (a superposition of resonances) as the OOS used by Sternheimer (1952) in his calculations of the density effect correction. In order to reproduce the high-energy stopping power given by the Bethe formula (Berger and Seltzer, 1982), the excitation energies and oscillator strengths must lead, through eq. (3.46), to the accepted value of the mean excitation energy $I$, i.e.

$$\sum_k f_k \ln W_k = Z \ln I. \quad (3.56)$$

Following Sternheimer (1952), we use this relation to determine the empirical adjustment factor $g \left(= W_j/U_j\right)$,

$$\ln g = \left[Z \ln I - \sum_i f_i \ln W_i - f_{eb} \ln W_{eb} - \sum_j f_j \ln U_j\right] \left(\sum_j f_j\right)^{-1}. \quad (3.57)$$

For a one-shell system, such as the hydrogen atom, the constraint (3.56) implies that the resonance energy $W_i$ is equal to $I$ [see the definition (3.46)]. Considering the $\sim W^{-3}$ dependence of the hydrogenic OOS, it is concluded that $g$ should be of the order of $\exp(1/2) = 1.65$ (Sternheimer et al., 1982). It is worth noting that the Sternheimer adjustment factor $g$ is a property of the considered medium; therefore, the DCSs for ionization of an outer shell of a given element in two different compounds may be slightly different.

The present theory is directly applicable to compounds (and mixtures), since the oscillators may pertain either to atoms or molecules. When the value of the mean excitation energy of the compound is not known, it may be estimated from Bragg’s additivity rule as follows. Consider a compound $X_xY_y$, in which the molecules consist of $x$ atoms of the element $X$ and $y$ atoms of the element $Y$. The number of electrons per molecule is $Z_M = xZ_X + yZ_Y$, where $Z_X$ stands for the atomic number of element $X$. According to the additivity rule, the GOS of the compound is approximated as the sum of the atomic GOSs of the atoms so that

$$Z_M \ln I = xZ_X \ln I_X + yZ_Y \ln I_Y, \quad (3.58)$$

where $I_X$ denotes the mean excitation energy of element $X$.

For heavy elements, and also for compounds and mixtures with several elements, the number of electron shells may be fairly large (of the order of sixty for an alloy of two heavy metals). In these cases, it would be impractical to treat all shells with the same detail/accuracy. In fact, the description of the outer shells can be simplified without sacrificing the reliability of the simulation results. In PENELLOPE, the maximum number of electron shells for each material is limited. When the number of actual shells is too large, outer shells that have similar resonance energies are grouped together and described by a single oscillator. The grouping is made in such a way that the contribution of the group oscillator to the mean excitation energy $I$ equals the sum of contributions of the grouped oscillators; this ensures that grouping will not alter the stopping power of fast particles (with $E$ substantially greater than the ionization energy of the grouped oscillators).
3.2. Inelastic collisions

3.2.2 Differential cross sections

The DCS for inelastic collisions obtained from our GOS model can be split into contributions from distant longitudinal, distant transverse and close interactions,

$$\frac{d^2\sigma_{in}}{dW\,dQ} = \frac{d^2\sigma_{\text{dis,l}}}{dW\,dQ} + \frac{d^2\sigma_{\text{dis,t}}}{dW\,dQ} + \frac{d^2\sigma_{\text{cp}}}{dW\,dQ}. \quad (3.59)$$

The DCS for distant longitudinal interactions is given by the first term in eq. (3.43),

$$\frac{d^2\sigma_{\text{dis,l}}}{dW\,dQ} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{2m_e c^2}{W \, Q (Q + 2m_e c^2)} \delta(W - W_k) \Theta(W_k - Q). \quad (3.60)$$

As mentioned above, the DCS for distant transverse interactions has a complicated expression. To simplify it, we shall ignore the (very small) angular deflections of the projectile in these interactions and replace the expression in curly brackets in eq. (3.43) by an averaged $W$-independent value that gives the exact contribution of the distant transverse interactions to the high-energy stopping power (Salvat and Fernández-Varea, 1992). This yields the following approximate expression for the DCS of distant transverse interactions,

$$\frac{d^2\sigma_{\text{dis,t}}}{dW\,dQ} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W} \left\{ \ln \left( \frac{1}{1 - \beta^2} \right) - \beta^2 - \delta_F \right\}$$

$$\times \delta(W - W_k) \Theta(W_k - Q_-) \delta(Q - Q_-), \quad (3.61)$$

where $Q_-$ is the minimum recoil energy\footnote{The recoil energy $Q_-$ corresponds to $\theta = 0$, i.e. we consider that the projectile is not deflected by distant transverse interactions.} for the energy transfer $W$, eq. (A.31), and $\delta_F$ is the Fermi density effect correction on the stopping power, which has been studied extensively in the past (Sternheimer, 1952; Fano, 1963). $\delta_F$ can be computed as (Fano, 1963)

$$\delta_F \equiv \frac{1}{Z} \int_0^\infty \frac{df(Q = 0, W)}{dW} \ln \left( 1 + \frac{L^2}{W^2} \right) dW - \frac{L^2}{\Omega_p^2} (1 - \beta^2), \quad (3.62)$$

where $L$ is a real-valued function of $\beta^2$ defined as the positive root of the following equation (Inokuti and Smith, 1982):

$$\mathcal{F}(L) \equiv \frac{1}{Z} \Omega_p^2 \int_0^\infty \frac{1}{W^2 + L^2} \frac{df(Q = 0, W)}{dW} dW = 1 - \beta^2. \quad (3.63)$$

The function $\mathcal{F}(L)$ decreases monotonically with $L$, and hence, the root $L(\beta^2)$ exists only when $1 - \beta^2 < \mathcal{F}(0)$; otherwise it is $\delta_F = 0$. Therefore, the function $L(\beta^2)$ starts with zero at $\beta^2 = 1 - \mathcal{F}(0)$ and grows monotonically with increasing $\beta^2$. With the OOS, given by eq. (3.53), we have

$$\mathcal{F}(L) = \frac{1}{Z} \Omega_p^2 \sum_k f_k \frac{1}{W_k^2 + L^2}. \quad (3.64)$$
and

\[ \delta_F \equiv \frac{1}{Z} \sum_k f_k \ln \left( 1 + \frac{L^2}{W_k^2} \right) - \frac{L^2}{\Omega_p^2} \left( 1 - \beta^2 \right). \]  

(3.65)

In the high-energy limit \((\beta \to 1)\), the \(L\) value resulting from eq. (3.63) is large \((L \gg W_k)\) and can be approximated as \(L^2 = \Omega_p^2/(1 - \beta^2)\). Then, using the Bethe sum rule \((\sum f_k = Z)\) and the relation (3.56), we obtain

\[ \delta_F \simeq \ln \left( \frac{\Omega_p^2}{(1 - \beta^2)I^2} \right) - 1, \quad \text{when } \beta \to 1. \]  

(3.66)

The DCS for close collisions is given by

\[ \frac{d^2\sigma_{\text{cb}}}{dW \, dQ} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W} \left( \frac{2m_e e^2}{W(W + 2m_e e^2)} + \frac{\beta^2 \sin^2 \theta_{\text{cb}}}{2m_e e^2} \right) \delta(W - Q) \Theta(W - W_k), \]

where \(\theta_{\text{cb}}\) is the recoil angle, defined by eq. (3.40) with \(Q = W\),

\[ \cos^2 \theta_{\text{cb}} = \frac{W}{E} \frac{E + 2m_e e^2}{W + 2m_e e^2}. \]  

(3.67)

We have

\[ \frac{d^2\sigma_{\text{cb}}}{dW \, dQ} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W^2} \left( 1 + \frac{\beta^2 (E - W) W - EW}{E(W + 2m_e e^2)} \right) \delta(W - Q) \Theta(W - W_k). \]  

(3.68)

**DCS for close collisions of electrons**

When the projectile is an electron, the DCS must be corrected to account for the indistinguishability of the projectile and the target electrons. For distant interactions, the effect of this correction is small (much smaller than the distortion introduced by our modelling of the GOS) and will be neglected. The energy loss DCS for binary collisions of electrons with free electrons at rest, obtained from the Born approximation with proper account of exchange, is given by the Møller (1932) formula,

\[ \frac{d^2\sigma_M}{dW \, dQ} = \frac{2\pi e^4}{m_e v^2} \frac{1}{W^2} \left[ 1 + \left( \frac{W}{E - W} \right)^2 - \frac{W}{E - W} \right. \\
\left. + a \left( \frac{W}{E - W} + \frac{W^2}{E^2} \right) \right] \delta(W - Q), \]  

(3.69)

where

\[ a = \left( \frac{E}{E + m_e e^2} \right)^2 = \left( \frac{\gamma - 1}{\gamma} \right)^2. \]  

(3.70)

To introduce exchange effects in the DCS for close interactions of electrons, we replace the factor in parenthesis in eq. (3.68) by the analogous factor in Møller’s formula, i.e. we take

\[ \frac{d^2\sigma_{\text{cb}}^{(-)}}{dW \, dQ} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W^2} F^{(-)}(E, W) \delta(W - Q) \Theta(W - W_k), \]  

(3.71)
3.2. Inelastic collisions

with

\[ F(E, W) \equiv 1 + \left( \frac{W}{E - W} \right)^2 - \frac{W}{E - W} + a \left( \frac{W}{E - W} \right)^2. \]  \( (3.72) \)

In the final state we have two indistinguishable free electrons, and it is natural to consider the fastest one as the “primary”. Accordingly, the maximum allowed energy transfer in close collisions is

\[ W_{\text{max}} = E/2. \]  \( (3.73) \)

After ionizing an inner shell \( i \), the primary electron has a kinetic energy \( E - W \), the “secondary” electron (delta ray) is ejected with kinetic energy \( W - U_i \), and the residual ion is left in an excited state, with a vacancy in the inner shell \( i \), which corresponds to an excitation energy equal to \( U_i \). This energy is released by emission of x rays and Auger electrons (see section 2.6). When the ionization occurs in an outer shell or in the conduction band, the initial energy of the secondary electron is set equal to \( W \) and no fluorescent radiation from the ionized atom is followed by the simulation program. This is equivalent to assuming that the secondary electron carries away the excitation energy of the target atom.

DCS for close collisions of positrons

Positrons in matter are unstable particles that annihilate with electrons giving photons (see section 3.4). On the other hand, electron-positron pairs can be created if enough electromagnetic energy (> \( 2m_e c^2 \)) is available (either from real or virtual photons). A positron does not interact with matter as a usual (stable) positively charged particle, since the competing process of annihilation followed by re-creation can cause the same transitions as “direct” scattering (see e.g. Sakurai, 1967). The DCS for binary collisions of positrons with free electrons at rest, obtained from the first Born approximation including the “annihilation/creation” mechanism, is given by the Bhabha (1936) formula,

\[ \frac{d^2 \sigma_B}{dW dQ} = \frac{2\pi e^4}{m_e c^2} \frac{1}{W^2} \left[ 1 - b_1 \frac{W}{E} + b_2 \left( \frac{W}{E} \right)^2 - b_3 \left( \frac{W}{E} \right)^3 + b_4 \left( \frac{W}{E} \right)^4 \right] \delta(W - Q), \]  \( (3.74) \)

where

\[ b_1 = \left( \frac{\gamma - 1}{\gamma} \right)^2 \frac{2(\gamma + 1)^2 - 1}{\gamma^2 - 1}, \quad b_2 = \left( \frac{\gamma - 1}{\gamma} \right)^2 \frac{3(\gamma + 1)^2 + 1}{(\gamma + 1)^2}, \]

\[ b_3 = \left( \frac{\gamma - 1}{\gamma} \right)^2 \frac{2\gamma(\gamma - 1)}{(\gamma + 1)^2}, \quad b_4 = \left( \frac{\gamma - 1}{\gamma} \right)^2 \frac{(\gamma - 1)^2}{(\gamma + 1)^2}. \]  \( (3.75) \)

To account approximately for the effect of annihilation/creation on the DCS for close inelastic interactions of positrons, we shall use the expression (3.68), with the factor in parenthesis replaced by the Bhabha factor,

\[ F(E, W) = 1 - b_1 \frac{W}{E} + b_2 \left( \frac{W}{E} \right)^2 - b_3 \left( \frac{W}{E} \right)^3 + b_4 \left( \frac{W}{E} \right)^4. \]  \( (3.76) \)
That is,
\[
\frac{d^2\sigma_{\text{el}}^{(+)}}{dW \, dQ} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W^2} F^{(+)}(E, W) \delta(W - Q) \Theta(W - W_k).
\] (3.77)

The ionization process is simulated as in the case of electrons. In the ionization of an inner shell \(i\), the delta ray is emitted with initial kinetic energy equal to \(W - U_i\) and the subsequent relaxation of the residual ion is followed. When the ionization is in an outer shell (or in the conduction band), the secondary electron is assumed to be ejected with kinetic energy \(W\) and no fluorescent radiation is followed. Notice that the maximum energy loss in collisions of positrons with energy \(E\) is \(W_{\text{max}} = E\).

### 3.2.3 Integrated cross sections

The energy-loss DCS is defined as
\[
\frac{d\sigma_{\text{in}}}{dW} \equiv \int_{Q_-}^{Q_+} \frac{d^2\sigma_{\text{in}}}{dW \, dQ} dQ = \frac{d\sigma_{\text{dis},1}}{dW} + \frac{d\sigma_{\text{dis},t}}{dW} + \frac{d\sigma_{\text{clo}}}{dW},
\] (3.78)

where \(Q_-\) and \(Q_+\) are the minimum and maximum kinematically allowed recoil energies given by eq. (A.31). The contributions from distant longitudinal and transverse interactions are
\[
\frac{d\sigma_{\text{dis},1}}{dW} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W_k} \ln \left( \frac{W_k Q_- + 2m_e e^2}{Q_- W_k + 2m_e e^2} \right) \delta(W - W_k) \Theta(W_k - Q_-)
\] (3.79)

and
\[
\frac{d\sigma_{\text{dis},t}}{dW} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W_k} \left\{ \ln \left( \frac{1}{1 - \beta^2} \right) - \beta^2 - \delta_F \right\} \delta(W - W_k) \Theta(W_k - Q_-),
\] (3.80)

respectively. The energy-loss DCS for close collisions is
\[
\frac{d\sigma_{\text{cl}}^{(\pm)}}{dW} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W^2} F^{(\pm)}(E, W) \Theta(W - W_k).
\] (3.81)

The PDF of the energy loss in a single inelastic collision is given by
\[
p_{\text{in}}(W) = \frac{1}{\sigma_{\text{in}}} \frac{d\sigma_{\text{in}}}{dW},
\] (3.82)

where
\[
\sigma_{\text{in}} = \int_0^{W_{\text{max}}} \frac{d\sigma_{\text{in}}}{dW} dW
\] (3.83)

is the total cross section for inelastic interactions. It is convenient to introduce the quantities
\[
\sigma_{\text{in}}^{(n)}(W) \equiv \int_0^{W_{\text{max}}} W^n \frac{d\sigma_{\text{in}}}{dW} dW = \sigma_{\text{in}}^{(n)} \int_0^{W_{\text{max}}} W^n p_{\text{in}}(W) dW = \sigma_{\text{in}}^{(n)} \langle W^n \rangle,
\] (3.84)
where $\langle W^n \rangle$ denotes the $n$-th moment of the energy loss in a single collision (notice that $\sigma_{\text{in}}^{(0)} = \sigma_{\text{in}}$, $\sigma_{\text{in}}^{(1)}$ and $\sigma_{\text{in}}^{(2)}$ are known as the stopping cross section and the energy straggling cross section (for inelastic collisions), respectively.

The mean free path $\lambda_{\text{in}}$ for inelastic collisions is

\[
\lambda_{\text{in}}^{-1} = N \sigma_{\text{in}},
\]

(3.85)

where $N$ is the number of scattering centres (atoms or molecules) per unit volume. The stopping power $S_{\text{in}}$ and the energy straggling parameter $\Omega_{\text{in}}^2$ are defined by

\[
S_{\text{in}} = N \sigma_{\text{in}}^{(1)} = \frac{\langle W \rangle}{\lambda_{\text{in}}},
\]

(3.86)

and

\[
\Omega_{\text{in}}^2 = N \sigma_{\text{in}}^{(2)} = \frac{\langle W^2 \rangle}{\lambda_{\text{in}}^2}.
\]

(3.87)

Notice that the stopping power gives the average energy loss per unit path length\(^4\). The physical meaning of the straggling parameter is less direct. Consider a monoenergetic electron (or positron) beam of energy $E$ that impinges normally on a foil of material of (small) thickness $ds$, and assume that the electrons do not scatter (i.e. they are not deflected) in the foil. The product $\Omega_{\text{in}}^2 ds$ then gives the variance of the energy distribution of the beam after traversing the foil (see also section 4.2).

The integrated cross sections $\sigma_{\text{in}}^{(n)}$ can be calculated as

\[
\sigma_{\text{in}}^{(n)} = \sigma_{\text{dis},1}^{(n)} + \sigma_{\text{dis},t}^{(n)} + \sigma_{\text{clo}}^{(n)}.
\]

(3.88)

The contributions from distant longitudinal and transverse interactions are

\[
\sigma_{\text{dis},1}^{(n)} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k W_k^{n-1} \ln \left( \frac{W_k Q - 2m_e c^2}{Q - W_k + 2m_e c^2} \right) \Theta(W_{\text{max}} - W_k)
\]

(3.89)

and

\[
\sigma_{\text{dis},t}^{(n)} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k W_k^{n-1} \left\{ \ln \left( \frac{1}{1 - \beta^2} \right) - \beta^2 - \delta_f \right\} \Theta(W_{\text{max}} - W_k),
\]

(3.90)

respectively. Notice that for distant interactions $W_{\text{max}} = E$, for both electrons and positrons.

The integrated cross sections for close collisions are

\[
\sigma_{\text{clo}}^{(n)} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \int_{W_k}^{W_{\text{max}}} W^{n-2} F^{(\pm)}(E, W) \, dW.
\]

(3.91)

In the case of electrons, the integrals in this formula are of the form

\[
J_n^{(-)} = \int W^{n-2} \left[ 1 + \left( \frac{W}{E - W} \right)^2 - \frac{(1 - a)W}{E - W} + \frac{aW^2}{E^2} \right] \, dW
\]

(3.92)

\(^4\)The term “stopping power” is somewhat misleading; in fact, $S_{\text{in}}$ has the dimensions of force.
and can be calculated analytically. For the orders 0, 1 and 2 we have

\[ J_0^{(-)} = -\frac{1}{W} + \frac{1}{E - W} + \frac{1 - a}{E} \ln \left( \frac{E - W}{W} \right) + \frac{aW}{E^2}, \quad (3.93) \]

\[ J_1^{(-)} = \ln W + \frac{E}{E - W} + (2 - a) \ln(E - W) + \frac{aW^2}{2E^2}, \quad (3.94) \]

and

\[ J_2^{(-)} = (2 - a)W + \frac{2E^2 - W^2}{E - W} + (3 - a)E \ln(E - W) + \frac{aW^3}{3E^2}. \quad (3.95) \]

For positrons, the integrals in (3.91),

\[ J_n^{(+)} = \int W^{n-2} \left[ 1 - b_1 \frac{W}{E} + b_2 \left( \frac{W}{E} \right)^2 - b_3 \left( \frac{W}{E} \right)^3 + b_4 \left( \frac{W}{E} \right)^4 \right] dW; \quad (3.96) \]

can also be evaluated analytically as

\[ J_0^{(+)} = -\frac{1}{W} - b_1 \frac{\ln W}{E} + b_2 \frac{W}{E^2} - b_3 \frac{W^2}{2E^3} + b_4 \frac{W^3}{3E^4}, \quad (3.97) \]

\[ J_1^{(+)} = \ln W - b_1 \frac{W}{E} + b_2 \frac{W^2}{2E^2} - b_3 \frac{W^3}{3E^3} + b_4 \frac{W^4}{4E^4} \quad (3.98) \]

and

\[ J_2^{(+)} = W - b_1 \frac{W^2}{2E} + b_2 \frac{W^3}{3E^2} - b_3 \frac{W^4}{4E^3} + b_4 \frac{W^5}{5E^4}. \quad (3.99) \]

Our analytical GOS model provides quite an accurate average description of inelastic collisions (see below). However, the continuous energy loss spectrum associated with single distant excitations of a given atomic electron shell is approximated here as a single resonance (a \( \delta \)-distribution). As a consequence, the simulated energy loss spectra show unphysically narrow peaks at energy losses that are multiples of the resonance energies. These spurious peaks are automatically smoothed out after multiple inelastic collisions and also when the bin width used to tally the energy loss distributions is larger than the difference between resonance energies of neighbouring oscillators.

Fig. 3.9 displays total inelastic cross sections for electrons in aluminium and gold, as well as contributions from various groups of shells, as functions of the kinetic energy of the projectile. The curves labelled “K” and “L1+...” represent cross sections for ionization in these shells. The cross section for ionization in a bound shell decreases rapidly with the shell ionization energy \( U_i \) (since energy transfers less than \( U_i \), which would promote the target electron to occupied states, are forbidden). As a consequence, collisions occur preferentially with electrons in the conduction band and in outer bound shells. Inner-shell ionization by electron/positron impact is a relatively unlikely process. It should be noted that our GOS model is too crude to provide an accurate description of inner-shell ionization. To illustrate this, fig. 3.9 includes K-shell ionization cross sections obtained from a more realistic semiempirical approximation (Mayol and Salvat, 1990), which agree reasonably well with experimental data. We see that there are significant
Figure 3.9: Total inelastic cross sections for electrons in aluminium and gold and contributions from the K-shell, L-shell, conduction band (cb) and outer shells, calculated from our model GOS ignoring density effect corrections (i.e, with $\delta_F = 0$). The short-dashed lines represent K-shell ionization cross sections calculated from a more elaborate theory (Mayol and Salvat, 1990), which yields results in close agreement with experimental data. Note: 1 barn$=10^{-24}$ cm$^2$.

Differences between the cross sections from the semiempirical approximation and the predictions of our model, which is designed to yield accurate stopping powers only. To get a realistic picture of inner-shell ionization, we should rely on much more elaborate physical schemes. In fact, even the Born approximation ceases to be appropriate for projectiles with kinetic energies near the ionization threshold.

Collision stopping powers for electrons in aluminium, silver and gold obtained from the present analytical model are compared with sample values from the ICRU 37 (1984) stopping power tables [given also in Berger and Seltzer (1982)] for $E \geq 10$ keV in fig. 3.10. At these energies, our results practically coincide with the values in the tables of reference. In fig. 3.11, inelastic mean free paths and stopping powers for low-energy electrons ($E = 100$ eV to 100 keV) in aluminium and gold obtained from the present model are compared with experimental data from several authors. We see that the theory predicts the energy variation of total integrated cross sections down to relatively low energies. It should be noted that the adopted value of $W_{cb}$, the resonance energy of conduction band electrons, has a strong effect on the calculated mean free paths. In the case of free-electron-like materials such as aluminium, $W_{cb}$ can be identified with
Figure 3.10: Collision stopping power $S_{in}/\rho$ for electrons and positrons in Al, Ag (×10) and Au (×100) as a function of the kinetic energy. Continuous and dashed curves are results from the present model. Crosses are data from the ICRU37 tables (1984) [also, Berger and Seltzer, 1982]. The dotted curves are predictions from the Bethe formula (3.105), for electrons and positrons.

the energy of plasmon excitations (which is the dominant energy-loss mechanism). For other solids, the outermost electrons have a broad energy loss spectrum and there is no simple way of predicting this parameter. Fortunately, the stopping power (and, hence, the global stopping process) is practically independent of the adopted value of $W_{cb}$. To generate the data for aluminium, Fig. 3.11, we have set $W_{cb} = 15$ eV, which is the measured energy of volume plasmons in the metal [eq. (3.51) with $f_{cb} = 3$ conduction electrons per atom gives $W_{cb} = 15.8$ eV]; in this case, the calculated mean free paths are seen to agree fairly well with measured data. In the case of gold, eq. (3.51) with $f_{cb} = 11$ conduction electrons per atom gives $W_{cb} = 30$ eV. Fig. 3.11 shows stopping powers and mean free paths for electrons in gold obtained with $W_{cb} = 30$ and 40 eV. We see that, as indicated above, the mean free path varies strongly with this parameter, but the stopping power is practically insensitive to it.

3.2.4 Stopping power of high-energy electrons and positrons

It is of interest to evaluate explicitly the stopping power for projectiles with high energies ($E \gg U_k$). We shall assume that $U_k \ll 2m_e c^2$ (for the most unfavourable case of the
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Figure 3.11: Collision mean free path and stopping power for low-energy electrons in Al and Au. The plotted quantities are $\rho \lambda_{\text{in}}$ and $S_{\text{in}}/\rho$. Special symbols are experimental data from different sources (see Fernández-Varea et al., 1993a); closed symbols for mean free paths and open symbols for stopping powers.

K shell of heavy elements, $U_k$ is of the order of $2m_e e^2/10$. Under these circumstances, $Q_\perp \ll 2m_e e^2$ and we can use the approximation [see eq. (A.35)]

$$Q_\perp \simeq W_k^2/(2m_e e^2 \beta^2).$$  \hspace{1cm} (3.100)

The contribution from distant (longitudinal and transverse) interactions to the stopping cross section is then [see eqs. (3.79) and (3.80)]

$$\sigma_{\text{dis}}^{(1)} \simeq \frac{2\pi e^4}{m_e v^2} \sum_k f_k \left\{ \ln \left( \frac{2m_e e^2}{W_k} \right) + \ln \left( \frac{1}{1 - \beta^2} \right) - \beta^2 - \delta_F \right\}. \hspace{1cm} (3.101)$$

The contribution of close interactions is given by

$$\sigma_{\text{clo}}^{(1)} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \int_{W_k}^{W_{\text{max}}} W^{-1} F^4(E, W) \, dW.$$

Recalling that $E \gg U_k$, we have

$$\sigma_{\text{clo}}^{(1)} \simeq \frac{2\pi e^4}{m_e v^2} \sum_k f_k \left\{ \ln \left( \frac{E}{W_k} \right) + 1 - \left[ 1 + \beta^2 + 2\sqrt{1 - \beta^2} \right] \ln 2 \right. \left. + \frac{1}{8} \left( 1 - \sqrt{1 - \beta^2} \right)^2 \right\}. \hspace{1cm} (3.103)$$
for electrons and
\[
\sigma_{\text{el}}^{(1)} \approx \frac{2\pi e^4}{m_e v^2} \sum_k f_k \left\{ \ln \left( \frac{E}{W_k} \right) - b_1 + \frac{b_2}{2} - \frac{b_3}{3} + \frac{b_4}{4} \right\}.
\] (3.104)

for positrons. Adding the distant and close stopping cross sections, and using the relation (3.56), we arrive at the familiar Bethe formula for the stopping power,
\[
S_{\text{in}} = N \left( \sigma_{\text{dis}}^{(1)} + \sigma_{\text{cl}}^{(1)} \right) = N \frac{2\pi e^4}{m_e v^2} Z \left\{ \ln \left( \frac{E^2 \gamma + 1}{I^2} \right) + f^{(\pm)}(\gamma) - \delta_f \right\},
\] (3.105)

where
\[
f^{(-)}(\gamma) = 1 - \beta^2 - \frac{2\gamma - 1}{\gamma^2} \ln 2 + \frac{1}{8} \left( \frac{\gamma - 1}{\gamma} \right)^2.
\] (3.106)

and
\[
f^{(+)}(\gamma) = 2 \ln 2 - \frac{\beta^2}{12} \left[ 23 + \frac{14}{\gamma + 1} + \frac{10}{(\gamma + 1)^2} + \frac{4}{(\gamma + 1)^3} \right].
\] (3.107)

for electrons and positrons, respectively. This formula can be derived from very general arguments that do not require knowing the fine details of the GOS; the only information needed is contained in the Bethe sum rule (3.45) and in the definition (3.46) of the mean excitation energy (see e.g. Fano, 1963). Since our approximate analytical GOS model is physically motivated, it satisfies the sum rule and reproduces the adopted value of the mean ionization energy, it yields the exact Bethe formula.

It is striking that the “asymptotic” Bethe formula is in fact valid down to fairly small energies, of the order of 10 keV for high-Z materials (see fig. 3.10). It also accounts for the differences between the stopping powers of electrons and positrons (to the same degree as our GOS model approximation).

For ultrarelativistic projectiles, for which the approximation (3.66) holds, the Bethe formula simplifies to
\[
S_{\text{in}} \approx N \frac{2\pi e^4}{m_e v^2} Z \left\{ \ln \left( \frac{E^2}{W_p} \gamma + 1 \right) + f^{(\pm)}(\gamma) + 1 \right\}.
\] (3.108)

The mean excitation energy \( I \) has disappeared from this formula, showing that at very high energies the stopping power depends only on the electron density \( N Z \) of the medium.

### 3.2.5 Simulation of hard inelastic collisions

The DCSs given by expressions (3.78)-(3.81) permit the random sampling of the energy loss \( W \) and the angular deflection \( \theta \) by using purely analytical methods. In the following we consider the case of mixed (class II) simulation, in which only hard collisions, with energy loss larger than a specified cutoff value \( W_{\text{ce}} \), are simulated (see chapter 4). As
the value of the cutoff energy loss can be selected arbitrarily, the sampling algorithm can also be used in detailed (interaction-by-interaction) simulations \((W_{cc} = 0)\).

The first stage of the simulation is the selection of the active oscillator, for which we need to know the restricted total cross section,

\[
\sigma(W_{cc}) = \int_{W_{cc}}^{W_{max}} \frac{d\sigma_{\text{in}}}{dW} dW = \sigma_{\text{dis},t}(W_{cc}) + \sigma_{\text{dis},l}(W_{cc}) + \sigma_{\text{dis},o}(W_{cc})
\]

\[
= \sum_k \sigma_k(W_{cc}),
\]

(3.109)
as well as the contribution of each oscillator, \(\sigma_k(W_{cc})\). The active oscillator is sampled from the point probabilities \(p_k = \sigma_k(W_{cc})/\sigma(W_{cc})\). Since these probabilities are calculated analytically, the sampling algorithm is relatively slow. In mixed simulations, the algorithm can be sped up by using a larger cutoff energy loss \(W_{cc}\), which eliminates all the oscillators with \(W_k < W_{cc}\) from the sum.

After selecting the active oscillator, the oscillator branch (distant or close) is determined and, finally, the variables \(W\) and \(Q\) (or \(\cos \theta\)) are sampled from the associated DCS. For close collisions, \(Q = W\) and, therefore, the scattering angle is obtained directly from the energy loss.

**Distant interactions**

In distant interactions with the \(k\)-th oscillator, \(W = W_k\). The contributions of transverse and longitudinal interactions to the restricted cross section define the relative probabilities of these interaction modes. If the interaction is (distant) transverse, the angular deflection of the projectile is neglected, i.e. \(\cos \theta = 1\). For distant longitudinal collisions, the (unnormalized) PDF of \(Q\) is given by [see eq. (3.60)]

\[
P_{\text{dis}}(Q) = \begin{cases} 
1 & \text{if } Q_- < Q < W_k, \\
0 & \text{otherwise},
\end{cases}
\]

(3.110)

where \(Q_-\) is the minimum recoil energy, eq. (A.31). Random sampling from this PDF can be performed by the inverse transform method, which gives the sampling formula

\[
Q = Q_S \left\{ \left[ \frac{Q_S}{W_k} \left( 1 + \frac{W_k}{2m_e c^2} \right) \right]^{\frac{1}{\varepsilon}} - \frac{Q_S}{2m_e c^2} \right\}^{-1},
\]

(3.111)

where

\[
Q_S = \frac{Q_-}{1 + Q_-/(2m_e c^2)}.
\]

(3.112)

Once the energy loss and the recoil energy have been sampled, the polar scattering angle \(\theta\) is determined from eq. (A.40),

\[
\cos \theta = \frac{E(E + 2m_e c^2) + (E - W)(E - W + 2m_e c^2) - Q(Q + 2m_e c^2)}{2\sqrt{E(E + 2m_e c^2)(E - W)(E - W + 2m_e c^2)}},
\]

(3.113)
The azimuthal scattering angle $\phi$ is sampled uniformly in the interval $(0, 2\pi)$.

**Hard close collisions of electrons**

For the formulation of the sampling algorithm, it is convenient to introduce the reduced energy loss $\kappa \equiv W/E$. The PDF of $\kappa$ in close collisions of electrons with the $k$-th oscillator is given by [see eqs. (3.71) and (3.72)]

$$P_k^{(-)}(\kappa) \equiv \kappa^{-2} E^{(-)}(E, W) \Theta(\kappa - \kappa_c) \Theta\left(\frac{1}{2} - \kappa\right) = \left[\frac{1}{\kappa^2} + \frac{1}{(1 - \kappa)^2}\right] - \frac{1}{\kappa(1 - \kappa)} + a \left[1 + \frac{1}{\kappa(1 - \kappa)}\right] \Theta(\kappa - \kappa_c) \Theta\left(\frac{1}{2} - \kappa\right),$$

(3.114)

with $\kappa_c \equiv \max(W_k, W_{cc})/E$. Notice that the maximum allowed value of $\kappa$ is 1/2. Here, normalization is irrelevant.

We introduce the distribution

$$\Phi^{(-)}(\kappa) \equiv (\kappa^{-2} + 5a) \Theta(\kappa - \kappa_c) \Theta\left(\frac{1}{2} - \kappa\right), \quad a \equiv \left(\frac{\gamma - 1}{\gamma}\right)^2.$$  

(3.115)

It may be shown that $\Phi^{(-)} \geq P_k^{(-)}$ in the interval $(\kappa_c, \frac{1}{2})$. Therefore, we can sample the reduced energy loss $\kappa$ from the PDF (3.114) by using the rejection method (see section 1.2.4) with trial values sampled from the distribution (3.114) and acceptance probability $P_k^{(-)} / \Phi^{(-)}$.

Random sampling from the PDF (3.115), can be performed by using the composition method (section 1.2.5). We consider the following decomposition of the (normalized) PDF given by eq. (3.115):

$$\Phi^{(-)}_{\text{norm}}(\kappa) = \frac{1}{1 + 5a\kappa_c/2} \left[p_1(\kappa) + (5a\kappa_c/2)p_2(\kappa)\right],$$

(3.116)

where

$$p_1(\kappa) = \frac{\kappa_c}{1 - 2\kappa_c} \kappa^{-2}, \quad p_2(\kappa) = \frac{2}{1 - 2\kappa_c}.$$  

(3.117)

are normalized PDFs in the interval $(\kappa_c, \frac{1}{2})$. Random values of $\kappa$ from the PDF (3.115) can be generated by using the following algorithm:

(i) Generate $\xi$.

(ii) Set $\zeta = (1 + 5a\kappa_c/2)\xi$.

(iii) If $\zeta < 1$, deliver the value $\kappa = \kappa_c/[1 - \zeta(1 - 2\kappa_c)]$.

(iv) If $\zeta > 1$, deliver the value $\kappa = \kappa_c + (\zeta - 1)(1 - 2\kappa_c)/(5a\kappa_c)$.

The rejection algorithm for random sampling of $\kappa$ from the PDF (3.114) proceeds as follows:
3.2. Inelastic collisions

(i) Sample \( \kappa \) from the distribution given by eq. (3.115).

(ii) Generate a random number \( \xi \).

(iii) If \( \xi (1 + 5a \kappa^2) < \kappa^2 P_k^{(-)}(\kappa) \), deliver \( \kappa \).

(iv) Go to step (i).

Notice that in the third step we accept the \( \kappa \) value with probability \( P_k^{(-)} / \Phi^{(-)} \), which approaches unity when \( \kappa \) is small.

The efficiency of this sampling method depends on the values of the energy \( E \) and the cutoff reduced energy loss \( \kappa_c \), as shown in table 3.1. For a given energy and for \( W_{ce} \) values which are not too large, the efficiency increases when \( W_{ce} \) decreases.

<table>
<thead>
<tr>
<th>( E ) (eV)</th>
<th>( \kappa_c )</th>
</tr>
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<tbody>
<tr>
<td></td>
<td>0.001</td>
</tr>
<tr>
<td>( 10^3 )</td>
<td>99.9</td>
</tr>
<tr>
<td>( 10^6 )</td>
<td>99.7</td>
</tr>
<tr>
<td>( 10^7 )</td>
<td>99</td>
</tr>
<tr>
<td>( 10^9 )</td>
<td>99</td>
</tr>
</tbody>
</table>

After sampling the energy loss \( W = \kappa E \), the polar scattering angle \( \theta \) is obtained from eq. (A.40) with \( Q = W \). This yields

\[
\cos^2 \theta = \frac{E - W}{E + 2m_e c^2} \frac{E + 2m_e c^2}{E - W + 2m_e c^2},
\]

which agrees with eq. (A.17). The azimuthal scattering angle \( \phi \) is sampled uniformly in the interval \((0, 2\pi)\).

**Hard close collisions of positrons**

The PDF of the reduced energy loss \( \kappa \equiv W/E \) in positron close collisions with the \( k \)-th oscillator is given by [see eqs. (3.76) and (3.77)]

\[
P_k^{(+)}(\kappa) = \kappa^{-2} f_k^{(+)}(E, W) \Theta(\kappa - \kappa_c) \Theta(1 - \kappa)
= \left[ \frac{1}{\kappa^2} - \frac{b_1}{\kappa} + b_2 - b_3 \kappa + b_4 \kappa^2 \right] \Theta(\kappa - \kappa_c) \Theta(1 - \kappa)
\]
with $\kappa_c \equiv \max(W_k, W_{ce})/E$. The maximum allowed reduced energy loss is 1. Again, normalization is not important.

Consider the distribution

$$\Phi^{(+)}(\kappa) \equiv \kappa^{-2} \Theta(\kappa - \kappa_c) \Theta(1 - \kappa).$$

(3.120)

It is easy to see that $\Phi^{(+)} > P^{(+)}_k$ in the interval $(\kappa_c, 1)$. Therefore, we can generate $\kappa$ from the PDF, eq. (3.119), by using the rejection method with trial values sampled from the distribution of eq. (3.120) and acceptance probability $P^{(+)}_k/\Phi^{(+)}$. Sampling from the PDF $\Phi^{(+)}$ can easily be performed with the inverse transform method.

The algorithm for random sampling from the PDF (3.119), is:

(i) Sample $\kappa$ from the PDF (3.120), as $\kappa = \kappa_c/[1 - \xi(1 - \kappa_c)]$.

(ii) Generate a new random number $\xi$.

(iii) If $\xi < \kappa^2 P^{(+)}_k(\kappa)$, deliver $\kappa$.

(iv) Go to step (i).

The efficiency of this algorithm, for given values of the kinetic energy and the cutoff reduced energy loss $\kappa_c$, practically coincides with that of the algorithm for electron collisions described above (see table 3.1).

Secondary electron emission

According to our GOS model, each oscillator $W_k$ corresponds to a shell with $f_k$ electrons and ionization energy $U_k$. As discussed above (see section 3.2.2), in the case of ionization of an inner shell $i$ we consider that a secondary electron (delta ray) is emitted with energy $E_s = W - U_i$ and that the residual ion is left with a vacancy in the shell $i$. In the case of ionization of outer shells, the simulated delta ray is emitted with kinetic energy $E_s = W$ and the target atom is assumed to remain in its ground state. To set the initial direction of the delta ray, we assume that the target electron was initially at rest, i.e. the delta ray is emitted in the direction of the momentum transfer $\mathbf{q}$. This implies that the polar emission angle $\theta_\mathbf{q}$ (see fig. 3.1) coincides with the recoil angle $\theta_r$ [which is given by eq. (A.42)],

$$\cos^2 \theta_\mathbf{q} = \frac{W^2/\beta^2}{Q(Q + 2m_e c^2)} \left( 1 + \frac{Q(Q + 2m_e c^2) - W^2}{2W(E + m_e c^2)} \right) ^2.$$  

(3.121)

In the case of close collisions ($Q = W$), this expression simplifies to

$$\cos \theta_\mathbf{q}(Q = W) = \left( \frac{W(E + 2m_e c^2)}{E(W + 2m_e c^2)} \right)^{1/2},$$

(3.122)

which agrees with the result for binary collisions with free electrons at rest, see eq. (A.18). Since the momentum transfer lies on the scattering plane (i.e. on the plane
formed by the initial and final momenta of the projectile), the azimuthal emission angle is \( \phi_s = \pi + \phi \).

In reality, the target electrons are not at rest and, therefore, the angular distribution of emitted delta rays is broad. Since the average momentum of bound electrons is zero, the average direction of delta rays coincides with the direction of \( \mathbf{q} \). Thus, our simple emission model correctly predicts the average initial direction of delta rays, but disregards the “Doppler broadening” of the angular distribution. This is not a serious drawback, because secondary electrons are usually emitted with initial kinetic energies that are much smaller than the initial energy of the projectile. This means that the direction of motion of the delta ray is randomized, by elastic and inelastic collisions, after a relatively short path length (much shorter than the transport mean free path of the projectile).

### 3.3 Bremsstrahlung emission

As a result of the acceleration caused by the electrostatic field of atoms, swift electrons (or positrons) emit bremsstrahlung (braking radiation). In each bremsstrahlung event, an electron with kinetic energy \( E \) generates a photon of energy \( W \), which takes values in the interval from 0 to \( E \). The process is described by an atomic DCS, differential in the energy loss \( W \), the final direction of the projectile and the direction of the emitted photon (Koch and Motz, 1959; Tsai, 1974). The habitual practice in Monte Carlo simulation is to sample the energy loss from the single-variable distribution obtained by integrating the DCS over the other variables. This permits the generation of \( W \) easily, but information on the angular distributions is completely lost and has to be regained from suitable approximations. Angular deflections of the projectile are considered to be accounted for by the elastic scattering DCS and, consequently, the direction of movement of the projectile is kept unaltered in the simulation of radiative events.

#### 3.3.1 The energy-loss scaled DCS

A simple description of the bremsstrahlung DCS is provided by the Bethe-Heitler formula with screening, which is derived within the Born approximation (Bethe and Heitler, 1934; Tsai, 1974). Although this formula is valid only when the kinetic energy of the electron before and after photon emission is much larger than its rest energy \( m_e c^2 \), it accounts for the most relevant features of the emission process. Within the Born approximation, bremsstrahlung emission is closely related to electron-positron pair production. In particular, the Bethe-Heitler DCS formulae for pair production and bremsstrahlung emission involve the same screening functions. Considering the exponential screening model (2.78), the Bethe-Heitler DCS for bremsstrahlung emission by electrons in the field of an atom of atomic number \( Z \) and screening radius \( R \) can be expressed as (Salvat
and Fernández-Varea, 1992)

\[
\frac{d\sigma_{br}}{dW} = r_0^2 \alpha Z(Z + \eta) \frac{1}{W} \left[ e^2 \varphi_1(b) + \frac{4}{3} (1 - \epsilon) \varphi_2(b) \right], \tag{3.123}
\]

where \( \alpha \) is the fine-structure constant, \( r_e \) is the classical electron radius,

\[
\epsilon = \frac{W}{E + m_e c^2} = \frac{W}{\gamma m_e c^2}, \quad b = \frac{R m_e c}{\hbar} \frac{1}{2\gamma} \frac{\epsilon}{1 - \epsilon}, \tag{3.124}
\]

and

\[
\varphi_1(b) = 4 \ln(R m_e c/\hbar) + 2 - 2 \ln(1 + b^2) - 4b \arctan(b^{-1}),
\]

\[
\varphi_2(b) = 4 \ln(R m_e c/\hbar) + \frac{7}{3} - 2 \ln(1 + b^2) - 6b \arctan(b^{-1})
\]

\[- b^2 \left[ 4 - 4b \arctan(b^{-1}) - 3 \ln(1 + b^{-2}) \right]. \tag{3.125}
\]

The quantity \( \eta \) in eq. (3.123) accounts for the production of bremsstrahlung in the field of the atomic electrons (see e.g. Seltzer and Berger, 1985); in the high-energy limit \( \eta \approx 1.2 \).

The Bethe-Heitler formula indicates that, for a given value of \( Z \), the quantity \( W d\sigma_{br}/dW \) varies smoothly with \( E \) and \( W \). It is therefore customary to express the DCS for bremsstrahlung emission by electrons in the form

\[
\frac{d\sigma_{br}}{dW} = \frac{Z^2}{\beta^2} \frac{1}{W} \chi(Z, E, \kappa), \tag{3.126}
\]

where \( W \) is the energy of the emitted photon, \( \kappa \) is the reduced photon energy, defined as

\[
\kappa \equiv W/E, \tag{3.127}
\]

which takes values between 0 and 1. The quantity

\[
\chi(Z, E, \kappa) = (\beta^2/Z^2)W \frac{d\sigma_{br}}{dW} \tag{3.128}
\]

is known as the “scaled” bremsstrahlung DCS; for a given element \( Z \), it varies smoothly with \( E \) and \( \kappa \). Seltzer and Berger (1985, 1986) produced extensive tables of the scaled DCS for all the elements \( Z = 1 \rightarrow 92 \) and for electron energies from 1 keV to 10 GeV. They tabulated the scaled DCSs for emission in the (screened) field of the nucleus (electron-nucleus bremsstrahlung) and in the field of atomic electrons (electron-electron bremsstrahlung) separately, as well as their sum, the total scaled DCS. The electron-nucleus bremsstrahlung DCS was calculated by combining analytical high-energy theories with results from partial-wave calculations by Pratt et al. (1977) for bremsstrahlung emission in screened atomic fields and energies below 2 MeV. The scaled DCS for
3.3. Bremsstrahlung emission

electron-electron bremsstrahlung was obtained from the theory of Haug (1975) combined with a screening correction that involves Hartree-Fock incoherent scattering functions. Seltzer and Berger’s scaled DCS tables constitute the most reliable theoretical representation of bremsstrahlung energy spectra available at present.

The PENELope database of scaled bremsstrahlung DCSs consists of 92 files, one for each element from hydrogen to uranium, which were generated from the original database of Seltzer and Berger. The file of the element Z contains the values of \( \chi(Z, E_i, \kappa_j) \) for a set of electron kinetic energies \( E_i \), which covers the range from 1 keV to 10 GeV and is suitably spaced to allow accurate natural cubic spline interpolation in \( \ln E \). For each energy \( E_i \) in this grid, the table contains the values of the scaled DCS for a given set of 32 reduced photon energies \( \kappa_j \) (the same for all elements), which span the interval (0,1), with a higher density at the upper end of this interval to reproduce the structure of the bremsstrahlung “tip” (see fig. 3.12). The spacing of the \( \kappa \)-grid is dense enough to allow linear interpolation of \( \chi(Z, E_i, \kappa_j) \) in \( \kappa \).

\[ \begin{align*}
\frac{d\sigma_{\text{br, mol}}}{dW} &= \frac{Z_X^2}{\beta^2} \frac{1}{W} \chi(Z_X, E, \kappa) + \frac{Z_Y^2}{\beta^2} \frac{1}{W} \chi(Z_Y, E, \kappa). 
\end{align*} \] (3.129)

**Figure 3.12:** Numerical scaled bremsstrahlung energy-loss DCSs of Al and Au for electrons with the indicated energies (Seltzer and Berger, 1986).

In the case of compounds (or mixtures) we use the additivity rule and compute the molecular DCS as the sum of the DCSs of all the atoms in a molecule. Consider a compound \( X_x Y_y \), whose molecules consist of \( x \) atoms of the element \( X \) and \( y \) atoms of the element \( Y \). The molecular DCS is
To simulate each radiative event in a compound, we should first select the element (X or Y) where the emission occurs and then sample the photon energy and direction from the corresponding atomic DCS. This is a lengthy process and requires storing the scaled DCSs for all the elements present. To simplify the simulation, we shall express the molecular DCS in the same form as the atomic DCS, eq. (3.126),

\[
\frac{d\sigma_{br,\text{mol}}}{dW} = \frac{Z_{eq}^2}{\beta^2} \frac{1}{W} \chi_{\text{mol}}(Z_{eq}, E, \kappa),
\]

(3.130)

where

\[
Z_{eq}^2 \equiv \frac{1}{x + y} \left(xZ_X^2 + yZ_Y^2\right)
\]

(3.131)

is the “equivalent” atomic number \( Z_{eq} \) and

\[
\chi_{\text{mol}}(Z_{eq}, E, \kappa) = \frac{xZ_X^2}{Z_{eq}^2} \chi(Z_X, E, \kappa) + \frac{yZ_Y^2}{Z_{eq}^2} \chi(Z_Y, E, \kappa)
\]

(3.132)

is the molecular scaled DCS. Radiative events will be sampled directly from the molecular DCS (3.130). This method may introduce slight inconsistencies in the angular distribution of the emitted photons (see below), which usually have a negligible effect on the simulation results.

The radiative DCS for positrons reduces to that of electrons in the high-energy limit but is smaller for intermediate and low energies. Owing to the lack of more accurate calculations, the DCS for positrons is obtained by multiplying the electron DCS by a \( \kappa \)-independent factor, i.e.

\[
\frac{d\sigma_{\text{he}^+}}{dW} = F_p(Z, E) \frac{d\sigma_{\text{he}^-}}{dW}.
\]

(3.133)

The factor \( F_p(Z, E) \) is set equal to the ratio of the radiative stopping powers for positrons and electrons, which has been calculated by Kim et al. (1986) (cf. Berger and Seltzer, 1982). In the calculations we use the following analytical approximation

\[
F_p(Z, E) = 1 - \exp(-1.2359 \times 10^{-1} t + 6.1274 \times 10^{-2} t^2 - 3.1516 \times 10^{-2} t^3 \\
+ 7.7446 \times 10^{-3} t^4 - 1.0595 \times 10^{-3} t^5 + 7.0568 \times 10^{-5} t^6 \\
- 1.8080 \times 10^{-6} t^7),
\]

(3.134)

where

\[
t = \ln \left(1 + \frac{10^6 E}{Z^2 m_e c^2}\right).
\]

(3.135)

Expression (3.134) reproduces the values of \( F_p(Z, E) \) tabulated by Kim et al. (1986) to an accuracy of about 0.5%.
3.3.2 Integrated cross sections

The total cross section for bremsstrahlung emission is infinite due to the divergence of the DCS (3.126) for small reduced photon energies. Nevertheless, the cross section for emission of photons with reduced energy larger than a given cutoff value $W_{cr}$ is finite. The corresponding mean free path is

$$\lambda_{br}^{-1}(E; W_{cr}) \equiv \mathcal{N} \int_{W_{cr}}^{E} \frac{d\sigma_{br}}{dW} \, dW = \mathcal{N} \frac{Z^2}{\beta^2} \int_{\kappa_{cr}}^{1} \frac{1}{\kappa} \chi(Z, E, \kappa) \, d\kappa,$$

(3.136)

where $\kappa_{cr} = W_{cr} / E$. The radiative stopping power and the radiative energy straggling parameter, defined by

$$S_{br}(E) \equiv \mathcal{N} \int_{0}^{E} W \frac{d\sigma_{br}}{dW} \, dW = \mathcal{N} \frac{Z^2}{\beta^2} E \int_{0}^{1} \chi(Z, E, \kappa) \, d\kappa$$

(3.137)

and

$$\Omega_{br}^2(E) \equiv \mathcal{N} \int_{0}^{E} W^2 \frac{d\sigma_{br}}{dW} \, dW = \mathcal{N} \frac{Z^2}{\beta^2} E^2 \int_{0}^{1} \kappa \chi(Z, E, \kappa) \, d\kappa,$$

(3.138)

are both finite. For the kinetic energies $E_i$ of the grid, these quantities are easily calculated from the tabulated scaled DCS by using linear interpolation in $\kappa$. For positrons, the definitions (3.136)-(3.138) must be multiplied by the factor $F_{\mu}(Z, E)$ [eq. (3.134)].

Radiative stopping powers of aluminium, silver and gold for electrons and positrons are shown as functions of the kinetic energy in fig. 3.13. The stopping powers computed from the DCS given by eq. (3.126) practically coincide with ICRU37 (1984) values (also Berger and Seltzer, 1982). To leave room for future improvements, PENELOE reads the radiative stopping power for electrons from the input material data file, and renormalizes the DCS, eq. (3.126), (i.e. multiplies it by a $\kappa$-independent factor) so as to exactly reproduce the input radiative stopping power.

CSDA range

As mentioned above, the stopping power gives the average energy loss per unit path length. Thus, when an electron/positron with kinetic energy $E$ advances a small distance $ds$ within a medium, it loses an (average) energy $dE = -S(E) \, ds$, where

$$S(E) = S_{in}(E) + S_{br}(E) = -\frac{dE}{ds}$$

(3.139)

is the total (collisional+radiative) stopping power. Many electron transport calculations and old Monte Carlo simulations are based on the so-called continuous slowing down approximation (CSDA), which assumes that particles lose energy in a continuous way and at a rate equal to the stopping power. Evidently, the CSDA disregards energy-loss fluctuations and, therefore, it should be used with caution.
Chapter 3. Electron and positron interactions

Figure 3.13: Radiative stopping power $S_{br}/\rho$ for electrons and positrons in Al, Ag ($\times 10$) and Au ($\times 100$) as a function of the kinetic energy. Solid and dashed curves are results from the present model. Crosses are data from the ICRU37 report (1984) (also in Berger and Seltzer, 1982).

Figure 3.14: CSDA ranges for electrons and positrons in Al and Au as functions of the kinetic energy of the particle.
A parameter of practical importance is the so-called CSDA range (or Bethe range), which is defined as the path length travelled by a particle (in an infinite medium) before being absorbed and is given by

$$R(E) = \int_{E_{abs}}^{E} \frac{dE'}{S(E')}$$  \hspace{1cm} (3.140)

where we have considered that particles are effectively absorbed when they reach the energy $E_{abs}$. Notice that the CSDA range gives the average path length, actual (or Monte Carlo generated) path lengths fluctuate about the mean $R(E)$; the distribution of ranges has been studied by Lewis (1952). Fig. 3.14 displays CSDA ranges for electrons and positrons in aluminium and gold, this information is useful e.g. in estimating the maximum penetration depth of a beam and for range rejection (a variance reduction method). Compare fig. 3.14 with figs. 3.10 and 3.13 (right plots only) to get a feeling of how differences in stopping power between electrons and positrons are reflected on the CSDA ranges of these particles.

### 3.3.3 Angular distribution of emitted photons

The direction of the emitted bremsstrahlung photon is determined by the polar angle $\theta$ (see fig. 3.1) and the azimuthal angle $\phi$. For isotropic media, with randomly oriented atoms or molecules, the bremsstrahlung DCS is independent of $\phi$ and can be expressed as

$$\frac{d^2\sigma_{br}}{d\Omega d(\cos \theta)} = \frac{d\sigma_{br}}{dW} p(Z, E, \kappa; \cos \theta) = \frac{Z^2}{\beta^2 W} \chi(Z, E, \kappa) p(Z, E, \kappa; \cos \theta),$$ \hspace{1cm} (3.141)

where $p(Z, E, \kappa; \cos \theta)$ is the PDF of $\cos \theta$, $\theta$ is the polar angle of the photon direction relative to the direction of the projectile (fig. 3.1).

Numerical values of the “shape function” $p(Z, E, \kappa; \cos \theta)$, calculated by partial-wave methods, have been published by Kissel et al. (1983) for the following benchmark cases: $Z = 2, 8, 13, 47, 79, 92; E = 1, 5, 10, 50, 100, 500$ keV and $\kappa = 0, 0.6, 0.8, 0.95$. These authors also gave a parameterization of the shape function in terms of Legendre polynomials. Unfortunately, their analytical form is not suited for random sampling of the photon direction. In PENELOE we use a different parameterization that allows the random sampling of $\cos \theta$ in a simple way. Owing to the lack of numerical data for positrons, it is assumed that the shape function for positrons is the same as for electrons.

In previous simulation studies of x-ray emission from solids bombarded by electron beams (Acosta et al., 1998), the angular distribution of bremsstrahlung photons was described by means of the semiempirical analytical formulae derived by Kirkpatrick and Wiedmann (1945) [and subsequently modified by Statham (1976)]. These formulae were obtained by fitting the bremsstrahlung DCS derived from Sommerfeld’s theory. The shape function obtained from the Kirkpatrick-Wiedmann-Statham fit reads

$$p_{KWS}(Z, E, \kappa; \cos \theta) = \frac{\sigma_x(1 - \cos^2 \theta) + \sigma_y(1 + \cos^2 \theta)}{(1 - \beta \cos \theta)^2},$$ \hspace{1cm} (3.142)
where the quantities $\sigma_x$ and $\sigma_y$ are independent of $\theta$. Although this simple formula predicts the global trends of the partial-wave shape functions of Kissel et al. (1983) in certain energy and atomic number ranges, its accuracy is not sufficient for general-purpose simulations. In a preliminary analysis, we tried to improve this formula and determined the parameters $\sigma_x$ and $\sigma_y$ by direct fitting to the numerical partial-wave shape functions, but the improvement was not substantial. However, this analysis confirmed that the analytical form (3.142) is flexible enough to approximate the “true” (partial-wave) shape.

The analytical form (3.142) is plausible even for projectiles with relatively high energies, say $E$ larger than 1 MeV, for which the angular distribution of emitted photons is peaked at forward directions. This can be understood by means of the following classical argument (see e.g. Jackson, 1975). Assume that the incident electron is moving in the direction of the $z$-axis of a reference frame $K$ at rest with respect to the laboratory frame. Let $(\theta', \phi')$ denote the polar and azimuthal angles of the direction of the emitted photon in a reference frame $K'$ that moves with the electron and whose axes are parallel to those of $K$. In $K'$, we expect that the angular distribution of the emitted photons will not depart much from the isotropic distribution. To be more specific, we consider the following ansatz (modified dipole distribution) for the shape function in $K'$,

$$p_d(\cos \theta') = A \left[ \frac{3}{8} (1 + \cos^2 \theta') + (1 - A) \frac{3}{4} (1 - \cos^2 \theta') \right], \quad (0 \leq A \leq 1),$$  \hspace{1cm} (3.143)

which is motivated by the relative success of the Kirkpatrick-Wiedmann-Statham formula at low energies (note that the projectile is at rest in $K'$). The direction of emission $(\theta, \phi)$ in $K$ is obtained by means of the Lorentz transformation

$$\cos \theta = \frac{\cos \theta' + \beta}{1 + \beta \cos \theta'}, \quad \phi = \phi'.$$ \hspace{1cm} (3.144)

Thus, the angular distribution in $K$ reads

$$p(\cos \theta) = p_d(\cos \theta') \frac{d(\cos \theta')}{d(\cos \theta)} = A \left[ \frac{3}{8} \left[ 1 + \left( \frac{\cos \theta - \beta}{1 - \beta \cos \theta} \right)^2 \right] \frac{1 - \beta^2}{(1 - \beta \cos \theta)^2} \right.$$ \hspace{1cm} (3.145)

$$+ (1 - A) \left[ \frac{3}{4} \left[ 1 - \left( \frac{\cos \theta - \beta}{1 - \beta \cos \theta} \right)^2 \right] \frac{1 - \beta^2}{(1 - \beta \cos \theta)^2} \right].$$

Now, it is clear that when $\beta$ tends to unity, the shape function concentrates at forward directions.

We found that the benchmark partial-wave shape functions of Kissel et al. (1983) can be closely approximated by the analytical form (3.145) if one considers $A$ and $\beta$ as
Figure 3.15: Shape functions (angular distributions) for bremsstrahlung emission by electrons of the indicated energies in the fields of Al and Au atoms. Dashed curves are partial-wave shape functions of Kissel et al. (1983). Continuous curves are the present analytical fits, eq. (3.146). For visual aid, some curves have been shifted upwards in the amounts indicated in parentheses.
adjustable parameters. Explicitly, we write

\[ p_{\text{fit}}(\cos \theta) = A \frac{3}{8} \left[ 1 + \left( \frac{\cos \theta - \beta'}{1 - \beta' \cos \theta} \right)^2 \right] \frac{1 - \beta'^2}{(1 - \beta' \cos \theta)^2} \]

\[ + (1 - A) \frac{3}{4} \left[ 1 - \left( \frac{\cos \theta - \beta'}{1 - \beta' \cos \theta} \right)^2 \right] \frac{1 - \beta'^2}{(1 - \beta' \cos \theta)^2}, \]

with \( \beta' = \beta(1 + B) \). The parameters \( A \) and \( B \) have been determined, by least squares fitting, for the 144 combinations of atomic number, electron energy and reduced photon energy corresponding to the benchmark shape functions tabulated by Kissel et al. (1983). Results of this fit are compared with the original partial-wave shape functions in fig. 3.15. The largest differences between the fits and the data were found for the higher atomic numbers, but even then the fits are very accurate, as shown in fig. 3.15. The quantities \( \ln(AZ\beta) \) and \( B\beta \) vary smoothly with \( Z \), \( \beta \) and \( \kappa \) and can be obtained by cubic spline interpolation of their values for the benchmark cases. This permits the fast evaluation of the shape function for any combination of \( Z \), \( \beta \) and \( \kappa \). Moreover, the random sampling of the photon direction, i.e. of \( \cos \theta \), can be performed by means of a simple, fast analytical algorithm (see below).

### 3.3.4 Simulation of hard radiative events

Let us now consider the simulation of hard radiative events (\( W > W_{\text{cr}} \)) from the DCS defined by eqs. (3.141) and (3.146). PENEOPE reads the scaled bremsstrahlung DCS from the database files and, by natural cubic spline interpolation/extrapolation in \( \ln E \), produces a table for a denser logarithmic grid of 200 energies (and for the “standard” mesh of 32 \( \kappa \)’s), which is stored in memory. This energy grid spans the full energy range considered in the simulation and allows accurate (and fast) linear interpolation of the scaled DCS in the variable \( \ln E \), which is more adequate than \( E \) when interpolation over a wide energy interval is required.

Notice that in the Monte Carlo simulation the kinetic energy of the transported electron (or positron) varies in a random way and may take arbitrary values within a certain domain. Hence, we must be able to simulate bremsstrahlung emission by electrons with energies \( E \) not included in the grid.

**Sampling of the photon energy**

The PDF for the reduced photon energy, \( \kappa = W/E \), is given by [see eq. (3.126)]

\[ p(E, \kappa) = \frac{1}{\kappa} \chi(Z, E, \kappa) \Theta(\kappa - \kappa_{\text{cr}}), \]

where \( \kappa_{\text{cr}} = W_{\text{cr}}/E \) and \( \chi(Z, E, \kappa) \) is calculated by linear interpolation, in both \( \ln E \) and \( \kappa \), in the stored table. That is, \( \chi(Z, E, \kappa) \) is considered to be a piecewise linear function
of $\kappa$. To sample $\kappa$ from the PDF (3.147) for an energy $E_i$ in the grid, we express the interpolated scaled DCS as

$$\chi(Z, E_i; \kappa) = a_j + b_j \kappa \quad \text{if } \kappa_j \leq \kappa \leq \kappa_{j+1},$$

(3.148)

and introduce the cumulative distribution function,

$$\mathcal{P}_j = \int_{\kappa_{j-1}}^{\kappa_j} \, p(E_i, \kappa) \, d\kappa,$$

(3.149)

which, for a piecewise linear $\chi$, can be computed exactly. We also define

$$\chi_{\text{max},j} = \max \left\{ \chi(Z, E_i, \kappa), \kappa \in (\kappa_j, \kappa_{j+1}) \right\} \quad j = 1, \ldots, 32.$$

(3.150)

With all this we can formulate the following sampling algorithm, which combines a numerical inverse transform and a rejection,

(i) Generate a random number $\xi$ and determine the index $j$ for which $\mathcal{P}_j \leq \xi \mathcal{P}_{32} \leq \mathcal{P}_{j+1}$ using the binary search method.

(ii) Sample $\kappa$ from the distribution $\kappa^{-1}$ in the interval $(\kappa_j, \kappa_{j+1})$, i.e.

$$\kappa = \kappa_j \left( \frac{\kappa_{j+1}}{\kappa_j} \right)^{\xi}.$$

(3.151)

(iii) If $\xi \chi_{\text{max},j} < a_j + b_j \kappa$, deliver $\kappa$.

(iv) Go to step (i).

This sampling algorithm is exact and very fast [notice that the binary search in step (i) requires at most 5 comparisons], but is only applicable for the energies in the grid where $\chi$ is tabulated.

To simulate bremsstrahlung emission by electrons with energies $E$ not included in the grid, we should first obtain the PDF $p(E, \kappa)$ by interpolation along the energy axis and then perform the random sampling of $\kappa$ from this PDF using the algorithm described above. This procedure is too time consuming. A faster method consists of assuming that the grid of energies is dense enough so that linear interpolation in $\ln E$ is sufficiently accurate. If $E_i < E < E_{i+1}$, we can express the interpolated PDF as

$$p_{\text{int}}(E, \kappa) = \pi_i p(E_i, \kappa) + \pi_{i+1} p(E_{i+1}, \kappa)$$

(3.152)

with

$$\pi_i = \frac{\ln E_{i+1} - \ln E}{\ln E_{i+1} - \ln E_i}, \quad \pi_{i+1} = \frac{\ln E - \ln E_i}{\ln E_{i+1} - \ln E_i}.$$ 

(3.153)

These “interpolation weights” are positive and add to unity, i.e. they can be interpreted as point probabilities. Therefore, to perform the random sampling of $\kappa$ from $p_{\text{int}}(E, \kappa)$ we can employ the composition method (section 1.2.5), which leads to the following algorithm:
(i) Sample the integer variable $I$, which can take the values $i$ or $i+1$ with point
probabilities $\pi_i$ and $\pi_{i+1}$, respectively.

(ii) Sample $\kappa$ from the distribution $p_{\text{int}}(E_i; \kappa)$.

With this “interpolation by weight” method we only need to sample $\kappa$ from the tabulated
PDFs, i.e. for the energies $E_i$ of the grid.

Angular distribution of emitted photons

The random sampling of $\cos \theta$ is simplified by noting that the PDF given by eq. (3.146)
results from a Lorentz transformation, with speed $\beta'$, of the PDF (3.143). This means
that we can sample the photon direction $\cos \theta'$ in the reference frame $K'$ from the PDF
(3.143) and then apply the transformation (3.144) (with $\beta'$ instead of $\beta$) to get the
direction $\cos \theta$ in the laboratory frame.

To generate random values of $\cos \theta$ from (3.146) we use the following algorithm, which combines the composition and rejection methods,

(i) Sample a random number $\xi_1$.

(ii) If $\xi_1 < A$, then
1) Sample a random number $\xi$ and set $\cos \theta = -1 + 2\xi$.
2) Sample a random number $\xi$.
3) If $2\xi > 1 + \cos^2 \theta'$, go to 1).

(iii) If $\xi_1 \geq A$, then
4) Sample a random number $\xi$ and set $\cos \theta = -1 + 2\xi$.
5) Sample a random number $\xi$.
6) If $\xi > 1 - \cos^2 \theta'$, go to 4).

(iv) Deliver $\cos \theta = \frac{\cos \theta + \beta'}{1 + \beta' \cos \theta'}$.

The efficiencies of the rejections in steps (ii) and (iii) are both equal to 0.66. That is,
on average, we need 4 random numbers to generate each value of $\cos \theta$.

3.4 Positron annihilation

Following Nelson et al. (1985), we consider that positrons penetrating a medium of
atomic number $Z$ with kinetic energy $E$ can annihilate with the electrons in the medium
by emission of two photons. We assume that the target electrons are free and at rest,
thus disregarding electron binding effects, which enable one-photon annihilation (Heitler,
1954). When annihilation occurs in flight, i.e. when the kinetic energy $E$ of the positron
is larger than the “absorption” energy, the two photons may have different energies, say
$E_-$ and $E_+$, which add to $E + 2m_e c^2$. In what follows, quantities referring to the photon with the lowest energy will be denoted by the subscript “−”. Each annihilation event is then completely characterized by the quantity

$$\zeta \equiv \frac{E_-}{E + 2m_e c^2}. \quad (3.154)$$

Assuming that the positron moves initially in the direction of the z-axis, from conservation of energy and momentum it follows that the two photons are emitted in directions with polar angles [see eqs. (A.21) and (A.22) in appendix A]

$$\cos \theta_- = (\gamma^2 - 1)^{-1/2}(\gamma + 1 - 1/\zeta) \quad (3.155)$$

and

$$\cos \theta_+ = (\gamma^2 - 1)^{-1/2}[\gamma + 1 - 1/(1 - \zeta)], \quad (3.156)$$

and azimuthal angles $\phi_-$ and $\phi_+ = \phi_- + \pi$. The quantity $\gamma = 1 + E/(m_e c^2)$ is the total energy of the positron in units of its rest energy.

The maximum value of $\zeta$ is $1/2$, its minimum value is found when $\cos \theta_- = -1$ and is given by

$$\zeta_{\min} = \frac{1}{\gamma + 1 + (\gamma^2 - 1)^{1/2}}. \quad (3.157)$$

The DCS (per electron) for two-photon annihilation, as observed in the centre-of-mass system of the positron and the electron, is given by Heitler (1954). Nelson et al. (1985) transformed this DCS to the laboratory system (where the electron is at rest), their result can be written as

$$\frac{d\sigma_{\text{an}}}{d\zeta} = \frac{\pi r_e^2}{(\gamma + 1)(\gamma^2 - 1)} [S(\zeta) + S(1 - \zeta)], \quad (3.158)$$

where

$$S(\zeta) = -(\gamma + 1)^2 + (\gamma^2 + 4\gamma + 1)\frac{1}{\zeta} - \frac{1}{\zeta^2}. \quad (3.159)$$

Owing to the axial symmetry of the process, the DCS is independent of the azimuthal angle $\phi_-$, which is uniformly distributed on the interval $(0, 2\pi)$. For fast positrons, annihilation photons are emitted preferentially at forward directions. When the kinetic energy of the positron decreases, the angular distribution of the generated photons becomes more isotropical (see fig. 3.16).

The cross section (per target electron) for two-photon annihilation is

$$\sigma_{\text{an}} = \int_{\zeta_{\min}}^{1/2} \frac{d\sigma_{\text{an}}}{d\zeta} d\zeta = \frac{\pi r_e^2}{(\gamma + 1)(\gamma^2 - 1)} \times \left\{(\gamma^2 + 4\gamma + 1) \ln \left[\gamma + \left(\gamma^2 - 1\right)^{1/2}\right] - (3 + \gamma) \left(\gamma^2 - 1\right)^{1/2}\right\}. \quad (3.160)$$
The annihilation mean free path is given by
\[ \lambda_{\text{an}}^{-1} = \mathcal{N} \mathcal{Z} \sigma_{\text{an}}, \]  
(3.161)
where \( \mathcal{N} \mathcal{Z} \) is the density of electrons in the medium. The annihilation cross section is displayed in fig. 3.16. The cross section decreases with the kinetic energy and, therefore, high-energy positrons can travel path lengths of the order of the CSDA range before annihilating.

### 3.4.1 Generation of emitted photons

The PDF of \( \zeta \) is given by (normalization is irrelevant here)
\[ p_{\text{an}}(\zeta) = S(\zeta) + S(1 - \zeta), \quad \zeta_{\text{min}} \leq \zeta \leq 1/2. \]  
(3.162)
To sample \( \zeta \), we may take advantage of the symmetry of this expression under the exchange of the two photons, which corresponds to exchanging \( \zeta \) and \( 1 - \zeta \). We first consider the distribution
\[ P(v) \equiv S(v), \quad \zeta_{\text{min}} \leq v \leq 1 - \zeta_{\text{min}} \]  
(3.163)
and write it in the form

\[ P(v) = \pi(v) g(v) \]  \hspace{1cm} (3.164)

with

\[ \pi(v) = \left[ \ln \left( \frac{1 - \zeta_{\text{min}}}{\zeta_{\text{min}}} \right) \right]^{-1} \frac{1}{v} \]  \hspace{1cm} (3.165)

and

\[ g(v) = \left[ -\left( \gamma + 1 \right)^2 v + \left( \gamma^2 + 4 \gamma + 1 \right) - \frac{1}{v} \right] . \]  \hspace{1cm} (3.166)

\( \pi(v) \) is a proper PDF (i.e. it is definite positive and normalized to unity) and \( g(v) \) is a monotonically decreasing function. Random values of \( v \) from the distribution \( P(v) \) can be generated by using the following algorithm (rejection method):

(i) Sample a value \( v \) from the distribution \( \pi(v) \). This is easily done with the inverse transform method, which yields the following sampling equation

\[ v = \zeta_{\text{min}} \left( \frac{1 - \zeta_{\text{min}}}{\zeta_{\text{min}}} \right) ^{\xi} . \]  \hspace{1cm} (3.167)

(ii) Generate a new random number \( \xi \).

(iii) If \( \xi g(\zeta_{\text{min}}) > g(v) \), go to step (i).

(iv) Deliver \( v \).

It is clear that the random value

\[ \zeta = \min(v, 1 - v) \]  \hspace{1cm} (3.168)

follows the distribution given by eq. (3.162) when \( v \) is sampled from the distribution \( P(v) \). The efficiency of this sampling algorithm practically equals 100% for positrons with kinetic energy \( E \) less than 10 keV, decreases when \( E \) increases to reach a minimum value of \( \sim 80\% \) at \( E \sim 10 \text{ MeV} \) and increases monotonically for larger energies.

As the result of annihilation, two photons with energies \( E_- = \zeta(E + 2m_e c^2) \) and \( E_+ = (1 - \zeta)(E + 2m_e c^2) \) are emitted in the directions given by eqs. (3.155) and (3.156).
Chapter 4

Electron/positron transport mechanics

In principle, the scattering model and sampling techniques described in chapter 3 allows the detailed Monte Carlo simulation of electron and positron transport in matter. However, detailed simulation is feasible only when the mean number of interactions per track is small (a few hundred at most). This occurs for electrons with low initial kinetic energies or for thin geometries. The number of interactions experienced by an electron or positron before being effectively stopped increases with its initial energy and, therefore, detailed simulation becomes impractical at high energies.

PENELLOPE implements a “mixed” simulation scheme (Berger, 1963; Reimer and Krefting, 1976; Andreo and Brahme, 1984), which combines the detailed simulation of hard events (i.e. events with polar angular deflection $\theta$ or energy loss $W$ larger than previously selected cutoff values $\theta_c$ and $W_c$) with condensed simulation of soft events, in which $\theta < \theta_c$ or $W < W_c$. Owing to the fact that for high-energy electrons the DCSs for the various interaction processes decrease rapidly with the polar scattering angle and the energy loss, cutoff values can be selected such that the mean number of hard events per electron track is sufficiently small to permit their detailed simulation. In general, this is accomplished by using relatively small cutoff values, so that each soft interaction has only a slight effect on the simulated track. The global effect of the (usually many) soft interactions that take place between each pair of consecutive hard events can then be simulated accurately by using a multiple scattering approach. Hard events occur much less frequently than soft events, but they have severe effects on the track evolution (i.e. they cause large angular deflections and lateral displacements or considerable energy losses), which can only be properly reproduced by detailed simulation. The computer time needed to simulate each track diminishes rapidly when the cutoff values for the angular deflection and the energy loss are increased. Mixed simulation algorithms are usually very stable under variations of the adopted cutoff values, whenever these are kept below some reasonable limits. Mixed simulation is then preferable to condensed simulation because 1) spatial distributions are simulated more accurately, 2) tracks in the vicinity of interfaces are properly handled, and 3) possible dependence of the results on user-defined parameters is largely reduced.
4.1 Elastic scattering

Let us start by considering electrons (or positrons) with kinetic energy \( E \) moving in a hypothetical infinite homogeneous medium, with \( N \) scattering centres per unit volume, in which they experience only pure elastic collisions (i.e. with no energy loss).

4.1.1 Multiple elastic scattering theory

Assume that an electron starts off from a certain position, which we select as the origin of our reference frame, moving in the direction of the \( z \)-axis. Let \( f(s; \mathbf{r}, \hat{d}) \) denote the probability density of finding the electron at the position \( \mathbf{r} = (x, y, z) \), moving in the direction given by the unit vector \( \hat{d} \) after having travelled a path length \( s \). The diffusion equation for this problem is (Lewis, 1950)

\[
\frac{\partial f}{\partial s} + \hat{d} \cdot \nabla f = N \int [f(s; \mathbf{r}, \hat{d}') - f(s; \mathbf{r}, \hat{d})] \frac{d\sigma_{el}(\theta)}{d\Omega} \ d\Omega, \tag{4.1}
\]

where \( \theta = \arccos(\hat{d} \cdot \hat{d}') \) is the scattering angle corresponding to the angular deflection \( \hat{d}' \rightarrow \hat{d} \). This equation has to be solved with the boundary condition \( f(0; \mathbf{r}, \hat{d}) = (1/\pi)\delta(\mathbf{r})\delta(1 - \cos \chi) \), where \( \chi \) is the polar angle of the direction \( \hat{d} \). By expanding \( f(s; \mathbf{r}, \hat{d}) \) in spherical harmonics, Lewis (1950) obtained exact expressions for the angular distribution and for the first moments of the spatial distribution after a given path length \( s \). The probability density \( F(s; \chi) \) of having a final direction in the solid angle element \( d\Omega \) around a direction defined by the polar angle \( \chi \) is given by

\[
F(s; \chi) = \int f(s; \mathbf{r}, \hat{d}) \ d\mathbf{r} = \sum_{\ell=0}^{\infty} \frac{2\ell + 1}{4\pi} \exp(-s/\lambda_{el,\ell}) P_{\ell}(\cos \chi), \tag{4.2}
\]

where \( P_{\ell}(\cos \chi) \) are Legendre polynomials and \( \lambda_{el,\ell} = 1/(N\sigma_{el,\ell}) \) is the \( \ell \)-th transport mean free path defined by eq. (3.14). The result given by eq. (4.2) coincides with the multiple scattering distribution obtained by Goudsmit and Saunderson (1940a, 1940b). Evidently, the distribution \( F(s; \chi) \) is symmetric about the \( z \)-axis, i.e. independent of the azimuthal angle of the final direction.

From the orthogonality of the Legendre polynomials, it follows that

\[
\langle P_{\ell}(\cos \chi) \rangle = 2\pi \int_{-1}^{1} P_{\ell}(\cos \chi) F(s; \chi) \ d(\cos \chi) = \exp(-s/\lambda_{el,\ell}). \tag{4.3}
\]

In particular, we have

\[
\langle \cos \chi \rangle = \exp(-s/\lambda_{el,1}) \tag{4.4}
\]

and

\[
\langle \cos^2 \chi \rangle = \frac{1}{3} [1 + 2 \exp(-s/\lambda_{el,2})]. \tag{4.5}
\]
4.1. Elastic scattering

Lewis (1950) also derived analytical formulae for the first moments of the spatial distribution and the correlation function of $z$ and $\cos \chi$. Neglecting energy losses, the results explicitly given in Lewis’ paper simplify to

$$
\langle z \rangle \equiv 2\pi \int z f(s; \mathbf{r}, \mathbf{d}) \, d(cos \chi) \, d\mathbf{r} = \lambda_{el,1} \left[ 1 - \exp(-s/\lambda_{el,1}) \right],
$$

(4.6)

$$
\langle x^2 + y^2 \rangle \equiv 2\pi \int \left( x^2 + y^2 \right) f(s; \mathbf{r}, \mathbf{d}) \, d(cos \chi) \, d\mathbf{r} = \frac{4}{3} \int_0^s dt \, \exp(-t/\lambda_{el,1}) \int_0^t \left[ 1 - \exp(-u/\lambda_{el,1}) \right] \exp(u/\lambda_{d,1}) \, du,
$$

(4.7)

$$
\langle z \cos \chi \rangle \equiv 2\pi \int z \cos \chi f(s; \mathbf{r}, \mathbf{d}) \, d(cos \chi) \, d\mathbf{r} = \exp(-s/\lambda_{el,1}) \int_0^s \left[ 1 + 2 \exp(-t/\lambda_{d,1}) \right] \exp(t/\lambda_{d,1}) \, dt.
$$

(4.8)

It is worth observing that the quantities (4.4)–(4.8) are completely determined by the values of the transport mean free paths $\lambda_{el,1}$ and $\lambda_{el,2}$; they are independent of the elastic mean free path $\lambda_{el}$.

4.1.2 Mixed simulation of elastic scattering

At high energies, where detailed simulation becomes impractical, $\lambda_{el,1} \gg \lambda_{el}$ (see fig. 3.3) so that the average angular deflection in each collision is small. In other words, the great majority of elastic collisions of fast electrons are soft collisions with very small deflections. We shall consider mixed simulation procedures (see Fernández-Varela et al., 1993b; Baró et al., 1994b) in which hard collisions, with scattering angle $\theta$ larger than a certain value $\theta_c$, are individually simulated and soft collisions (with $\theta < \theta_c$) are described by means of a multiple scattering approach.

In practice, the mixed algorithm will be defined by specifying the mean free path $\lambda_{el}^{(h)}$ between hard elastic events, defined by [see eqs. (3.10) and (3.12)]

$$
\frac{1}{\lambda_{el}^{(h)}} = N \frac{2\pi}{\int \sigma_{el}(\theta) \, d\Omega \sin \theta \, d\theta}.
$$

(4.9)

This equation determines the cutoff angle $\theta_c$ as a function of $\lambda_{el}^{(h)}$. A convenient recipe to set the mean free path $\lambda_{el}^{(h)}$ is

$$
\lambda_{el}^{(h)}(E) = \max \{ \lambda_{el}(E), C_1 \lambda_{el,1}(E) \},
$$

(4.10)

where $C_1$ is a pre-selected small constant (say, less than $\sim 0.1$). For increasing energies, $\lambda_{el}$ attains a constant value and $\lambda_{el,1}$ increases steadily (see fig. 3.3) so that the formula (4.10) gives a mean free path for hard collisions that increases with energy, i.e. hard collisions are less frequent when the scattering effect is weaker. The recipe (4.10) also ensures that $\lambda_{el}^{(h)}$ will reduce to the actual mean free path $\lambda_{el}$ for low energies. In this case,
soft collisions cease to occur ($\theta_c = 0$) and mixed simulation becomes purely detailed. It is worth noticing that, when mixed simulation is effective (i.e. when $\lambda_{el}^{(h)} > \lambda_{el}$), the mean angular deflection in a path length $\lambda_{el}^{(h)}$ is [see eq. (4.4)]

$$1 - \langle \cos \chi \rangle = 1 - \exp(-\lambda_{el}^{(h)}/\lambda_{el,1}) \simeq C_1.$$  \hspace{1cm} (4.11)

Hence, when using the prescription (4.10), the average angular deflection due to all elastic collisions occurring along a path length $\lambda_{el}^{(h)}$ equals $C_1$.

The PDF of the step length $s$ between two successive hard collisions is

$$p(s) = \frac{1}{\lambda_{el}^{(h)}} \exp(-s/\lambda_{el}^{(h)}),$$  \hspace{1cm} (4.12)

and random values of $s$ can be generated by means of the sampling formula, eq. (1.36)

$$s = -\lambda_{el}^{(h)} \ln \xi.$$  \hspace{1cm} (4.13)

The (unnormalized) PDF of the polar deflection $\theta$ in single hard collisions is

$$p^{(h)}(\theta) = \frac{d\sigma_{el}(\theta)}{d\Omega} \sin \theta \Theta(\theta - \theta_c),$$  \hspace{1cm} (4.14)

where $\Theta(x)$ stands for the step function.

The inverse transport mean free paths $\lambda_{el,\ell}^{-1}$, see eq. (3.14), for the actual scattering process can be split into contributions from soft and hard collisions, i.e.

$$\frac{1}{\lambda_{el,\ell}} = \frac{1}{\lambda_{el,\ell}^{(s)}} + \frac{1}{\lambda_{el,\ell}^{(h)}},$$  \hspace{1cm} (4.15)

where

$$\frac{1}{\lambda_{el,\ell}^{(s)}} = \mathcal{N}2\pi \int_0^{\theta_c} [1 - P_\ell(\cos \theta)] \frac{d\sigma_{el}(\theta)}{d\Omega} \sin \theta d\theta$$  \hspace{1cm} (4.16a)

and

$$\frac{1}{\lambda_{el,\ell}^{(h)}} = \mathcal{N}2\pi \int_\theta^{\pi} [1 - P_\ell(\cos \theta)] \frac{d\sigma_{el}(\theta)}{d\Omega} \sin \theta d\theta.$$  \hspace{1cm} (4.16b)

Let us assume that an electron starts off from the origin of coordinates moving in the direction of the $z$-axis and undergoes the first hard collision after travelling a path length $s$. The exact angular distribution produced by the soft collisions along this step is

$$F^{(s)}(s; \chi) = \sum_{\ell=0}^{\infty} \frac{2\ell + 1}{4\pi} \exp(-s/\lambda_{el,\ell}^{(s)}) P_\ell(\cos \chi).$$  \hspace{1cm} (4.17)

The exact average longitudinal and transverse displacements at the end of the step are given by [see eqs. (4.6) and (4.7)]

$$\langle z \rangle^{(s)} = \lambda_{el,1}^{(s)} \left[1 - \exp(-s/\lambda_{el,1}^{(s)})\right] = s \left[1 - \frac{1}{2} \left(\frac{s}{\lambda_{el,1}^{(s)}}\right) + \frac{1}{6} \left(\frac{s}{\lambda_{el,1}^{(s)}}\right)^2 - \ldots\right],$$  \hspace{1cm} (4.18)
\[ \langle x^2 + y^2 \rangle_{(s)} = \frac{2}{9} \frac{s^3}{\lambda_{el,1}^{(s)}} \left[ 1 - \frac{1}{4} \left( 1 + \frac{\lambda_{el,1}^{(s)}}{\lambda_{el,2}^{(s)}} \right) \left( \frac{s}{\lambda_{el,1}^{(s)}} \right) \right] + \cdots \], \quad (4.19)

where \( \lambda_{el,1}^{(s)} \), the first transport mean free path for soft collisions, is larger than \( \lambda_{el,1} \). As the mean free path between hard collisions is normally much less than \( \lambda_{el,1}^{(s)} \) (depending on the value of \( C_1 \)), the value \( s/\lambda_{el,1}^{(s)} \) is, on average, much less than unity (note that \( \langle s \rangle = \lambda_{el,1}^{(h)} \)). Therefore, the global effect of the soft collisions in the step, i.e. the change in direction of movement and the lateral displacement, is very small (part of the deflection is caused by the hard interaction at the end of the step).

In PENELOPE, the angular deflection and the lateral displacement due to the multiple soft collisions in a step of length \( s \) are simulated by means of the random hinge method\(^1\) (Fernández-Varea et al., 1993b). The associated algorithm can be formulated as follows (see fig. 4.1),

(i) The electron first moves a random distance \( \tau \), which is sampled uniformly in the interval \((0, s)\), in the initial direction.

(ii) Then a single artificial soft scattering event (a hinge) takes place, in which the electron changes its direction of movement according to the multiple scattering distribution \( F^{(s)}(s; \chi) \).

(iii) Finally, the electron moves a distance \( s - \tau \) in the new direction.

![Figure 4.1: Simulation of the global effect of soft collisions between two consecutive hard collisions by the random hinge method.](image)

Obviously, this algorithm leads to the exact angular distribution at the end of the step. The average longitudinal displacement at the end of the simulated step is

\[ \langle z \rangle_{\text{sim}} = \frac{s}{2} + \frac{s}{2} \langle \cos \chi \rangle_{\text{sim}} = s \left[ 1 - \frac{1}{2} \left( \frac{s}{\lambda_{el,1}^{(s)}} \right) + \frac{1}{4} \left( \frac{s}{\lambda_{el,1}^{(s)}} \right)^2 - \cdots \right], \quad (4.20) \]

which agrees closely with the exact result given by eq. (4.18). Moreover, the average simulated transverse displacement is

\[ \langle x^2 + y^2 \rangle_{\text{sim}} = \langle (s - \tau)^2 \sin^2 \chi \rangle_{\text{sim}} = \frac{1}{3} s^2 \left( 1 - \langle \cos^2 \chi \rangle_{\text{sim}} \right) \]

\(^1\)The name was coined by Ron Kensek.
which does not differ much from the exact value given by eq. (4.19). From these facts, we may conclude that the random hinge method provides a faithful description of the transport when the step length \( s \) is much shorter than the first transport mean free path \( \lambda_{d,1} \), so that the global angular deflection and lateral displacement are small. Surprisingly, it does work well also in condensed (class I) simulations, where this requirement is not met. In spite of its simplicity, the random hinge method competes in accuracy and speed with other, much more sophisticated transport algorithms (see Bielajew and Salvat, 2001, and references therein). It seems that the randomness of the hinge position \( r \) leads to correlations between the angular deflection and the displacement that are close to the actual correlations.

The random hinge algorithm can be readily adapted to simulate multiple elastic scattering processes in limited material structures, which may consist of several regions of different compositions separated by well-defined surfaces (interfaces). In these geometries, when the track crosses an interface, we simply stop it at the crossing point, and resume the simulation in the new material. In spite of its simplicity, this recipe gives a fairly accurate description of interface crossing. To see this, consider that a hard collision has occurred at the position \( r \) in region “1” and assume that the following hard collision occurs in region “2”. The step length \( s \) between these two hard collisions is larger than the distance \( t \) from \( r \) to the interface (see fig. 4.2). If the artificial soft elastic collision occurs in region “1”, the angular deflection in this collision is sampled from the distribution \( F^{(s)}(s; \chi) \). Otherwise, the electron reaches the interface without changing its direction of movement. Assuming \( s \ll \lambda_{d,1}^{(s)} \), the mean angular deflection in a soft collision is

\[
1 - \langle \cos \chi \rangle^{(s)} = 1 - \exp(-s/\lambda_{d,1}^{(s)}) \simeq \frac{s}{\lambda_{d,1}^{(s)}}. \tag{4.22}
\]

Moreover, when this assumption is valid, lateral displacements due to soft collisions are small and can be neglected to a first approximation. As the probability for the soft collision to occur within region “1” equals \( t/s \), the average angular deflection of the

![Figure 4.2: Simulation of a track near the crossing of an interface.](image)
simulated electron track when it reaches the interface is

\[ 1 - \langle \cos \chi \rangle = \frac{t}{s} \left( 1 - \langle \cos \chi \rangle^{(s)} \right) \simeq \frac{t}{\lambda^{(s)}_{d,1}}, \]  

(4.23)

which practically coincides with the exact mean deviation after the path length \( t \) within region “1”, as required. Thus, by sampling the position of the soft collision uniformly in the segment \((0, s)\) we make sure that the electron reaches the interface with the correct average direction of movement.

### Angular deflections in soft scattering events

In the random hinge method, the global effect of the soft collisions experienced by the particle along a path segment of length \( s \) between two consecutive hard events is simulated as a single artificial soft scattering event. The angular deflection follows the multiple scattering distribution \( F^{(s)}(s; \chi) \). Unfortunately, the exact Legendre expansion, eq. (4.17), is not appropriate for Monte Carlo simulation, since this expansion converges very slowly (because the associated single scattering DCS is not continuous) and the sum varies rapidly with the path length \( s \).

Whenever the cutoff angle \( \theta_c \) is small, the distribution \( F^{(s)}(s; \chi) \) may be calculated by using the small angle approximation (see e.g. Lewis, 1950). Notice that \( \theta_c \) can be made as small as desired by selecting a small enough value of \( C_1 \), see eqs. (4.9) and (4.10). Introducing the limiting form of the Legendre polynomials

\[ P_\ell(\cos \theta) \simeq 1 - \frac{1}{4} \ell (\ell + 1) \theta^2 \]  

(4.24)

into eq. (4.16a) we get

\[ \frac{1}{\lambda^{(s)}_{d,\ell}} = \mathcal{N} \frac{2 \pi}{4} \frac{\ell (\ell + 1)}{4} \int_0^{\theta_c} \theta^2 \frac{d\sigma_{\ell \ell}(\theta)}{d\Omega} \sin \theta \, d\theta = \frac{\ell (\ell + 1)}{2} \frac{1}{\lambda^{(s)}_{d,1}}, \]  

(4.25)

i.e. the transport mean free paths \( \lambda^{(s)}_{d,\ell} \) are completely determined by the single value \( \lambda^{(s)}_{d,1} \). The angular distribution \( F^{(s)} \) then simplifies to

\[ F^{(s)}(s; \chi) = \sum_{\ell=0}^{\infty} \frac{2\ell + 1}{4\pi} \exp \left[ -\frac{\ell (\ell + 1)}{2} \frac{s}{\lambda^{(s)}_{d,1}} \right] P_\ell(\cos \chi). \]  

(4.26)

This expression can be evaluated by using the Molière (1948) approximation for the Legendre polynomials, we obtain (see Fernández-Varea et al., 1993b)

\[ F^{(s)}(s; \chi) = \frac{1}{2\pi} \left( \frac{\chi}{\sin \chi} \right)^{1/2} \lambda^{(s)}_{d,1} \frac{s}{\sin \chi} \exp \left[ -\frac{s}{8\lambda^{(s)}_{d,1} \chi^2} \right]. \]  

(4.27)

which does not differ significantly from the Gaussian distribution with variance \( s/\lambda^{(s)}_{d,1} \).

This result is accurate whenever \( s \ll \lambda^{(s)}_{d,1} \) and \( \theta_c \ll 1 \). It offers a possible method
of generating the angular deflection in artificial soft events. When the result given by
eq. (4.27) is applicable, the single parameter $\lambda_{\text{el},1}^{(s)}$ completely determines the multiple
scattering distribution due to soft collisions, i.e. other details of the DCS for scattering
angles less than $\theta_r$ are irrelevant. However, in actual Monte Carlo simulations, the
small-angle approximation is seldom applicable.

In most practical cases the number of hard collisions per electron track can be made
relatively large by simply using a small value of the parameter $C_1$ [see eq. (4.10)]]. When
the number of steps is large enough, say larger than $\sim 10$, it is not necessary to use the
exact distribution $F^{(s)}(s; \chi)$ to sample the angular deflection in artificial soft collisions.
Instead, we may use a simpler distribution, $F_n(s; \chi)$, with the same mean and variance,
without appreciably distorting the simulation results. This is so because the details of
the adopted distribution are washed out after a sufficiently large number of steps and
will not be seen in the simulated distributions. Notice that, within the small angle
approximation, it is necessary to keep only the proper value of the first moment to
get the correct final distributions. However, if the cutoff angle $\theta_c$ is not small enough,
the angular distribution $F^{(s)}(s; \chi)$ may become sensitive to higher-order moments of
the soft single scattering distribution. Thus, by also keeping the proper value of the
variance, the range of validity of the simulation algorithm is extended, i.e. we can speed
up the simulation by using larger values of $C_1$ (or of $\lambda_{\text{el}}^{(h)}$) and still obtain the correct
distributions.

We now return to the notation of section 3.1, and use the variable $\mu \equiv (1 - \cos \chi)/2$
to describe angular deflections in soft scattering events. The exact first and second
moments of the multiple scattering distribution $F^{(s)}(s; \mu)$ are

$$\langle \mu \rangle^{(s)} = \int_0^1 \mu F_n(s; \mu) \, d\mu = \frac{1}{2} \left[ 1 - \exp \left( -s / \lambda_{\text{el},1}^{(s)} \right) \right]$$  \hspace{1cm} (4.28)

and

$$\langle \mu^2 \rangle^{(s)} = \int_0^1 \mu^2 F_n(s; \mu) \, d\mu = \langle \mu \rangle^{(s)} - \frac{1}{6} \left[ 1 - \exp \left( -s / \lambda_{\text{el},2}^{(s)} \right) \right].$$  \hspace{1cm} (4.29)

The angular deflection in soft scattering events will be generated from a distribution
$F_n(s; \mu)$, which is required to satisfy eqs. (4.28) and (4.29), but is otherwise arbitrary.

PENELOPE uses the following,

$$F_n(s; \mu) = a U_{0,1}(\mu) + (1 - a) U_{1,1}(\mu),$$  \hspace{1cm} (4.30)

where $U_{u,v}(x)$ denotes the normalized uniform distribution in the interval $(u, v),

$$U_{u,v}(x) = \begin{cases} 
1/(v - u) & \text{if } u \leq x \leq v, \\
0 & \text{otherwise.}
\end{cases}$$  \hspace{1cm} (4.31)

The parameters $a$ and $b$, obtained from the conditions (4.28) and (4.29), are

$$b = \frac{2\langle \mu \rangle^{(s)} - 3\langle \mu^2 \rangle^{(s)}}{1 - 2\langle \mu \rangle^{(s)}} , \quad a = 1 - 2\langle \mu \rangle^{(s)} + b.$$  \hspace{1cm} (4.32)
4.1. Elastic scattering

The simple distribution (4.30) is flexible enough to reproduce the combinations of first and second moments encountered in the simulations [notice that \( \langle \mu \rangle^{(s)} \), eq. (4.28), is always less than 1/2] and allows fast random sampling of \( \mu \).

4.1.3 Simulating with the MW model

PENELOPE simulates elastic scattering by using the MW model (see section 3.1), which allows the formulation of the mixed simulation algorithm in closed analytical form.

The mean free path \( \lambda_{el}^{(h)} \) between hard elastic events and the cutoff deflection \( \mu_c = (1 - \cos \theta_c)/2 \) are related through [see eqs. (3.18) and (4.9)]

\[
\frac{1}{\lambda_{el}^{(h)}} = \frac{1}{\lambda_{el}} \int_{\mu_c}^1 p_{MW}(\mu) \, d\mu.
\]  
(4.33)

This equation can be easily inverted to give

\[
\mu_c = \mathcal{P}_{MW}^{-1}(\xi_c),
\]  
(4.34)

where

\[
\xi_c \equiv 1 - \frac{\lambda_{el}}{\lambda_{el}^{(h)}}
\]  
(4.35)

and \( \mathcal{P}_{MW}^{-1} \) is the inverse of the single scattering cumulative distribution function given by eqs. (3.31) and (3.36).

In the following, we assume that the MW distribution is that of case I, eq. (3.24); the formulae for case II can be derived in a similar way. The random sampling of the angular deflection \( \mu \) in hard collisions is performed by the inverse transform method (section 1.2.2); random values of \( \mu \) are obtained from the sampling equation

\[
\int_{\mu_c}^\mu p_{MW}(\mu') \, d\mu' = \xi \int_{\mu_c}^1 p_{MW}(\mu') \, d\mu'.
\]  
(4.36)

With the MW distribution, eq. (3.24), this equation can be solved analytically to give

\[
\mu = \mathcal{P}_{MW}^{-1} \left(1 - \frac{\lambda_{el}}{\lambda_{el}^{(h)}}(1 - \xi)\right).
\]  
(4.37)

To determine the angular distribution of soft events \( F_s(s; \mu) \), eq. (4.30), we need the first and second transport mean free paths for soft collisions, which are given by

\[
(\lambda_{el,1}^{(s)})^{-1} = \frac{2}{\lambda_{el}} T_1(\mu_c) \quad \text{and} \quad (\lambda_{el,2}^{(s)})^{-1} = \frac{6}{\lambda_{el}} [T_1(\mu_c) - T_2(\mu_c)]
\]  
(4.38)

with

\[
T_1(\mu_c) = \int_0^{\mu_c} \mu p_{MW}(\mu) \, d\mu \quad \text{and} \quad T_2(\mu_c) = \int_0^{\mu_c} \mu^2 p_{MW}(\mu) \, d\mu.
\]  
(4.39)
These latter quantities can be computed analytically as

\[
T_1(\mu_c) = \begin{cases} 
(1 - B)I_1(\mu_c) & \text{if } 0 \leq \xi_c < \xi_0 \\
(1 - B)I_1(\mu_0) + (\xi_c - \xi_0)\mu_0 & \text{if } \xi_0 \leq \xi_c < \xi_0 + B \\
(1 - B)I_1(\mu_c) + B\mu_0 & \text{if } \xi_0 + B \leq \xi_c \leq 1
\end{cases} \tag{4.40}
\]

and

\[
T_2(\mu_c) = \begin{cases} 
(1 - B)I_2(\mu_c) & \text{if } 0 \leq \xi_c < \xi_0 \\
(1 - B)I_2(\mu_0) + (\xi_c - \xi_0)\mu_0^2 & \text{if } \xi_0 \leq \xi_c < \xi_0 + B \\
(1 - B)I_2(\mu_c) + B\mu_0^2 & \text{if } \xi_0 + B \leq \xi_c \leq 1
\end{cases} \tag{4.41}
\]

with

\[
I_1(\mu) \equiv A \left[ (1 + A) \ln \left( \frac{A + \mu}{A} \right) - \frac{(1 + A)\mu}{A + \mu} \right] \tag{4.42}
\]

and

\[
I_2(\mu) \equiv A \left[ \frac{(1 + A)\mu^2}{A + \mu} - 2I_1(\mu) \right]. \tag{4.43}
\]

The quantities \(\xi_0\) and \(\xi_c\) are defined by eqs. (3.34) and (4.35), respectively.

### 4.2 Soft energy losses

The high-energy codes currently available implement different approximate methods to simulate inelastic collisions. Thus, ETRAN and ITS3 make use of the multiple scattering theories of Landau (1944) and Bhanck and Leisegang (1950) to obtain the energy loss distribution due to inelastic collisions after a given path length: the production of secondary electrons is simulated by means of the Møller (1932) and Bhabha (1936) DCSs, which neglect binding effects. This approach accounts for the whole energy straggling, within the accuracy of the multiple scattering theory, but disregards the correlation between delta ray emission and energy loss in each track segment. As a consequence, energetic delta rays can be generated in a track segment where the energy lost by the primary particle is smaller than the energy of the emitted delta rays. EGS4 uses a mixed procedure to simulate collision energy losses: hard inelastic collisions are simulated from the Møller and Bhabha DCSs, thus neglecting binding effects, and soft inelastic collisions are described by means of the continuous slowing down approximation (CSDA), i.e. energy straggling due to soft inelastic collisions is ignored. As regards bremsstrahlung emission, EGS4 implements a mixed procedure in which hard radiative events are simulated in detail and use is made of the CSDA to simulate the effect of soft photon emission; ETRAN uses strictly detailed simulation.

To make the arguments more precise, we introduce the cutoff values \(W_{cc}\) and \(W_{cr}\), and consider inelastic collisions with energy loss \(W < W_{cc}\) and emission of bremsstrahlung photons with \(W < W_{cr}\) as soft stopping interactions. The use of the CSDA to describe
4.2. Soft energy losses

soft interactions is well justified when the energy straggling due to these interactions
is negligible, as happens when the cutoff energies $W_{cc}$ and $W_{cr}$ are both small, so that
the fraction of the stopping power due to soft interactions is also small. To improve
the description of energy straggling one should reduce the cutoff energies, but this enlarges
the number of hard inelastic and radiative events to be simulated along each track and
hence the simulation time. Our purpose is to go beyond the CSDA by introducing energy
straggling in the description of soft stopping interactions. It is clear that, by proceeding
in this way, we will be able to use larger values of the cutoff energies $W_{cc}$ and $W_{cr}$, and
hence speed up the simulation, without distorting the energy distributions.

In previous versions of PENELope, soft energy losses were simulated by using the
mixed simulation algorithm described by Baró et al. (1995). The quantities that define
the algorithm are the mean free paths $\lambda_{\text{in}}^{(h)}$ and $\lambda_{\text{br}}^{(h)}$ between hard collisions and hard ra-
diative events, the stopping power $S_s$ and the energy straggling parameter $\Omega_s^2$ associated
with soft interactions. These quantities are given by

$$\lambda_{\text{in}}^{(h)} (E) = \left( \mathcal{N} \int_{W_{cc}}^{E} \frac{d\sigma_{\text{in}}}{dW} \, dW \right)^{-1}, \quad (4.44)$$

$$\lambda_{\text{br}}^{(h)} (E) = \left( \mathcal{N} \int_{W_{cr}}^{E} \frac{d\sigma_{\text{br}}}{dW} \, dW \right)^{-1}, \quad (4.45)$$

$$S_s (E) = \mathcal{N} \int_{0}^{W_{cc}} W \frac{d\sigma_{\text{in}}}{dW} \, dW + \mathcal{N} \int_{0}^{W_{cr}} W \frac{d\sigma_{\text{br}}}{dW} \, dW \quad (4.46)$$

and

$$\Omega_s^2 (E) = \mathcal{N} \int_{0}^{W_{cr}} W^2 \frac{d\sigma_{\text{in}}}{dW} \, dW + \mathcal{N} \int_{0}^{W_{cr}} W^2 \frac{d\sigma_{\text{br}}}{dW} \, dW \quad (4.47)$$

To prevent $\lambda_{\text{br}}^{(h)} (E)$ from vanishing (infrared divergence), in PENELope the radiative
cutoff energy $W_{cr}$ is required to be larger than or equal to 10 eV.

Let us consider that a particle, electron or positron, travels a step of length $s$ be-
tween two consecutive hard events of any kind (i.e. hard elastic or inelastic collisions,
and annihilation in the case of positrons). Along this step, the particle is assumed to interact only through soft inelastic collisions and soft bremsstrahlung emission. We consider that the average energy loss in this path length,
$S_s (E) s$, is much less than the initial energy $E$ so that the DCSs can be assumed to stay essentially constant along the step. Let $G(s; \omega)$ denote the PDF of the energy loss $\omega$ along the path length $s$; this distribution satisfies the transport equation (Landau, 1944)

$$\frac{\partial G(s; \omega)}{\partial s} = \mathcal{N} \int_{0}^{\infty} \left[ G(s; \omega - W) - G(s; \omega) \right] \sigma_s (E; W) \, dW \quad (4.48)$$

with the initial value $G(0; \omega) = \delta(\omega)$. Here, $\sigma_s (E; W)$ stands for the DCS for soft stopping interactions, i.e.

$$\sigma_s (E; W) = \frac{d\sigma_{\text{in}}}{dW} \Theta(W_{cc} - W) + \frac{d\sigma_{\text{br}}}{dW} \Theta(W_{cr} - W), \quad (4.49)$$
where $\Theta(x)$ is the step function. A closed formal solution of the integral equation (4.48) may be obtained by considering its Fourier, or Laplace, transform with respect to $\omega$ (see e.g. Landau, 1944, Blunk and Leisegang, 1950). For our purposes it is only necessary to know the first moments of the energy loss distribution after the path length $s$,

$$
\langle \omega^n \rangle \equiv \int_0^\infty \omega^n G(s; \omega) \, d\omega.
$$

(4.50)

From eq. (4.48) it follows that

$$
\frac{d}{ds} \langle \omega^n \rangle = N \int_0^\infty d\omega \int_0^\infty dW \omega^n \left[ G(s; \omega - W) - G(s; \omega) \right] \sigma_s(E; W)
$$

$$
= N \left( \int_0^\infty d\omega' \int_0^\infty dW (\omega' + W)^n G(s; \omega') \sigma_s(E; W) - \langle \omega^n \rangle \int_0^\infty \sigma_s(E; W) \, dW \right)
$$

$$
= \sum_{k=1}^n \frac{n!}{k!(n-k)!} \langle \omega^{n-k} \rangle N \int_0^\infty W^k \sigma_s(E; W) \, dW,
$$

(4.51)

where use has been made of the fact that $\sigma_s(E; W)$ vanishes when $W < 0$. In particular, we have

$$
\frac{d}{ds} \langle \omega \rangle = N \int_0^\infty W \sigma_s(E; W) \, dW = S_s,
$$

(4.52)

$$
\frac{d}{ds} \langle \omega^2 \rangle = 2\langle \omega \rangle N \int_0^\infty W \sigma_s(E; W) \, dW + N \int_0^\infty W^2 \sigma_s(E; W) \, dW
$$

$$
= 2\langle \omega \rangle S_s + \Omega_s^2
$$

(4.53)

and, hence,

$$
\langle \omega \rangle = S_s s,
$$

(4.54)

$$
\langle \omega^2 \rangle = (S_s s)^2 + \Omega_s^2 s.
$$

(4.55)

The variance of the energy loss distribution is

$$
\text{var}(\omega) = \langle \omega^2 \rangle - \langle \omega \rangle^2 = \Omega_s^2 s,
$$

(4.56)

i.e. the energy straggling parameter $\Omega_s^2$ equals the variance increase per unit path length.

The key point in our argument is that soft interactions involve only comparatively small energy losses. If the number of soft interactions along the path length $s$ is statistically sufficient, it follows from the central limit theorem that the energy loss distribution is Gaussian with mean $S_s s$ and variance $\Omega_s^2 s$, i.e.

$$
G(s; \omega) \approx \frac{1}{(2\pi \Omega_s^2 s)^{1/2}} \exp \left[ -\frac{(\omega - S_s(E)s)^2}{2\Omega_s^2 s} \right].
$$

(4.57)

This result is accurate only if 1) the average energy loss $S_s(E)s$ is much smaller than $E$ (so that the DCS $d\sigma_s/dW$ is nearly constant along the step) and 2) its standard
deviation \( \frac{\Omega_s^2(E)^{1/2}}{s} \) is much smaller than its mean \( s_s(E) \) (otherwise there would be a finite probability of negative energy losses), i.e.

\[
\frac{\Omega_s^2(E)^{1/2}}{s} \ll s_s(E) \ll E. \tag{4.58}
\]

Requirement 1) implies that the cutoff energies \( W_c \) and \( W_c \) for delta ray production and photon emission have to be relatively small. The second requirement holds for path lengths larger than \( s_{\text{crit}} = \frac{\Omega_s^2}{s_s^2} \).

Now, we address ourselves to the problem of simulating the energy losses due to soft stopping interactions between two consecutive hard events. The distribution (4.57) gives the desired result when conditions (4.58) are satisfied. In fact, the use of a Gaussian distribution to simulate the effect of soft stopping interactions was previously proposed by Andreo and Brahma (1984). Unfortunately, the step lengths found in our simulations are frequently too short for conditions (4.58) to hold (i.e. \( s \) is usually less than \( s_{\text{crit}} \)). To get over this problem, we replace the actual energy loss distribution \( G(s; \omega) \) by a simpler “equivalent” distribution \( G_a(s; \omega) \) with the same mean and variance, given by eqs. (4.54) and (4.56). Other details of the adopted distribution have no effect on the simulation results, provided that the number of steps along each track is statistically sufficient (say, larger than \( \sim 20 \)). PENELOPE generates \( \omega \) from the following distributions

- Case I. If \( \langle \omega \rangle^2 > 9 \text{var}(\omega) \), we use a truncated Gaussian distribution,

\[
G_a,1(s; \omega) = \begin{cases} 
\exp \left[ -\frac{(\omega - \langle \omega \rangle)^2}{2(1.015387 \sigma)^2} \right] & \text{if } |\omega - \langle \omega \rangle| < 3 \sigma, \\
0 & \text{otherwise}.
\end{cases} \tag{4.59}
\]

where \( \sigma = \sqrt{\text{var}(\omega)} \) is the standard deviation and the numerical factor 1.015387 corrects for the effect of the truncation. Notice that the shape of this distribution is very similar to that of the “true” energy-loss distribution, eq. (4.57). Random sampling from (4.59) is performed by means of the Box-Müller method, eq. (1.54), rejecting the generated \( \omega \)’s that are outside the interval \( \langle \omega \rangle \pm 3 \sigma \).

- Case II. When \( 3 \text{var}(\omega) < \langle \omega \rangle^2 < 9 \text{var}(\omega) \), the energy loss is sampled from the uniform distribution

\[
G_{a,2}(s; \omega) = U_{\omega_1, \omega_2}(\omega) \tag{4.60}
\]

with

\[
\omega_1 = \langle \omega \rangle - \sqrt{3} \sigma, \quad \omega_2 = \langle \omega \rangle + \sqrt{3} \sigma. \tag{4.61}
\]

- Case III. Finally, when \( \langle \omega \rangle^2 < 3 \text{var}(\omega) \), the adopted distribution is an admixture of a delta and a uniform distribution,

\[
G_{a,3}(s; \omega) = a \delta(\omega) + (1 - a) U_{\omega_0, \omega}(\omega) \tag{4.62}
\]

with

\[
a = \frac{3 \text{var}(\omega) - \langle \omega \rangle^2}{3 \text{var}(\omega) + 3 \langle \omega \rangle^2} \quad \text{and} \quad \omega_0 = \frac{3 \text{var}(\omega) + 3 \langle \omega \rangle^2}{2 \langle \omega \rangle}. \tag{4.63}
\]
It can be easily verified that these distributions have the required mean and variance. It is also worth noticing that they yield \( \omega \) values that are less than

\[
\omega_{\text{max}} = \begin{cases} 
\langle \omega \rangle + 3\sigma & \text{in case I,} \\
\omega_2 & \text{in case II,} \\
\omega_0 & \text{in case III.}
\end{cases}
\] (4.64)

\( \omega_{\text{max}} \) is normally much less than the kinetic energy \( E \) of the transported particle. Energy losses larger than \( E \) might be generated only when the step length \( s \) has a value of the order of the Bethe range, but this never happens in practical simulation (see below). It is worth noticing that, after a moderately large number of steps, this simple simulation scheme effectively yields an energy loss distribution that has the correct first and second moments and is similar in shape to the “true” distribution. Further improvements of the distribution of soft energy losses would mean considering higher order moments of the single scattering inelastic DCS given by eq. (4.49).

In spatial-dose calculations, the energy loss \( \omega \) due to soft stopping interactions can be considered to be locally deposited at a random position uniformly distributed along the step. This procedure yields dose distributions identical to those obtained by assuming that the energy loss is deposited at a constant rate along the step, but is computationally simpler. According to this, PENELLOPE simulates the combined effect of all soft elastic collisions and soft stopping interactions that occur between a pair of successive hard events, separated a distance \( s \), as a single event (a hinge) in which the particle changes its direction of movement according to the distribution \( F_n(s; \mu) \), eqs. (4.30)-(4.32), and loses energy \( \omega \) that is generated from the distribution \( G_n(s; \omega) \), eqs. (4.59)-(4.63). The position of the hinge is sampled uniformly along the step, as in the case of purely elastic scattering (section 4.1.2). When the step crosses an interface (see fig. 4.2), the artificial event is simulated only when its position lies in the initial material; otherwise the track is stopped at the interface and restarted in the new material. It can be easily verified that the particle reaches the interface not only with the correct average direction of movement, but also with the correct average energy, \( E = S_t \).

### 4.2.1 Energy dependence of the soft DCS

The simulation model for soft energy losses described above is based on the assumption that the associated energy-loss DCS does not vary with the energy of the transported particle. To account for the energy dependence of the DCS in a rigorous way, we have to start from the transport equation [cf. eq. (4.48)]

\[
\frac{\partial G(s; \omega)}{\partial s} = N \int_0^\infty G(s; \omega - W) \sigma_s(E_0 - \omega + W; W) \, dW - N \int_0^\infty G(s; \omega) \sigma_s(E_0 - \omega; W) \, dW;
\] (4.65)

where \( E_0 \) denotes the kinetic energy of the particle at the beginning of the step. We desire to obtain expressions for the first and second moments, \( \langle \omega \rangle \) and \( \langle \omega^2 \rangle \), of the multiple
4.2. Soft energy losses

scattering energy-loss distribution, which define the artificial distribution \( G_\alpha(s; \omega) \) as described above. Unfortunately, for a realistic DCS, these moments can only be obtained after arduous numerical calculations and we have to rely on simple approximations that can be easily implemented in the simulation code.

Let us consider that, at least for relatively small fractional energy losses, the DCS varies linearly with the kinetic energy of the particle,

\[
\sigma_s(E_0 - \omega; W) \simeq \sigma_s(E_0; W) - \left[ \frac{\partial \sigma_s(E; W)}{\partial E} \right]_{E=E_0} \omega. \tag{4.66}
\]

We recall that we are considering only soft energy-loss interactions (inelastic collisions and bremsstrahlung emission) for which the cutoff energies, \( \omega_{\infty} \) and \( \omega_{\text{cr}} \), do not vary with \( E \). Therefore, the upper limit of the integrals in the right hand side of eq. (4.65) is finite and independent of the energy of the particle. The stopping power \( S_s(E_0 - \omega) \) can then be approximated as

\[
S_s(E_0 - \omega) \equiv \mathcal{N} \int W \sigma_s(E_0 - \omega; W) \, dW \simeq S_s(E_0) - S_s'(E_0) \omega, \tag{4.67}
\]

where the prime denotes the derivative with respect to \( E \). Similarly, for the straggling parameter \( \Omega_s^2(E) \) we have

\[
\Omega_s^2(E_0 - \omega) \equiv \mathcal{N} \int W^2 \sigma_s(E_0 - \omega; W) \, dW \simeq \Omega_s^2(E_0) - \Omega_s^{2'}(E_0) \omega. \tag{4.68}
\]

From eq. (4.65) it follows that the moments of the multiple scattering distribution,

\[
\langle \omega^n \rangle = \int \omega^n G(s; \omega) \, d\omega,
\]

satisfy the equations

\[
\frac{d}{ds} \langle \omega^n \rangle = \mathcal{N} \int dw \int dW \left[ (w + W)^n G(s; w) \sigma_s(E_0 - w; W) \right]
\]

\[
- \mathcal{N} \int dw \int dW \, w^n G(s; w) \sigma_s(E_0 - w; W)
\]

\[
= \mathcal{N} \sum_{k=1}^{n} \frac{n!}{k!(n-k)!} \int d\omega \int dW \, \omega^{n-k} W^k G(s; \omega) \sigma_s(E_0 - \omega; W). \tag{4.69}
\]

By inserting the approximation (4.66), we obtain

\[
\frac{d}{ds} \langle \omega^n \rangle = \sum_{k=1}^{n} \frac{n!}{k!(n-k)!} \left( \langle \omega^{n-k} \rangle M_k - \langle \omega^{n-k+1} \rangle M_k' \right), \tag{4.70}
\]

where

\[
M_k \equiv \mathcal{N} \int W^k \sigma_s(E_0; W) \, dW \tag{4.71}
\]
and

\[ M'_k \equiv \mathcal{N} \int W^k \left[ \frac{\partial \sigma_s(E; W)}{\partial E} \right]_{E=E_0} \, dW = \left[ \frac{dM_k}{dE} \right]_{E=E_0}. \]  

(4.72)

The equations (4.70) with the boundary conditions \( \langle \omega^n \rangle_{s=0} = 0 \) can now be solved sequentially to any order. For \( n = 1 \) we have

\[ \frac{d}{ds} \langle \omega \rangle = S_s(E_0) - S'_s(E_0) \langle \omega \rangle, \]  

(4.73)

which yields

\[ \langle \omega \rangle = \frac{S_s(E_0)}{S'_s(E_0)} \left\{ 1 - \exp \left[ -S'_s(E_0)s \right] \right\}. \]  

(4.74)

The equation for \( n = 2 \) reads,

\[ \frac{d}{ds} \langle \omega^2 \rangle = \Omega_s^2(E_0) \left[ 2S_s(E_0) - \Omega_s^{2f}(E_0) \right] \langle \omega \rangle - 2S'_s(E_0) \langle \omega \rangle, \]  

(4.75)

and its solution is

\[ \langle \omega^2 \rangle = \frac{\Omega_s^2(E_0)}{2S'_s(E_0)} \left[ 1 - \exp \left[ -2S'_s(E_0)s \right] \right] \]  

\[ + s \left[ 2S_s(E_0) - \Omega_s^{2f}(E_0) \right] S_s(E_0) \left( \frac{1 - \exp \left[ -S'_s(E_0)s \right]}{2S'_s(E_0)} \right)^2. \]  

(4.76)

Hence,

\[ \text{var}(\omega) = \langle \omega^2 \rangle - \langle \omega \rangle^2 \]

\[ = \frac{\Omega_s^2(E_0)}{2S'_s(E_0)} \left[ 1 - \exp \left[ -2S'_s(E_0)s \right] \right] - 2\Omega_s^{2f}(E_0)S_s(E_0) \left( \frac{1 - \exp \left[ -S'_s(E_0)s \right]}{2S'_s(E_0)} \right)^2. \]  

(4.77)

Since these expressions are derived from the linear approximation, eq. (4.66), it is consistent to evaluate \( \langle \omega \rangle \) and \( \text{var}(\omega) \) from their Taylor expansions to second order,

\[ \langle \omega \rangle = S_s(E_0) s \left[ 1 - \frac{1}{2} S'_s(E_0) s + \mathcal{O}(s^3) \right] \]

\[ \approx S_s(E_0) s \left\{ 1 - \frac{1}{2} \left[ \frac{d \ln S_s(E)}{dE} \right]_{E=E_0} S_s(E_0) s \right\} \]  

(4.78)

and

\[ \text{var}(\omega) = \Omega_s^2(E_0)s - \left[ \frac{1}{2} \Omega_s^{2f}(E_0) S_s(E_0) + \Omega_s^2(E_0) S'_s(E_0) \right] s^2 + \mathcal{O}(s^3) \]

\[ \approx \Omega_s^2(0)s \left\{ 1 - \left[ \frac{1}{2} \frac{d \ln \Omega_s^2(E)}{dE} + \frac{d \ln S_s(E)}{dE} \right]_{E=E_0} S_s(E_0) s \right\}, \]  

(4.79)
where the logarithmic derivatives have been introduced for numerical convenience. The factors in curly brackets account for the global effect of the energy dependence of the soft energy-loss DCS (within the linear approximation). To simulate soft energy losses, we sample $\omega$ from the artificial distribution $G_\omega(\omega; s)$, eqs. (4.59) to (4.63), with the “correct” first moment and variance, given by expressions (4.78) and (4.79). In PENELOE, we use step lengths $s$ such that the fractional energy loss along each step is relatively small (see below) and, consequently, the energy-dependence correction is also small (i.e. the correcting factors are close to unity).

### 4.3 Combined scattering and energy loss

Up to this point, soft scattering and energy loss have been regarded as essentially independent processes, while in reality they coexist. In this section, we consider their interplay and set the basis of an algorithm that simulates their combined effect.

Ours is a mixed algorithm, where hard simulations are described individually from the associated DCSs (see chapter 3). These interactions are 1) hard elastic collisions, “el”, 2) hard inelastic collisions, “in”, 3) hard bremsstrahlung photon emission “br”, and, in the case of positrons, 4) positron annihilation, “an”. The mean free path between consecutive hard events, $\lambda_T^{(h)}$, is given by

$$
\left[ \lambda_T^{(h)} \right]^{-1} = N \sigma_T^{(h)} = N \left[ \sigma_d^{(h)} + \sigma_{\text{in}}^{(h)} + \sigma_{\text{br}}^{(h)} \right] \equiv \Sigma_h,
$$

(4.80)

where $\sigma_T^{(h)}$ is the total atomic cross section for hard interactions. We recall that the inverse mean free path, $\Sigma_h$, gives the interaction probability per unit path length. In the absence of soft energy-loss events, the PDF of the step length $s$ between two successive hard events (or from a given point in the track to the next hard event) is

$$
p(s) = \Sigma_h \exp \left( -\Sigma_h s \right).
$$

(4.81)

In each hard event, one and only one interaction (i.e. “el”, “in”, “br” or “an”) occurs with probability

$$
p_i = \sigma_i^{(h)} / \sigma_T^{(h)}.
$$

(4.82)

When soft energy-losses are considered, the PDF of the distance $s$ travelled by the particle to the following hard interaction is not given by eq. (4.81), because the mean free path $\lambda_T^{(h)}$ varies with energy and may change appreciably along a single step. The simplest way to cope with this problem is to limit the length of the step to make sure that the average energy loss is much smaller than the kinetic energy $E$ at the beginning of the step, and consider that $\lambda_T^{(h)}(E)$ remains essentially constant along the step. Then, the mean energy loss in a step is given by

$$
\langle \Delta E \rangle = \lambda_T^{(h)} S(E),
$$

(4.83)
where
\[ S(E) = S_{in}(E) + S_{br}(E) \] (4.84)
is the total stopping power. Since the mean free path between consecutive hard events of any kind is shorter than the mean free path between hard elastic events, the energy loss per step can be limited by re-defining the hard mean free path. If we wish to tolerate average fractional energy losses \( \Delta E/E \) along a step of the order of \( C_2 \) (a small value, say, 0.05), we simply take
\[
\lambda_{el}^{(h)}(E) = \max \left\{ \lambda_{el}(E), \min \left[ C_1 \lambda_{el,1}(E), C_2 \frac{E}{S(E)} \right] \right\}. \] (4.85)

This effectively limits the average energy loss per step at the expense of increasing the frequency of hard elastic events. The parameters \( C_1 \) and \( C_2 \) in eq. (4.85), to be selected by the user, determine the computer time needed to simulate each track. Ideally, they should not have any influence on the accuracy of the simulation results. This happens only when their values are sufficiently small (see below).

**Figure 4.3:** Elastic mean free path \( \lambda_{el} \), first transport mean free path \( \lambda_{el,1} \), and \( E/S(E) \) for electrons in aluminium and lead. The solid line represents the mean free path between hard elastic events \( \lambda_{el}^{(h)} \) obtained from eq. (4.85) with \( C_1 = C_2 = 0.05 \).

It should be noted that \( C_1 \) and \( C_2 \) act on different energy domains. This is illustrated in fig. 4.3, where the lengths \( \lambda_{el} \), \( \lambda_{el,1} \) and \( E/S \) for electrons in aluminium and lead are represented as functions of the kinetic energy. The mean free path \( \lambda_{el}^{(h)} \) for hard elastic
events, determined from the prescription \((4.85)\) with \(C_1 = C_2 = 0.05\) is also plotted. For low energies, \(\lambda_{el}^{(h)} = \lambda_{el}\) and the simulation is purely detailed (\(\mu_c = 0\)). For intermediate energies, \(\lambda_{el}^{(h)} = C_1 \lambda_{el,1}\), whereas \(\lambda_{el}^{(h)} = C_2 E / S(E)\) in the high-energy domain. From fig. 4.3 it is clear that increasing the value of \(C_2\) does not have any effect on the simulation of electron tracks with initial energies that are less than \(\sim 10\) MeV.

4.3.1 Variation of \(\lambda_T^{(h)}\) with energy

With the definition \((4.85)\) of the hard elastic mean free path, we only set a limit on the \textit{average} step length. However, since \(s\) is sampled from the exponential distribution, its realizations fluctuate amply about the average value. On the other hand, the soft energy loss \(\omega\) along a step of given length \(s\) also fluctuates about the mean value \(\langle \omega \rangle\) given by eq. \((4.78)\). This means that the inverse mean free path \(\Sigma_h(E)\) varies along the step in an essentially unpredictable way.

Let us consider for a moment that the CSDA is applicable (i.e. that the effect of soft energy straggling is negligible). In this case, there is a one-by-one correspondence between the kinetic energy \(E\) of the electron and the travelled path length \(s\),

\[
s = \int_{E}^{E_0} \frac{dE'}{S_s(E')}, \tag{4.86}
\]

where \(S_s(E)\) is the soft stopping power, eq. \((4.46)\) [we consider that no hard interactions occur along the step]. Equivalently,

\[
\frac{ds}{dE} = \frac{1}{S_s(E)}. \tag{4.87}
\]

Thus, the inverse mean free path \(\Sigma_h\) can be formally considered as a function of the path length \(s\). The probability \(p(s) ds\) of having the first hard interaction when the particle has travelled a length in the interval \((s, s + ds)\) is determined by the equation [cf. eq. \((1.72)\)]

\[
p(s) = \Sigma_h(s) \int_{s}^{\infty} p(s') ds', \tag{4.88}
\]

with the normalization condition,

\[
\int_{0}^{\infty} p(s) ds = 1. \tag{4.89}
\]

Instead of the path length \(s\), it is convenient to consider the dimensionless variable

\[
q \equiv \int_{E}^{E_0} \frac{\Sigma_h(E')}{S_s(E')} dE' = \int_{0}^{s} \Sigma_h(s') ds', \tag{4.90}
\]

which varies with energy and

\[
\frac{dq}{dE} = \frac{\Sigma_h(E)}{S_s(E)} \tag{4.91}
\]
The PDF of $q$ is
\[ \pi(q) = p(s) \frac{ds}{dq} = p(s) \frac{ds}{dE} \frac{dE}{dq} = p(s) \frac{1}{\Sigma_h(s)}. \] (4.92)

From eq. (4.88) it follows that $\pi(q)$ satisfies the equation
\[ \pi(q) = \int_q^\infty \pi(q') dq'. \] (4.93)

Therefore, $q$ is distributed exponentially,
\[ \pi(q) = \exp(-q). \] (4.94)

The PDF of the step length $s$ is obtained by inverting the transformation (4.90),
\[ p(s) = \Sigma_h(s) \exp\left(-\int_0^s \Sigma_h(s') ds'\right). \] (4.95)

It is not practical to sample $s$ from this complicated PDF. It is much more convenient to sample $q$ [as $-\ln \xi$, cf. eq. (1.36)] and then determine $s$ from (4.90), which can be inverted numerically (for practical details, see Berger, 1998). Although this sampling method effectively accounts for the energy dependence of $\Sigma_s(E)$, it is applicable only to simulations in the CSDA.

A more versatile algorithm for sampling the position of hard events, still within the CSDA, is the following. We let the electron move in steps of maximum length $s_{\text{max}}$, a value specified by the user. This determines the maximum energy loss along the step,
\[ \omega_{\text{max}} = \int_0^{s_{\text{max}}} S_h(s) ds. \] (4.96)

Let $\Sigma_{h,\text{max}}$ denote an upper bound for the inverse mean free path of hard events in the swept energy interval, i.e.
\[ \Sigma_{h,\text{max}} > \max\{\Sigma_h(E), E \in (E_0 - \omega_{\text{max}}, E_0)\}. \] (4.97)

We now assume that the electron may undergo fictitious events in which the energy and direction remain unaltered (delta interactions). The inverse mean free path of these interactions is defined as
\[ \Sigma_\delta(E) = \Sigma_{h,\text{max}} - \Sigma_h(E), \] (4.98)
so that the inverse mean free path of the combined process (delta interactions + hard events) equals $\Sigma_{h,\text{max}}$, a constant. Owing to the Markovian character of the processes, the introduction of delta interactions does not influence the path-length distribution between hard events. Therefore, the occurrence of hard events can be sampled by means of the following simple algorithm,

(i) Sample a distance $s$ from the exponential distribution with inverse mean free path $\Sigma_{h,\text{max}}$, i.e. $s = (-\ln \xi)/\Sigma_{h,\text{max}}$. 

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(ii) If $s > s_{\text{max}}$, move the electron a path length $s_{\text{max}}$ and determine the soft energy loss $\omega$ along this path length. Modify the electron energy\(^2\), $E \leftarrow E - \omega$, and assume that a delta interaction occurs at the end of the step.

(iii) If $s < s_{\text{max}}$, move the electron a step of length $s$. Determine the energy loss $\omega$ and update the energy, $E \leftarrow E - \omega$. Sample a random number $\xi$.

1. If $\xi \Sigma_{h_{\text{max}}} < \Sigma_h(E)$, simulate a hard interaction
2. Otherwise, assume that the particle undergoes a delta interaction.

(iv) Return to (i).

It is clear that the path-length $s$ to the first hard interaction generated with this algorithm follows the PDF (4.95). The interesting peculiarity of this algorithm is that it makes no explicit reference to the CSDA. Therefore, it can be adopted in mixed simulations with soft-energy-loss straggling, provided only that an upper bound exists for the energy $\omega$ lost along the path length $s_{\text{max}}$.

![Graphs showing inverse mean free path for Al and Au](image)

**Figure 4.4:** Inverse mean free path (interaction probability per unit path length) for hard interactions of electrons in Al and Au for the indicated values of the simulation parameters. The plotted curves were calculated with $W_{cc} = W_{cr} = 100$ eV.

Fortunately, the energy loss generated from the artificial distribution $G_a(\omega; s)$, eqs. (4.59)-(4.63), is always less than $\omega_{\text{max}}$, eq. (4.64). Indeed, in case I we use the truncated

\(^2\)In the description of the algorithms we use the symbol $\leftarrow$ in expressions such as “$a \leftarrow b$” to indicate that the value $b$ replaces the value of $a$. 
Gaussian distribution (4.59) just to enforce this property. In our mixed simulation we shall select a maximum step length $s_{\text{max}}$, which serves to set an upper bound for the energy that the transported electrons may lose along each step. Since the hard inverse mean free path $\Sigma_h(E)$ has a broad minimum (and no local maxima) in the whole energy interval of interest (see fig. 4.4), the maximum value of $\Sigma_h$ within a certain energy interval ($E_1, E_2$) occurs at one of the end points. This makes the practical implementation of the above algorithm very easy.

### 4.3.2 Scattering by atomic electrons

Most of the existing high-energy simulation codes have difficulties in accounting for the angular deflections of the projectile due to inelastic collisions (see e.g. Jenkins et al., 1988). The inelastic cross section differential in the scattering angle can be calculated approximately in terms of the incoherent scattering function (see e.g. Mott and Massey, 1965). This was the approach followed by Fano (1954) in order to introduce electron scattering effects in the Molière (1948) multiple scattering theory. However, the DCS calculated in this way accounts for all excitations and, hence, it is not adequate for mixed simulations, where the part of electron scattering due to hard collisions is explicitly simulated. Moreover, the calculation of the DCS from the incoherent scattering function involves an average over excitation energies that cannot be performed exactly; instead an effective “minimum momentum transfer” is introduced, which must be estimated empirically. This may cause inconsistencies for low-energy projectiles. A more consistent approach (Baró et al., 1995) is obtained by simply computing the restricted angular DCS, for soft collisions with $W < W_{cc}$, from our inelastic scattering model (see section 3.2), as follows.

We recall that the recoil energy $Q$ is given by (see appendix B)

$$ Q(Q + 2m_e c^2) = c^2 (p^2 + p'^2 - 2pp' \cos \theta), \quad (4.99) $$

where $p$ and $p'$ are the magnitudes of the momentum of the projectile before and after the collision,

$$(cp)^2 = E(E + 2m_e c^2) \quad \text{and} \quad (cp')^2 = (E - W)(E - W + 2m_e c^2). \quad (4.100)$$

In soft distant interactions, the angular deflection $\mu = (1 - \cos \theta)/2$ and the recoil energy $Q$ are related through

$$ Q(Q + 2m_e c^2) = 4cp\, c_p\, \mu + (cp - c_p)^2, \quad (4.101) $$

where $p_k$ is the momentum of the projectile after the collision,

$$(c_p)^2 = (E - W_k)(E - W_k + 2m_e c^2). \quad (4.102)$$

The cross section for soft distant interactions\(^3\), eq. (3.60), can then be expressed in terms

\(^3\)Distant transverse interactions do not cause scattering.
of the variable $\mu$ as

$$
\frac{d\sigma_{\text{dis},l}}{d\mu} = \frac{4\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W_k} \frac{1}{R_k + \mu} \frac{1}{4 \alpha' c \rho_k} \frac{m_e e^2}{2(Q + m_e e^2)},
$$

(4.103)

Considering that $Q \ll m_e e^2$ for the majority of soft distant collisions, we have

$$
\frac{d\sigma_{\text{dis},l}}{d\mu} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W_k} \frac{1}{R_k + \mu}, \quad 0 < \mu < \mu_k,
$$

(4.104)

where

$$
R_k = \frac{(c \rho - c \rho_k)^2}{4 \alpha' c \rho_k}
$$

(4.105)

and

$$
\mu_k = \mu(Q = W_k) = \frac{W_k(W_k + 2m_e e^2) - (c \rho - c \rho_k)^2}{4 \alpha' c \rho_k}.
$$

(4.106)

On the other hand, the DCS per unit oscillator strength for soft ($W < W_{cc}$) close collisions with the $i$-th oscillator is given by [see eqs. (3.71) and (3.77)]

$$
\frac{d\sigma_{\text{clo}}^{(\pm)}}{dW} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W^2} f^{(\pm)}(E, W).
$$

(4.107)

The angular deflection and the energy loss are related by (3.118), which implies that

$$
W = \frac{E(E + 2m_e e^2)(\mu - \mu^2)}{2E(\mu - \mu^2) + m_e e^2}
$$

(4.108)

and

$$
\frac{dW}{d\mu} = \frac{E(E + 2m_e e^2)m_e e^2(1 - 2\mu)}{[2E(\mu - \mu^2) + m_e e^2]^2}.
$$

(4.109)

Therefore,

$$
\frac{d\sigma_{\text{clo}}^{(\pm)}}{d\mu} = \frac{2\pi e^4}{m_e v^2} \sum_k f_k \frac{1}{W^2} f^{(\pm)}(E, W) \frac{dW}{d\mu}, \quad \mu_k < \mu < \mu_{cc},
$$

(4.110)

where

$$
\mu_{cc} = \mu(Q = W_{cc}) = \frac{W_{cc}(W_{cc} + 2m_e e^2) - (c \rho - c \rho_{cc})^2}{4 \alpha' c \rho_{cc}}
$$

(4.111)

with

$$(c \rho_{cc})^2 = (E - W_{cc})(E - W_{cc} + 2m_e e^2).
$$

(4.112)

The angular DCS for soft inelastic interactions is then given by

$$
\frac{d\sigma_{\text{s}}}{d\mu} = \frac{d\sigma_{\text{dis},l}}{d\mu} + \frac{d\sigma_{\text{clo}}^{(\pm)}}{d\mu}
$$

$$
= \frac{2\pi e^4}{m_e v^2} \sum_k f_k \left\{ \frac{1}{W_k} \frac{1}{R_k + \mu} + \frac{1}{W^2} f^{(\pm)}(E, W) \frac{dW}{d\mu} \right\},
$$

(4.113)
where the summations extend over the oscillators with resonance energy less than \( W_{cc} \) and greater than \( W_{\text{max}} \), and each term contributes only for the \( \mu \)-intervals indicated above. The mean free path and the first and second transport mean free paths for soft inelastic scattering are

\[
\left[ \lambda_{\text{in}}^{(s)} \right]^{-1} = N \int_{0}^{\mu_2} \frac{d\sigma_{\text{in}}^{(s)}}{d\mu} \, d\mu,
\]

(4.114)

\[
\left[ \lambda_{\text{in},1}^{(s)} \right]^{-1} = N \int_{0}^{\mu_2} \frac{d\sigma_{\text{in}}^{(s)}}{d\mu} \, d\mu
\]

(4.115)

and

\[
\left[ \lambda_{\text{in},2}^{(s)} \right]^{-1} = N \int_{0}^{\mu_2} \frac{d\sigma_{\text{in}}^{(s)}}{d\mu} \, d\mu.
\]

(4.116)

In PENEOPE, soft electronic scattering is simulated together with soft elastic scattering, by means of the artificial distribution (4.30). The combined process is described by the transport mean free paths

\[
\left[ \lambda_{\text{comb},1}^{(s)} \right]^{-1} = \left[ \lambda_{\text{el},1}^{(s)} \right]^{-1} + \left[ \lambda_{\text{in},1}^{(s)} \right]^{-1}
\]

(4.117)

and

\[
\left[ \lambda_{\text{comb},2}^{(s)} \right]^{-1} = \left[ \lambda_{\text{el},2}^{(s)} \right]^{-1} + \left[ \lambda_{\text{in},2}^{(s)} \right]^{-1}.
\]

(4.118)

Thus, to account for soft electronic scattering we only have to replace the soft elastic transport mean free paths by those of the combined process.

### 4.3.3 Bielajew’s alternate random hinge

Angular deflections due to soft interactions along a step of length \( s \) are generated from the artificial distribution (4.30) with first and second moments given by eqs. (4.28) and (4.29), which are determined by the transport mean free paths \( \lambda_{\text{comb},1}^{(s)} \) and \( \lambda_{\text{comb},2}^{(s)} \). To account (at least partially) for the energy dependence of these quantities we use a trick due to Alex Bielajew. The soft energy loss and angular deflection (which occur at the hinge) are considered as independent processes and are simulated in random order. That is, the soft angular deflection is evaluated for the energy at either the beginning or the end of the step, with equal probabilities. This is equivalent to assuming that the transport mean free paths \( \lambda_{\text{comb},1}^{(s)}(E) \) and \( \lambda_{\text{comb},2}^{(s)}(E) \) vary linearly with energy. The method is fairly accurate and computationally inexpensive provided only that the fractional energy loss along each step (which is of the order of \( C_2 \)) is sufficiently small.

### 4.4 Generation of random tracks

Each simulated electron or positron history consists of a chronological succession of events. These can be either hard events, artificial soft events (hinges) or other relevant
stages of the particle history (such as its initial state, the crossing of an interface or the effective absorption after slowing down). The trajectory of the particle between a pair of successive events is straight and will be referred to as a “segment”. We keep the term “step” to designate the portion of a track between two hard events, which consists of two segments and a hinge (when mixed simulation is effective).

Simulation with PENELope is controlled by the constants $C_1$ and $C_2$ [see eq. (4.85)] and the cutoff energies $W_{cc}$ and $W_{cr}$. Hereafter, these four quantities will be referred to as simulation parameters. The parameter $C_1$, which determines the mean free path $\lambda^{(h)}_d$ between hard elastic events, should be small enough to ensure reliable simulation results. PENELope admits values of $C_1$ from 0 (detailed simulation) up to 0.2, which corresponds to a mean angular deflection $\langle \theta \rangle \sim 37$ deg after a step length $\lambda^{(h)}_d$. The simulation parameter $C_2$ gives the maximum average fractional energy loss in a single step and it is effective only at high energies. From the discussion in section 4.3, it is clear that $C_2$ should also be small. PENELope allows values of $C_2$ between zero and 0.2. The cutoff energies $W_{cc}$ and $W_{cr}$ mainly influence the simulated energy distributions. The simulation speeds up by using larger cutoff energies, but if these are too large the simulated distributions may be somewhat distorted. In practice, simulated energy distributions are found to be quite insensitive to the adopted values of $W_{cc}$ and $W_{cr}$ when these are less than the bin width used to tally the energy distributions. Thus, the desired energy resolution determines the maximum allowed cutoff energies.

The combined effect of all soft elastic and stopping interactions in a step is simulated as a single artificial event or hinge, in which the particle changes its direction of movement and loses energy. When $W_{cc}$ is less than the lowest oscillator resonance energy, the simulation of inelastic collisions becomes purely detailed, i.e. inelastic collisions do not contribute to the soft stopping power. On the other hand, the simulation of bremsstrahlung emission is only possible by means of a mixed scheme, because of the divergence of the DCS at $W = 0$ [see eq. (3.126)]. To test the accuracy of mixed algorithms, and also in studies of low-energy electron and positron transport (with, say, $E < 100$ keV), it may be convenient to perform strictly detailed simulations (see below). For this purpose, PENELope allows the user to switch off the emission of soft bremsstrahlung photons with energy less than 10 eV. This option is activated when the $W_{cr}$ value selected by the user is negative, in which case the program sets $W_{cr} = 10$ eV, disregards soft bremsstrahlung events and simulates hard events (with $W > 10$ eV) in a detailed way. The generation of the angular deflection in artificial events is discontinued when the simulation of elastic and inelastic scattering becomes detailed (i.e. when $\lambda^{(h)}_d = \lambda_d$, $W_{cc} = 0$).

As indicated above, the length of the steps generated by PENELope is always less than $s_{max}$, an upper bound selected by the user. The simulation code limits the step length by placing delta interactions along the particle track. These are fictitious interactions that do not alter the state of the particle. Their only effect is to interrupt the sequence of simulation operations, which requires altering the values of inner control variables to permit resuming the simulation in a consistent way. The use of bounded step lengths is necessary to account for the energy dependence of the DCSs for soft interactions.
However, this is not the only reason for limiting the step length. Since energy losses and deflections at the hinges are sampled from artificial distributions, the number of hinges per primary track must be "statistically sufficient," i.e. larger than \( \sim 10 \), to smear off the unphysical details of the adopted artificial distributions. Therefore, when the particle is in a thin region, it is advisable to use a small value of \( s_{\text{max}} \) to make sure that the number of hinges within the material is sufficient. In PENELOE, the parameter \( s_{\text{max}} \) can be varied freely during the course of the simulation of a single track. To ensure internal consistency, \( s_{\text{max}} \) is required to be less than \( 3\lambda_{\text{T}}^{(h)} \). When the user-selected value is larger, the code sets \( s_{\text{max}} = 3\lambda_{\text{T}}^{(h)} \); in this case, about 5 per cent of the sampled steps have lengths that exceed \( s_{\text{max}} \) and are terminated by a delta interaction. This slows down the simulation a little (\( \sim 5\% \)), but ensures that the energy dependence of \( \lambda_{\text{T}}^{(h)} \) is correctly accounted for. Instead of the \( s_{\text{max}} \) value set by the user, PENELOE uses a random maximum step length \([0,s_{\text{max}}]\) that averages to half the user's value; this is used to eliminate an artifact in the depth-dose distribution from parallel electron/positron beams near the entrance interface. Incidentally, limiting the step length is also necessary to perform simulation of electron/positron transport in external static electromagnetic fields (see appendix C).

The state of the particle immediately after an event is defined by its position coordinates \( \mathbf{r} \), energy \( E \) and direction cosines of its direction of movement \( \mathbf{d} \), as seen from the laboratory reference frame. It is assumed that particles are locally absorbed when their energy becomes smaller than a preselected value \( E_{\text{abs}} \); positrons are considered to annihilate after absorption. The practical generation of random electron and positron tracks in arbitrary material structures, which may consist of several homogeneous regions of different compositions separated by well-defined surfaces (interfaces), proceeds as follows:

(i) Set the initial position \( \mathbf{r} \), kinetic energy \( E \) and direction of movement \( \mathbf{d} \) of the primary particle.

(ii) Determine the maximum allowed soft energy loss \( \omega_{\text{max}} \) along a step and set the value of inverse mean free path for hard events (see section 4.3). The results depend on the adopted \( s_{\text{max}} \), which can vary along the simulated track.

(iii) Sample the distance \( s \) to be travelled to the following hard event (or delta interaction) as

\[
s = -\ln \frac{\xi}{\Sigma_{\text{h,max}}}. \tag{4.119}\]

If \( s > s_{\text{max}} \), truncate the step by setting \( s = s_{\text{max}} \).

(iv) Generate the length \( \tau = s\xi \) of the step to the next hinge. Let the particle advance this distance in the direction \( \mathbf{d} \): \( \mathbf{r} \leftarrow \mathbf{r} + \tau \mathbf{d} \).

(v) If the track has crossed an interface:

   Stop it at the crossing point (i.e. redefine \( \mathbf{r} \) as equal to the position of this point and set \( \tau \) equal to the travelled distance).

   Go to (ii) to continue the simulation in the new material, or go to (xi) if the new material is the outer vacuum.

(vi) Simulate the energy loss and deflection at the hinge. This step consists of two actions:
4.4. Generation of random tracks

a) Sample the polar angular deflection \( \mu = (1 - \cos \theta)/2 \) from the distribution \( F_\alpha(s; \mu) \), eq. (4.30), corresponding to the current energy \( E \). Sample the azimuthal scattering angle as \( \phi = 2\pi \xi \). Perform a rotation \( R(\theta, \phi) \) of the vector \( \hat{d} \) according to the sampled polar and azimuthal angular deflections (as described in section 1.4.2) to obtain the new direction: \( \hat{d} \leftarrow R(\theta, \phi)\hat{d} \).

b) Sample the energy loss \( \omega \) due to soft stopping interactions along the step \( s \) from the distribution \( G_\alpha(s; \omega) \), eqs. (4.59)-(4.63), and reduce the kinetic energy: \( E \leftarrow E - \omega \).

These two actions are performed in random order to account for the energy dependence of the soft transport mean free paths (see section 4.3.3).

Go to (xi) if \( E < E_{\text{abs}} \).

(vii) Let the particle advance the distance \( s - \tau \) in the direction \( \hat{d} \): \( \mathbf{r} \leftarrow \mathbf{r} + (s - \tau)\hat{d} \).

(viii) Do as in (v).

(ix) If in step (iii) the step length was truncated, i.e. \( s = s_{\text{max}} \), simulate a delta interaction.

Go to (ii).

(x) Simulate the hard event:

Sample the kind of interaction according to the point probabilities,

\[
p_{\text{el}} = \frac{N\sigma_{\text{el}}}{\sum_{h,\text{max}}} \quad p_{\text{in}} = \frac{N\sigma_{\text{in}}}{\sum_{h,\text{max}}} \quad p_{\text{br}} = \frac{N\sigma_{\text{br}}}{\sum_{h,\text{max}}} \quad p_{\delta} = \frac{N\sigma_{\delta}}{\sum_{h,\text{max}}}
\]

and \( p_{\text{an}} = \frac{N\sigma_{\text{an}}}{\sum_{h,\text{max}}} \) in the case of positrons. (4.120)

If the event is a delta interaction, return to (ii).

Sample the polar scattering angle \( \theta \) and the energy loss \( W \) from the corresponding DCS. Generate the azimuthal scattering angle as \( \phi = 2\pi \xi \). Perform a rotation \( R(\theta, \phi) \) of the vector \( \hat{d} \) to obtain the new direction: \( \hat{d} \leftarrow R(\theta, \phi)\hat{d} \).

Reduce the kinetic energy of the particle: \( E \leftarrow E - W \).

If, as a result of the interaction, a secondary particle is emitted in a direction \( \hat{d}_s \), with energy \( E_s > E_{\text{abs}} \), store its initial state \((\mathbf{r}, E_s, \hat{d}_s)\).

Go to (ii) if \( E > E_{\text{abs}} \).

(xi) Simulate the tracks of the secondary electrons and photons produced by the primary particle (or by other secondaries previously followed) before starting a new primary track.

4.4.1 Stability of the simulation algorithm

The present simulation scheme for electrons/positrons is relatively stable under variations of the simulation parameters, due mostly to the effectiveness of the energy-loss corrections. This implies that the simulation parameters can be varied amply without practically altering the accuracy of the results. For the important case of low-energy electrons/positrons (with energies of the order of 500 keV or less), the relevant parameters are \( E_{\text{abs}}, C_1, W_\infty \) and \( s_{\text{max}} \), because \( C_2 \) is not effective (see fig. 4.3) and radiative
Figure 4.5: Results from the simulations of 500 keV electrons in aluminium described in the text. Crosses, detailed simulation; continuous curves, mixed simulation. $p(z)$ is the PDF of the $z$-coordinate of the final electron position, after travelling the prescribed 200 µm. $p(\theta)$ and $p(E)$ are the PDFs of the direction of motion (specified by the polar angle $\theta$) and the kinetic energy $E$ of the electrons at the end of the simulated tracks. The function $D(z)$ represents the “depth-dose” function, i.e. the average energy deposited in the material per unit length along the $z$-direction (the residual energy at the end of the track is not included in the dose).
emission is unimportant (hard bremsstrahlung events occur very seldom and, therefore, \(W_{cr}\) has no influence). The value of the parameter \(s_{\text{max}}\) is important to ensure the reliability of the results; a safe recipe is to set \(s_{\text{max}}\) equal to one tenth of the “expected track length” or less. Since the values of \(E_{\text{abs}}\) and \(W_{cr}\) are dictated by the characteristics of the considered experiment, it follows that the only “critical” parameter, with a direct influence on the speed of the simulation, is \(C_1\). As mentioned above, PENEOPE accepts values of \(C_1\) ranging from 0 (detailed simulation of elastic scattering) to 0.2.

In practice, the value of \(C_1\) does not influence the accuracy of the simulation results when the other parameters are given “safe” values. This is illustrated in fig. 4.5, which displays results from simulations of 500 keV electrons in aluminimum (infinite medium). Electrons started off from the origin of coordinates moving in the direction of the z axis. During the generation of each electron track, we scored the energy deposited at different “depths” (z-coordinate) to get the “depth-dose” distribution. The simulation of a track was discontinued when the electron had travelled a path length \(s\) equal to 200 \(\mu\)m, and the PDFs of the final electron energy and position coordinates were tallied. Notice that no secondary radiation was followed and that the kinetic energy of the electrons at \(s = 200 \mu\)m was not included in the dose distribution (i.e. the calculated “dose” does not represent the quantity that would be obtained from a measurement).

The results displayed in fig. 4.5 are from equivalent detailed and mixed simulations with \(E_{\text{abs}} = 10\) keV and \(s_{\text{max}} = 40 \mu\)m. The detailed simulation was performed by setting \(C_1 = C_2 = 0, W_{cr} = 0\) and \(W_{cr} = -100\). Notice that when the user enters a negative value of the cutoff energy loss for radiative events, PENEOPE sets \(W_{cr} = 10\) eV, disregards the emission of soft bremsstrahlung photons with \(W < 10\) eV (which represents a negligible fraction of the stopping power) and simulates hard bremsstrahlung events as usually, i.e. in a detailed way. The mixed simulation results shown in fig. 4.5 were generated with \(C_1 = C_2 = 0.2, W_{cr} = 1\) keV and \(W_{cr} = -100\) (i.e. radiative events were described as in the detailed simulation).

In the detailed simulation, about 15 million electron tracks were generated by running a modified version of the code PENLAB.F (see section 6.2.1) on a 666 MHz PII computer for 85 hours, which corresponds to a simulation speed of 49 tracks/s. The average numbers of elastic, inelastic and bremsstrahlung interactions that had to be simulated to produce each detailed track were 1297 and 1222 and 0.03, respectively. On the same computer, the mixed simulation generated 25 million tracks in about 2 hours, which represents a simulation speed of 3421 tracks/s, 71 times faster than that of detailed simulation. The reason is that, on average, there were only 2.4 hard elastic collisions, 6.2 hard inelastic collisions, 0.03 hard bremsstrahlung events and 6.8 delta interactions along each track. From fig. 4.5 we conclude that, in this case, the mixed algorithm is not only stable under variations of the parameter \(C_1\) over the accepted range (0.0,2), but also provides results that are essentially equivalent to those from the detailed simulation. It is worth recalling that detailed simulation is nominally exact, the results are affected only by statistical uncertainties.

In general, our mixed simulation algorithm yields very accurate results (i.e. agree-
ing with those from detailed simulation) for electron and positron transport in infinite media, but not necessarily for limited geometries. The existence of interfaces poses considerable problems to condensed (class I) simulation, for which a satisfactory solution/approximation is not yet known. The present mixed (class II) algorithm handles interface crossing in a more accurate, but still approximate way. The rule to ensure accuracy for transport in the vicinity of interfaces is to use a small enough value of $s_{\text{max}}$. 
Chapter 5

Constructive quadric geometry

Practical simulations of radiation transport in material systems involve two different kinds of operations, namely, physical (determination of the path length to the next interaction, random sampling of the different interactions) and geometrical (space displacements, interface crossings, ...) operations. In the case of material systems with complex geometries, geometrical operations can take a large fraction of the simulation time. These operations are normally performed by dedicated subroutine packages, whose characteristics depend on the kind of algorithm used to simulate the interactions. The material system is assumed to consist of a number of homogeneous bodies limited by well-defined surfaces. The evolution of particles within each homogeneous body is dictated by the physical simulation routines, which operate as if particles were moving in an infinite medium with a given composition. Normally, the physical routines can handle a number of different media, whose interaction properties have been previously stored in memory. The job of the geometry routines is to steer the simulation of particle histories in the actual material system. They must determine the active medium, change it when the particle crosses an interface (i.e., a surface that separates two different media) and, for certain simulation algorithms, they must also keep control of the proximity of interfaces.

In this chapter, we describe the FORTRAN subroutine package PENGEOM, which is adequate for detailed simulation algorithms (i.e., algorithms where all single interactions in the history of a particle are simulated in chronological succession). With these algorithms, the description of interface crossings is very simple: when the particle reaches an interface, its track is stopped just after entering a new material body and restarted again with the new active medium. This method (stopping and restarting a track when it crosses an interface) is applicable even when we have the same medium on both sides of the surface. That is, detailed simulations with a single homogeneous body and with the same body split into two parts by an arbitrary surface yield the same results (apart from statistical uncertainties).

As we have seen, detailed simulation is feasible only for photon transport and low-energy electron transport. For high-energy electrons and positrons, most Monte Carlo codes [e.g., ETRAN (Berger and Seltzer, 1988), ITS3 (Halbleib et al., 1992), EGS4 (Nel-
son et al., 1985), EGSnrc (Kawrakow and Rogers, 2000), GEANT (Brun et al., 1986)] have recourse to condensed (class I) or mixed (class II) simulation, where the global effect of multiple interactions along a path segment of a given length is evaluated using available multiple scattering theories. To avoid large step lengths that could place the particle inside a different medium, these condensed procedures require the evaluation of the distance from the current position to the nearest interface, an operation with a high computational cost (see e.g. Bielajew, 1995). The mixed procedure implemented in PENELope is, at least computationally, analogous to detailed simulation (it gives a “jump-and-knock” description of particle tracks). In fact, the structure of PENELope’s tracking algorithm was designed to minimize the influence of the geometry on the transport physics. This algorithm operates independently of the proximity of interfaces and only requires knowledge of the material at the current position of the particle. As a consequence, the geometry package PENGEOM is directly linkable to PENELope. However, since PENGEOM does not evaluate the distance to the closest interface, it cannot be used with condensed simulation codes, such as those mentioned above.

Let us mention, in passing, that in simulations of high-energy photon transport complex geometries can be handled by means of relatively simple methods, which do not require control of interface crossings (see e.g. Snyder et al., 1969). Unfortunately, similar techniques are not applicable to electron and positron transport, mainly because these particles have much shorter track lengths and, hence, the transport process is strongly influenced by inhomogeneities of the medium. With the analogue simulation scheme adopted in PENELope, it is necessary to determine when a particle track crosses an interface, not only for electrons and positrons but also for photons.

PENGEOM evolved from a subroutine package of the same name provided with the 1996.02.29 version of the PENELope code system. This package was aimed at describing simple structures with a small number of homogeneous bodies limited by quadric surfaces. Although it was robust and very flexible, its speed deteriorated rapidly when the number of surfaces increased. The need for developing a more efficient geometry package became evident when we started to use PENELope to simulate radiation transport in accelerator heads (the description of which requires of the order of 100 surfaces) or in studies of total body irradiation (the definition of a realistic anthropomorphic phantom may involve a few hundred surfaces).

With PENGEOM we can describe any material system consisting of homogeneous bodies limited by quadric surfaces. To speed up the geometry operations, the bodies of the material system can be grouped into modules (connected volumes, limited by quadric surfaces, that contain one or several bodies); modules can in turn form part of larger modules, and so on. This hierarchic modular structure allows a reduction of the work of the geometry routines, which becomes more effective when the complexity of the system increases.

Except for trivial cases, the correctness of the geometry definition is difficult to check and, moreover, 3D structures with interpenetrating bodies are difficult to visualize. A pair of programs, named GVIEW2D and GVIEW3D, have been written to display the ge-
ometry on the computer screen. These programs use specific computer graphics software and, therefore, they are not portable. The executable files included in the PENelope distribution package run on IBM-compatible personal computers under Microsoft Windows 9x; they are simple and effective tools for debugging the geometry definition file.

5.1 Rotations and translations

The definition of various parts of the material system (quadric surfaces in reduced form and modules) involves rotations and translations. To describe these transformations, we shall adopt the active point of view: the reference frame remains fixed and only the space points (vectors) are translated or rotated.

In what follows, and in the computer programs, all lengths are in cm. The position and direction of movement of a particle are referred to the laboratory coordinate system, a Cartesian reference frame which is defined by the position of its origin of coordinates and the unit vectors \( \hat{x} = (1, 0, 0) \), \( \hat{y} = (0, 1, 0) \) and \( \hat{z} = (0, 0, 1) \) along the directions of its axes.

A translation \( T(t) \), defined by the displacement vector \( t = (t_x, t_y, t_z) \), transforms the vector \( r = (x, y, z) \) into

\[
T(t) r = r + t = (x + t_x, y + t_y, z + t_z).
\] (5.1)

Evidently, the inverse translation \( T^{-1}(t) \) corresponds to the displacement vector \(-t\), i.e. \( T^{-1}(t) = T(-t) \).

A rotation \( R \) is defined through the Euler angles \( \omega, \theta \) and \( \phi \), which specify a sequence of rotations about the coordinate axes\(^1\): first a rotation of angle \( \omega \) about the \( z \)-axis, followed by a rotation of angle \( \theta \) about the \( y \)-axis and, finally, a rotation of angle \( \phi \) about the \( z \)-axis. A positive rotation about a given axis would carry a right-handed screw in the positive direction along that axis. Positive (negative) angles define positive (negative) rotations.

The rotation \( R(\omega, \theta, \phi) \) transforms the vector \( r = (x, y, z) \) into a vector

\[
r' = R(\omega, \theta, \phi) r = (x', y', z'),
\] (5.2)

whose coordinates are given by

\[
\begin{pmatrix}
x' \\
y' \\
z'
\end{pmatrix} = R(\omega, \theta, \phi) \begin{pmatrix}
x \\
y \\
z
\end{pmatrix},
\] (5.3)

\(^1\)This definition of the Euler angles is the one usually adopted in Quantum Mechanics (see e.g. Edmonds, 1960).
where
\[
R(\omega, \theta, \phi) = \begin{pmatrix} R_{xx} & R_{xy} & R_{xz} \\ R_{yx} & R_{yy} & R_{yz} \\ R_{zx} & R_{zy} & R_{zz} \end{pmatrix}
\] (5.4)
is the rotation matrix. To obtain its explicit form, we recall that the matrices for rotations about the z- and y-axes are
\[
R_z(\phi) = \begin{pmatrix} \cos \phi & -\sin \phi & 0 \\ \sin \phi & \cos \phi & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad \text{and} \quad R_y(\theta) = \begin{pmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{pmatrix},
\]
respectively. Hence,
\[
R(\omega, \theta, \phi) = R_z(\phi)R_y(\theta)R_z(\omega)
= \frac{1}{\sin \phi} \begin{pmatrix} \cos \phi & -\sin \phi & 0 \\ \sin \phi & \cos \phi & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{pmatrix} \begin{pmatrix} \cos \omega & -\sin \omega & 0 \\ \sin \omega & \cos \omega & 0 \\ 0 & 0 & 1 \end{pmatrix},
\]
(5.6)
The inverse of the rotation \(R(\omega, \theta, \phi)\) is \(R(-\phi, -\theta, -\omega)\) and its matrix is the transpose of \(R(\omega, \theta, \phi)\), i.e.
\[
R^{-1}(\omega, \theta, \phi) = R(-\phi, -\theta, -\omega) = R_z(-\omega)R_y(-\theta)R_z(-\phi) = R^T(\omega, \theta, \phi).
\]
(5.7)

Let us now consider transformations \(C = T(t)R(\omega, \theta, \phi)\) that are products of a rotation \(R(\omega, \theta, \phi)\) and a translation \(T(t)\). \(C\) transforms a point \(r\) into
\[
r' = C(r) = T(t)R(\omega, \theta, \phi) r
\]
or, in matrix form,
\[
\begin{pmatrix} x' \\ y' \\ z' \end{pmatrix} = R(\omega, \theta, \phi) \begin{pmatrix} x \\ y \\ z \end{pmatrix} + \begin{pmatrix} t_x \\ t_y \\ t_z \end{pmatrix}.
\]
(5.9)
Notice that the order of the factors does matter; the product of the same factors in reverse order \(D = R(\omega, \theta, \phi) T(t)\) transforms \(r\) into a point \(r' = D(r)\) with coordinates
\[
\begin{pmatrix} x' \\ y' \\ z' \end{pmatrix} = R(\omega, \theta, \phi) \begin{pmatrix} x + t_x \\ y + t_y \\ z + t_z \end{pmatrix}.
\]
(5.10)
5.2. Quadric surfaces

Given a function $F(\mathbf{r})$, the equation $F(\mathbf{r}) = 0$ defines a surface in implicit form. We can generate a new surface by applying a rotation $\mathcal{R}(\omega, \theta, \phi)$ followed by a translation $\mathcal{T}(t)$ (we shall always adopt this order). The implicit equation of the transformed surface is

$$G(\mathbf{r}) = F \left[ \mathcal{R}^{-1}(\omega, \theta, \phi) \mathcal{T}^{-1}(t) \mathbf{r} \right] = 0,$$

which simply expresses the fact that $G(\mathbf{r})$ equals the value of the original function at the point $\mathbf{r}' = \mathcal{R}^{-1}(\omega, \theta, \phi) \mathcal{T}^{-1}(t) \mathbf{r}$ that transforms into $\mathbf{r}$.

5.2 Quadric surfaces

As already mentioned, the material system consists of a number of homogeneous bodies, defined by their composition (material) and limiting surfaces. For practical reasons, all limiting surfaces are assumed to be quadrics given by the implicit equation

$$F(x, y, z) = A_{xx}x^2 + A_{xy}xy + A_{xz}xz + A_{yy}y^2 + A_{yz}yz + A_{zz}z^2 + A_x x + A_y y + A_z z + A_0 = 0,$$

which includes planes, pairs of planes, spheres, cylinders, cones, ellipsoids, paraboloids, hyperboloids, etc. In practice, limiting surfaces are frequently known in “graphical” form and it may be very difficult to obtain the corresponding quadric parameters. Try with a simple example: calculate the parameters of a circular cylinder of radius $R$ such that its symmetry axis goes through the origin and is parallel to the vector $(1, 1, 1)$. To facilitate the definition of the geometry, each quadric surface can be specified either through its implicit equation or by means of its reduced form, which defines the “shape” of the surface (see fig. 5.1), and a few simple geometrical transformations.

A reduced quadric is defined by the expression

$$F_r(x, y, z) = I_1 x^2 + I_2 y^2 + I_3 z^2 + I_4 xz + I_5 = 0,$$

where the coefficients (indices) $I_1$ to $I_5$ can only take the values $-1$, $0$ or $1$. Notice that reduced quadrics have central symmetry about the $z$-axis, i.e. $F_r(-x, -y, z) = F_r(x, y, z)$. The possible (real) reduced quadrics are given in table 5.1.

A general quadric is obtained from the corresponding reduced form by applying the following transformations (in the quoted order)\(^2\).

(i) An expansion along the directions of the axes, defined by the scaling factors $X$-SCALE $= a$, $Y$-SCALE $= b$ and $Z$-SCALE $= c$. The equation of the scaled quadric is

$$F_r(x, y, z) = I_1 \left( \frac{x}{a} \right)^2 + I_2 \left( \frac{y}{b} \right)^2 + I_3 \left( \frac{z}{c} \right)^2 + I_4 \frac{x}{c} + I_5 = 0.$$  

\(^2\)Keywords used to denote the various parameters in the geometry definition file are written in typewriter font, e.g. X-SCALE. See section 5.4.
<table>
<thead>
<tr>
<th>Index</th>
<th>Surface Representation</th>
<th>Index</th>
<th>Surface Representation</th>
<th>Index</th>
<th>Surface Representation</th>
</tr>
</thead>
<tbody>
<tr>
<td>1,1,1,0,−1</td>
<td><img src="sphere.png" alt="Sphere" /></td>
<td>1,1,0,0,−1</td>
<td><img src="cylinder.png" alt="Cylinder" /></td>
<td>1,−1,0,0,−1</td>
<td><img src="quadric.png" alt="Quadric" /></td>
</tr>
<tr>
<td>1,1,−1,0,0</td>
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<td>1,1,−1,0,−1</td>
<td><img src="quadric2.png" alt="Quadric" /></td>
<td>1,1,−1,0,1</td>
<td><img src="quadric3.png" alt="Quadric" /></td>
</tr>
<tr>
<td>1,1,0,−1,0</td>
<td><img src="quadric4.png" alt="Quadric" /></td>
<td>1,0,0,−1,0</td>
<td><img src="quadric5.png" alt="Quadric" /></td>
<td>1,−1,0,−1,0</td>
<td><img src="quadric6.png" alt="Quadric" /></td>
</tr>
</tbody>
</table>

**Figure 5.1:** Non-planar reduced quadric surfaces and their indices [see eq. (5.13)]. In all cases, the perspective is the same as for the sphere.
Table 5.1: Reduced quadrics.

<table>
<thead>
<tr>
<th>Reduced form</th>
<th>Indices</th>
<th>Quadric</th>
</tr>
</thead>
<tbody>
<tr>
<td>$z - 1 = 0$</td>
<td>0 0 0 1 -1</td>
<td>plane</td>
</tr>
<tr>
<td>$z^2 - 1 = 0$</td>
<td>0 0 1 0 -1</td>
<td>pair of parallel planes</td>
</tr>
<tr>
<td>$x^2 + y^2 + z^2 - 1 = 0$</td>
<td>1 1 1 0 -1</td>
<td>sphere</td>
</tr>
<tr>
<td>$x^2 + y^2 - z^2 = 0$</td>
<td>1 1 0 0 0</td>
<td>cylinder</td>
</tr>
<tr>
<td>$x^2 - y^2 - 1 = 0$</td>
<td>1 -1 0 0 -1</td>
<td>hyperbolic cylinder</td>
</tr>
<tr>
<td>$x^2 + y^2 - z^2 = 0$</td>
<td>1 1 -1 0 0</td>
<td>cone</td>
</tr>
<tr>
<td>$x^2 + y^2 - z^2 - 1 = 0$</td>
<td>1 1 -1 0 -1</td>
<td>one sheet hyperboloid</td>
</tr>
<tr>
<td>$x^2 + y^2 - z^2 + 1 = 0$</td>
<td>1 1 0 0 1</td>
<td>two sheet hyperboloid</td>
</tr>
<tr>
<td>$x^2 + y^2 - z = 0$</td>
<td>1 0 0 -1 0</td>
<td>paraboloid</td>
</tr>
<tr>
<td>$x^2 - z = 0$</td>
<td>1 -1 0 -1 0</td>
<td>hyperbolic paraboloid</td>
</tr>
</tbody>
</table>

... and permutations of $x$, $y$ and $z$ that preserve the central symmetry with respect to the $z$-axis.

For instance, this transforms the reduced sphere into an ellipsoid with semiaxes equal to the scaling factors.

(ii) A rotation, $\mathcal{R}(\omega, \theta, \phi)$, defined through the Euler angles $\text{OMEGA} = \omega$, $\text{THETA} = \theta$ and $\text{PHI} = \phi$. Notice that the rotation $\mathcal{R}(\omega, \theta, \phi)$ transforms a plane perpendicular to the $z$-axis into a plane perpendicular to the direction with polar and azimuthal angles $\text{THETA}$ and $\text{PHI}$, respectively. The first Euler angle, $\omega$ has no effect when the initial (scaled) quadric is symmetric about the $z$-axis.

(iii) A translation, defined by the components of the displacement vector $t$ ($\text{X-SHIFT}= t_x$, $\text{Y-SHIFT}= t_y$, $\text{Z-SHIFT}= t_z$).

A quadric is completely specified by giving the set of indices $(I_1, I_2, I_3, I_4, I_5)$, the scale factors $(\text{X-SCALE, Y-SCALE, Z-SCALE})$, the Euler angles $(\text{OMEGA, THETA, PHI})$ and the displacement vector $(\text{X-SHIFT, Y-SHIFT, Z-SHIFT})$. Any quadric surface can be expressed in this way. The implicit equation of the quadric is obtained as follows. We define the matrix

$$
\mathcal{A} = \begin{pmatrix}
A_{xx} & \frac{1}{2}A_{xy} & \frac{1}{2}A_{xz} \\
\frac{1}{2}A_{xy} & A_{yy} & \frac{1}{2}A_{yz} \\
\frac{1}{2}A_{xz} & \frac{1}{2}A_{yz} & A_{zz}
\end{pmatrix}
$$

and write the generic quadric equation (5.12) in matrix form

$$
\mathbf{r}^T \mathcal{A} \mathbf{r} + \mathcal{A}^T \mathbf{r} + \mathbf{a}_0 = 0,
$$

(5.16)
where \( \mathbf{r} \) and \( \mathbf{A} \equiv (A_x, A_y, A_z) \) are considered here as one-column matrices. Notice that the matrix \( \mathbf{A} \) is symmetric (\( \mathbf{A}^T = \mathbf{A} \)). Expressing the scaled quadric (5.14) in the form (5.16), the equation for the rotated and shifted quadric is [see eq. (5.11)]

\[
(r - t)^T \mathbf{R} \mathbf{A} \mathbf{R}^T (r - t) + (\mathbf{R} \mathbf{A})^T (r - t) + A_0 = 0,
\]

which can be written in the generic form (5.16)

\[
r^T \mathbf{A}' \mathbf{r} + \mathbf{A}'^T \mathbf{r} + A'_0 = 0
\]

with

\[
\mathbf{A}' = \mathbf{R} \mathbf{A} \mathbf{R}^T, \quad \mathbf{A}' = \mathbf{R} \mathbf{A} - 2 \mathbf{A} \mathbf{t}, \quad A'_0 = A_0 + \mathbf{t}^T (\mathbf{A}' \mathbf{t} - \mathbf{R} \mathbf{A}).
\]

From these relations, the parameters of the implicit equation (5.12) are easily obtained.

A quadric surface \( F(x, y, z) = 0 \) divides the space into two exclusive regions that are identified by the sign of \( F(x, y, z) \), the surface side pointer. A point with coordinates \((x_0, y_0, z_0)\) is said to be inside the surface if \( F(x_0, y_0, z_0) \leq 0 \) (side pointer = -1), and outside it if \( F(x_0, y_0, z_0) > 0 \) (side pointer = +1).

### 5.3 Constructive quadric geometry

A body is defined as a space volume limited by quadric surfaces and filled with a homogeneous material. To specify a body we have to define its limiting quadric surfaces \( F(\mathbf{r}) = 0 \), with corresponding side pointers (+1 or -1), and its composition (i.e. the integer label used by PENEOPE to identify the material). It is considered that bodies are defined in “ascending”, exclusive order so that previously defined bodies effectively delimit the new ones. This is convenient e.g. to describe bodies with inclusions. The work of the geometry routines is much easier when bodies are completely defined by their limiting surfaces, but this is not always possible or convenient for the user. The example in section 5.7 describes an arrow inside a sphere (fig. 5.2); the arrow is defined first so that it limits the volume filled by the material inside the sphere. It is impossible to define the hollow sphere (as a single body) by means of only its limiting quadric surfaces. It is clear that, by defining a conveniently large number of surfaces and bodies, we can describe any quadric geometry.

The subroutine package PENGEOM contains a subroutine, named \texttt{LOCATE}, that “locates” a point \( \mathbf{r} \), i.e. determines the body that contains it, if any. The obvious method is to compute the side pointers [i.e. the sign of \( F(\mathbf{r}) \)] for all surfaces and, then, explore the bodies in ascending order looking for the first one that fits the given side pointers. This brute force procedure was used in older versions of PENGEOM; it has the advantage of being robust (and easy to program) but becomes too slow for complex systems. A second subroutine, named \texttt{STEP}, “moves” the particle from a given position \( \mathbf{r}_0 \) within a body \( B \) a certain distance \( s \) in a given direction \( \mathbf{d} \). \texttt{STEP} also checks if the particle leaves the active medium and, when this occurs, stops the particle just after entering
the new material. To do this, we must determine the intersections of the track segment \( \mathbf{r}_0 + t\mathbf{d} \) (\( 0 < t \leq s \)) with all the surfaces that limit the body \( B \) (including those that limit other bodies that limit \( B \)), and check if the final position \( \mathbf{r}_0 + s\mathbf{d} \) remains in \( B \) or not. The reason for using only quadric surfaces is that these intersections are easily calculated by solving a quadratic equation.

Notice that bodies can be concave, i.e., the straight segment joining any two points in a body may not be wholly contained in the body. Hence, even when the final position of the particle lies within the initial body, we must analyze all the intersections of the path segment with the limiting surfaces of \( B \) and check if the particle has left the body after any of the intersections. When the particle leaves the initial body, say after travelling a distance \( s' \) (\( < s \)), we have to locate the point \( \mathbf{r}' = \mathbf{r}_0 + s'\mathbf{d} \). The easiest method consists of computing the side pointers of all surfaces of the system at \( \mathbf{r}' \), and determining the body \( B' \) that contains \( \mathbf{r}' \) by analyzing the side pointers of the different bodies in ascending order. It is clear that, for complex geometries, this is a very slow process. We can speed it up by simply disregarding those elements of the system that cannot be reached in a single step (e.g. bodies that are “screened” by other bodies). Unfortunately, as a body can be limited by all the other bodies that have been defined previously, the algorithm can be improved only at the expense of providing it with additional information. We shall adopt a simple strategy that consists of lumping groups of bodies together to form modules.

A module is defined as a connected volume\(^3\), limited by quadric surfaces, that contains one or several bodies. A module can contain other modules, which will be referred to as submodules of the first. The volume of a module is filled with a homogeneous medium, which automatically fills the cavities of the module (i.e. volumes that do not

\(^3\)A space volume is said to be connected when any two points in the volume can be joined by an arc of curve that is completely contained within the volume.
correspond to a body or to a submodule); these filled cavities are considered as a single
new body. A body that is connected and limited only by surfaces can be declared either
as a body or as a module. For the sake of simplicity, modules are required to satisfy
the following conditions: 1) the bodies and submodules of a module must be completely
contained within the parent module (i.e. it is not allowed to have portions of bodies or
submodules that lie outside the module) and 2) a submodule of a module cannot overlap
with other submodules and bodies of the same module (this is necessary to make sure
that a particle can only enter or leave a module through its limiting surfaces). Notice
however, that the bodies of a module are still assumed to be defined in ascending order,
i.e. a body is limited by its surfaces and by the previously defined bodies of the same module,
so that inclusions and interpenetrating bodies can be easily defined. Of course,
overlapping bodies must be in the same module.

A module (with its possible submodules) can represent a rigid part (e.g. a radioactive
source, an accelerator head, a detector, a phantom, etc.) of a more complex material
system. To facilitate the definition of the geometry, it is useful to allow free translations
and rotations of the individual modules. The definition of a module (see below) includes
the parameters of a rotation $R(\omega, \theta, \phi)$ and a translation $T(t)$, which are optional and
serve to modify the position and orientation of the module (and its submodules) with
respect to the laboratory reference frame. As before, the rotation is applied first. All
submodules and bodies of the same module are shifted and rotated together.

In practical simulations, it may be useful to limit the region of space where particles
have to be transported. For instance, to simulate the response of a detector with a given
photon source, it is advisable to stop the simulation of a particle when it is far enough
from the detector. This can be done automatically by considering an “enclosure” of
the material system, which is defined as a module that contains the complete system.
If such a covering module is not explicitly defined, the subroutines set the enclosure
as a sphere of $10^{15}$ cm radius. It is assumed that there is perfect vacuum outside the
enclosure, and in any inner volume that is not a part of a body or of a filled module.
Hence, particles that leave the enclosure are lost and will never return to the material
system.

For programming purposes, it is useful to imagine each module as the mother of its
bodies and submodules, and as the daughter of the module that contains it. We thus
have a kind of genealogical tree with various generations of modules and bodies (see fig.
5.3). The first generation reduces to the enclosure (which is the only motherless module).
The members of the second generation are bodies and modules that are daughters of
the enclosure. The $n$-th generation consists of modules and bodies whose mothers
belong to the $(n - 1)$-th generation. Each module is defined by its limiting surfaces
(which determine the border with the external world) and those of their descendants
(which determine the module’s internal structure); this is not true for bodies (childless
members of the tree), which can be limited either by surfaces, by other sister bodies or
by a combination of both. A body that is limited only by surfaces can be defined as a
module, which has the advantage of allowing free rotation and translation.
5.4 Geometry definition file

The geometry is defined from the input file (UNIT=IRD). In principle, this permits the simulation of different geometries by using the same main program. The input file consists of a series of data sets, which define the different elements (surfaces, bodies and modules). A data set consists of a number of strictly formatted text lines; it starts and ends with a separation line filled with zeros. The first line after each separation line must start with one of the defining 8-character strings “SURFACE-”, “BODY----”, “MODULE--”, “END-----” or “INCLUDE-” (here, blank characters are denoted by “-”; they are essential!). Informative text (as many lines as desired) can be written at the beginning of the file, before the first separation line. A line starting with the string “END-----” after a separation line discontinues the reading of geometry data. Each element is identified by its type (surface, body or module) and a three-digit integer label. Although the element label can be given an arbitrary value (-99 to 999) in the input file, PANGEOM redefines it so that elements of a given kind are numbered consecutively, according to their input order. Notice that bodies and modules are considered as elements of the same kind (i.e. assigning the same label to a body and to a module will cause an error of the reading routine).

In the input file, numerical quantities must be written within the parentheses in the specified format. All lengths are in cm; angles can be given in either degrees (DEG) or radians (RAD). When angles are in degrees, it is not necessary to specify the unit. The parameters in each data set can be entered in any order. They can even be defined several times, in which case, only the last input value is accepted. This is useful, e.g.
to study variations caused by changing these parameters without duplicating the input file. Comments can be written at the end of each line, at the right of the last keyword or after the closing parenthesis of numerical fields.

- The format of the data set of a surface defined in reduced form is the following,

```
SURFACE ( I3 ) TEXT DESCRIBING THE SURFACE ...
INDICES=(I2,I2,I2,I2,I2,I2)
X-SCALE=( E22.15 , I3 ) (DEFAULT=1.0)
Y-SCALE=( E22.15 , I3 ) (DEFAULT=1.0)
Z-SCALE=( E22.15 , I3 ) (DEFAULT=1.0)
OMEGA=( E22.15 , I3 ) (DEFAULT=0.0)
THETA=( E22.15 , I3 ) (DEFAULT=0.0)
PHI=( E22.15 , I3 ) (DEFAULT=0.0)
X-SHIFT=( E22.15 , I3 ) (DEFAULT=0.0)
Y-SHIFT=( E22.15 , I3 ) (DEFAULT=0.0)
Z-SHIFT=( E22.15 , I3 ) (DEFAULT=0.0)
```

- Surface parameters are optional and can be entered in any order. Default values are assigned to parameters not defined in the input file. Thus, to define an elliptic cylinder centred on the z-axis, only the parameters X-SCALE and Y-SCALE are required. Notice that scale parameters must be greater than zero.

- The I3 value following each parameter must be set equal to zero (or negative) to make the parameter value effective. When this field contains a positive integer IP, the parameter is set equal to the value stored in the IP-th component of the array PARINP, an input argument of subroutine GEOMIN (see section 5.6). This permits the user to modify the geometry parameters from the MAIN program.

- Limiting surfaces can also be defined in implicit form. When a quadric surface is defined in this way, the indices must be set to zero; this switches the reading subroutine GEOMIN to implicit mode. The format of an implicit surface data set is

```
SURFACE ( I3 ) TEXT DESCRIBING THE SURFACE ...
INDICES=( 0, 0, 0, 0, 0, 0)
AXX=( E22.15 , I3 ) (DEFAULT=0.0)
AYY=( E22.15 , I3 ) (DEFAULT=0.0)
AZZ=( E22.15 , I3 ) (DEFAULT=0.0)
AX=( E22.15 , I3 ) (DEFAULT=0.0)
AY=( E22.15 , I3 ) (DEFAULT=0.0)
AZ=( E22.15 , I3 ) (DEFAULT=0.0)
A0=( E22.15 , I3 ) (DEFAULT=0.0)
```

- Surface parameters are optional and can be entered in any order. The default value 0.0 is assigned to parameters not defined in the input file.

- The I3 value following each parameter must be negative or zero to make the parameter value effective. Otherwise, its actual value will be set by subroutine GEOMIN.
5.4. Geometry definition file

- The format of a body data set is

```
0000000000000000000000000000000000000000000000000000000000000000
BODY ( I3 ) TEXT DESCRIBING THE BODY ...
MATERIAL ( I3 )
SURFACE ( I3 ), SIDE POINTER=(I2)
SURFACE ( I3 ), SIDE POINTER=(I2) ...
BODY ( I3 )
BODY ( I3 ) ...
```

- The indicator of each material (2nd line) must agree with the convention adopted in PENELope. Void inner volumes can be described as material bodies with MATERIAL set to 0.
- A line is required to define each limiting surface, with its side pointer, and each limiting body. Limiting surfaces and bodies can be entered in any order.
- Bodies are assumed to be defined in ascending order so that, in principle, it would not be necessary to declare the limiting bodies. However, to speed up the calculations, it is required to declare explicitly all the elements (surfaces and bodies) that actually limit the body that is being defined. Omission of a limiting body will cause inconsistencies unless the materials in the limiting and the limited bodies are the same.

- The format for the definition of a module is the following:

```
0000000000000000000000000000000000000000000000000000000000000000
MODULE ( I3 ) TEXT DESCRIBING THE MODULE...
MATERIAL ( I3 )
SURFACE ( I3 ), SIDE POINTER=(I2)
SURFACE ( I3 ), SIDE POINTER=(I2) ...
BODY ( I3 )
BODY ( I3 ) ...
MODULE ( I3 )
MODULE ( I3 ) ...
```

- The material (which must be explicitly declared) fills the cavities of the module.
- As in the case of bodies, MATERIAL = 0 corresponds to vacuum.
- The limiting surfaces must define a connected volume. All inner bodies and modules (submodules) must be declared. Notice that these cannot extend outside the module’s volume and that a submodule cannot overlap with the other submodules and bodies.
- Limiting surfaces, inner bodies and submodules can be entered in any order.
- The enclosure of the system can be defined as a module (the last in the geometry file). The enclosure must contain all modules and all bodies that do not belong to modules. Otherwise, the subroutines will define a new enclosure (a sphere with
$10^{15}$ cm radius) that is supposed to satisfy this condition. Make sure that the system does not contain bodies that extend outside the enclosure (this normally will cause errors).

- The rotation and the translation are optional and apply to all elements of the module. The line filled with 1’s ends the definition of elements and starts that of transformation parameters (it can be skipped if no transformation parameters are entered). The 13 value following each parameter has the same meaning as above. It has to be negative or zero to make the parameter value effective; otherwise, the parameter must be set from the main program.

A single surface can be used to define several bodies and/or submodules in the same module; unnecessary duplication of a surface reduces the calculation speed. Notice, however, that rotation or translation of a module modifies all the surfaces of its descendants and, therefore, a transformed surface must be redefined to be used again. Thus, if the system contains two identical modules in different positions (e.g. two detectors in a coincidence experiment), each of them must be defined explicitly. This does not require too much editing work; after generating the first of the two modules, we can just duplicate its definition data sets and change their labels.

The INCLUDE option allows inserting a predefined structure (e.g. a scintillation detector, an encapsulated nuclear source, ...) within the geometry file. The inserted structure is defined by a complete definition file (i.e. ending with an “END------” line). The labels of the objects in the included file must be different from the labels used in the main file and in any other included file. The format of an INCLUDE block is the following,

```
000000000000000000000000000000000000000000000000000
INCLUDE
FILE= (filename.ext)
000000000000000000000000000000000000000000000000000
```

The name of the included file must be written between the parentheses. It may be up to twelve characters long; if it is shorter, the blanks must be left at the right end of the field. Only one-level INCLUDES are allowed, i.e. an included file cannot contain any INCLUDE blocks.

The definition of the geometry may seem somewhat more laborious than with combinatorial methods, where the system is described by combining basic bodies of several simple shapes [see Jenkins et al. (1988) and references therein]. In practice, however, defining the various surfaces that limit a body may be more convenient, and intuitive, than considering all the parameters needed to specify that body. The example of a right elliptical cylinder, which needs 9 parameters, is quite illustrative. With our method, this body can be defined as a module by means of two planes perpendicular to the z-axis (only one parameter if the base is the $z = 0$ plane) and a scaled cylinder (2 parameters); the rotation (3 parameters) of the module gives the required orientation and the translation (3 parameters) puts it in the required position. The definition as a proper body requires defining the three surfaces that limit the cylinder in its actual position,
which is a bit more inconvenient. In any case, the important issue is not how to define
the geometry, but the amount of computation needed to follow a particle through the
material system.

5.5 The subroutine package PENGEO

The package PENGEO consists of the following subroutines:

- **SUBROUTINE GEOMIN(PARINP,NPINP,NMAT,NBOD,IRD,IWR)**
  - Reads geometry data from the input file and initializes the geometry package.
  - Input arguments:
    - **PARINP**: Array containing optional parameters, which may replace the ones entered
      from the input file. This array must be declared in the **MAIN** program, even when
      **NPINP = 0**.
    - **NPINP**: Number of parameters defined in **PARINP** (positive).
    - **IRD**: Input file unit (opened in the main program).
    - **IWR**: Output file unit (opened in the main program).
  - Output arguments:
    - **NMA T**: Number of different materials in full bodies (excluding void regions).
    - **NBOD**: Number of defined bodies and modules.

Subroutine **GEOMIN** labels elements of the various kinds (surfaces, bodies and mod-
ules) in strictly increasing order; it may also redefine some of the geometry param-
eters, whose actual values are entered through the array **PARINP**. A copy of the
geometry definition file, with the effective parameter values and with the element
labels assigned by **GEOMIN**, is printed on the output file (**UNIT IWR**). This part of
the output file describes the actual geometry used in the simulation.

- **SUBROUTINE LOCATE**
  - Determines the body that contains the point with coordinates (**X**, **Y**, **Z**).
  - Input values (through **COMMON/TRACK/)**:
    - **X**, **Y**, **Z**: Particle position coordinates.
    - **U**, **V**, **W**: Direction cosines of the direction of movement.
  - Output values (through **COMMON/TRACK/)**:
    - **IBODY**: Body where the particle moves.
    - **MAT**: Material in **IBODY**. The output **MAT = 0** indicates that the particle is in a
      void region.

- **SUBROUTINE STEP(DS,DSEF,NCROSS)**
  - Used in conjunction with **PENelope**, this subroutine performs the geometrical part
    of the track simulation. The particle starts from the point (**X,Y,Z**) and proceeds
to travel a length **DS** in the direction (**U,V,W**) within the material where it moves.

---

*Most of the input/output of the geometry routines is through **COMMON/TRACK/**, which is the common
block used by **PENelope** to transfer particle state variables.*
STEP displaces the particle and stops it at the end of the step, or just after entering a new material. The output value DSEF is the distance travelled within the initial material. If the particle enters a void region, STEP continues the particle track, as a straight segment, until it penetrates a material body or leaves the system (the path length through inner void regions is not included in DSEF). When the particle arrives from a void region (MAT = 0), it is stopped after entering the first material body. The output value MAT = 0 indicates that the particle has escaped from the system.

- Input-output values (through COMMON/TRACK/):
  - X, Y, Z: Input: coordinates of the initial position.
  - Output: coordinates of the final position.
  - U, V, W: Direction cosines of the displacement. They are kept unaltered.
  - IBODY: Input: initial body, i.e. the one that contains the initial position.
  - Output: final body.
  - MAT: Material in body IBODY (automatically changed when the particle crosses an interface).

- Input argument:
  - DS: Distance to travel (unaltered).

- Output arguments:
  - DSEF: Travelled path length before leaving the initial material or completing the jump (less than DS if the track crosses an interface).
  - NCROSS: Number of interface crossings (=0 if the particle does not leave the initial material, greater than 0 if the particle enters a new material).

For the handling and storage of geometric information we take advantage of the structure of the genealogical tree. It is assumed that an enclosure has been defined so that it is the only common ancestor for all bodies and modules. To understand the operation of the geometry routines, it is convenient to define a matrix FLAG(KB, KS) as follows (the indices KS and KB indicate the label of a surface and a body or module, respectively),

\[ \text{FLAG}(KB, KS) = \begin{cases} 
1, & \text{if } KS \text{ is a limiting surface of } KB \text{ and } KB \text{ is inside } KS, \\
2, & \text{if } KS \text{ is a limiting surface of } KB \text{ and } KB \text{ is outside } KS, \\
3, & \text{if } KB \text{ is a body and } KS \text{ does not directly limit } KB, \text{ but appears in the definition of a body that limits } KB, \\
4, & \text{if } KB \text{ is a module and } KS \text{ limits one of its daughters (bodies and submodules), but does not appear in the definition of } KB, \\
5, & \text{otherwise.} 
\end{cases} \]

To locate a point we call subroutine LOCATE, where we proceed upwards in the genealogical tree of modules. If the point is outside the enclosure, we set MAT = 0 and return to the main program. Otherwise, we look for a module or body of the second generation that contains the point. If it exists, we continue analyzing its descendants.
(if any) and so on. The process ends when we have determined the body \texttt{IBODY} that contains the point, or as soon as we conclude that the point is outside the material system (i.e. in a void region). Notice that, when we have found that a module \texttt{KS} does contain the point, to do the next step we only need to consider the surfaces \texttt{KS} such that \texttt{FLAG(KB, KS)} = 1, 2 or 4.

After the body \texttt{IBODY} that contains the initial position of the particle has been identified, we can call subroutine \texttt{STEP} to move the particle a certain distance \texttt{DS}, dictated by \texttt{PENELOPE}, along the direction \texttt{(U,V,W)}. We start by checking whether the track segment crosses any of the surfaces that limit \texttt{IBODY}. If after travelling the distance \texttt{DS} the particle remains within the same body, \texttt{DSEP} is set equal to \texttt{DS} and control is returned to the main program. It is worth noting that the surfaces \texttt{KS} that define the initial body are those with \texttt{FLAG(IBODY, KS)}=1 and 2 (proper limiting surfaces) or =3 (limiting surfaces of limiting bodies). Although it may happen that a surface with \texttt{FLAG}=3 does not directly limit the body, subroutine \texttt{STEP} cannot know this from the information at hand and, consequently, all surfaces with \texttt{FLAG}=3 are analyzed after each move. It is clear that, to reduce the number of surfaces to be considered, we should minimize the number of bodies used to delimit other bodies.

When the particle leaves \texttt{IBODY} and enters a new material, \texttt{STEP} stops it just after crossing the interface and determines the new body and material (in this case, the output values of \texttt{IBODY} and \texttt{MAT} are different from the input ones). To do this, the limiting surfaces of the parent module and of all the sisters of the initial body must be analyzed (if they exist). If the new position is outside the parent module, we must analyze all surfaces that limit the parent’s sisters and go downward in the genealogical tree to determine the module that contains the point and, if necessary, go upwards again to find out what the new body is. If the new material is the same as in the initial body, the particle is allowed to move the remaining distance. Void regions (strict vacuum) are crossed freely (i.e. the distance travelled within these regions is not counted). Furthermore, when the particle starts from outside the enclosure, it is allowed to propagate freely until it reaches a material body. The particle is stopped when it penetrates a different material or when it leaves the system (i.e. when, after leaving a material body, its straight trajectory does not intersect a non-void body; in this case, the value \texttt{MAT}=0 is returned). Evidently, the speed of the geometry subroutines depends greatly on the structure of the modules’ genealogical tree. The responsibility of optimizing it rests with the user.

When \texttt{STEP} moves the particle across an interface, there is a risk that, owing to numerical truncation errors, the particle is placed on the wrong side of the interface (i.e. the track is stopped just before the interface). If this occurs, the program could go into an endless loop in which \texttt{STEP} repeatedly tries to move the particle a very small distance (of the order of $10^{-15}$ cm) towards the interface but does not succeed, i.e. the particle is trapped at the interface. To avoid this collapse of the trajectory, after each interface crossing, \texttt{STEP} applies an additional small displacement ($\sim 10^{-8}$ cm) in the direction of movement, which is physically irrelevant and sufficient to compensate for the effect of truncation errors. The same strategy is used in subroutine \texttt{LOCATE}: when the particle is too close to an interface, it is moved $10^{-8}$ cm along the surface gradient direction.
or its opposite, depending on whether the particle approaches or leaves the interface. Notice that this strategy requires that the direction of movement \((U, V, W)\) be defined before calling LOCATE. The extra displacement effectively eliminates the risk of particle trapping at interfaces; but it also sets a limit to the space resolution (geometrical details that are less than \(\sim 10\ \text{Å}\) in size cannot be described).

PENGEOM admits up to 250 surfaces and 125 bodies and modules. When the input file contains a larger number of elements, the program stops and a corresponding error message is printed. To describe such complex material systems, it is necessary to edit the source file PENGEOM.F and increase the values of the parameters NS (maximum number of surfaces) and NB (maximum number of bodies) in all subroutines. It is assumed that the number of bodies in a module is less than \(\mathbf{NX} = 100\), which is also the upper limit for the number of surfaces that can be used to define a body or a module (those with \(\mathbf{FLAG} < 5\)). When \(\mathbf{NX}\) is too small, the module that causes the trouble should be decomposed into several submodules. Although it is possible to increase the parameter \(\mathbf{NX}\), this would waste a lot of memory. As a consequence, a system with more than 100 surfaces or bodies must be decomposed into modules.

### 5.6 Debugging and viewing the geometry

A pair of computer programs named GVIE\(\text{W}2\)D and GVIE\(\text{W}3\)D have been written to visualize the geometry and to help the user to debug the definition file. These codes generate two- and three-dimensional 24-bit colour images of the system using specific graphics routines. The executable codes included in the distribution package run on personal computers under Microsoft Windows 9x.

The most characteristic (and useful) feature of GVIE\(\text{W}2\)D is that displayed pictures are generated by using the PENGEOM package and, therefore, errors and inconsistencies in the geometry definition file that would affect the results of actual simulations are readily identified. The method to generate the image consists of following a particle that moves on a plane perpendicular to an axis of the reference frame, which is mapped on the window. The particle starts from a position that corresponds to the left-most pixel and moves along a straight trajectory to the right of the window. To do this, we call subroutine \texttt{STEP} repeatedly, maintaining the direction of movement and with a large value of \(\mathbf{DS}\) (such that each body is crossed in a single step). A colour code is assigned to each material, and pixels are lit up with the active colour when they are crossed by the particle trajectory. The active colour is changed when the particle enters a new material. The final picture is a map of the bodies and materials intersected by the window plane. The orientation of the window plane, as well as the position and size of the window view, may be changed interactively by entering one of the one-character commands shown in table 5.2, directly from the graphics window (upper- and lower-case letters may work differently). With GVIE\(\text{W}2\)D we can inspect the internal structure of the system with arbitrary magnification (limited only by the intrinsic resolution of the PENGEOM routines).
Table 5.2: One-character commands of the GVIEV2D geometry viewer.

<table>
<thead>
<tr>
<th>Command</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>+</td>
<td>change window orientation, x-axis, +</td>
</tr>
<tr>
<td>y</td>
<td>change window orientation, y-axis, +</td>
</tr>
<tr>
<td>z</td>
<td>change window orientation, z-axis, +</td>
</tr>
<tr>
<td>r</td>
<td>shift right, l, left --&gt; shift left, +</td>
</tr>
<tr>
<td>u, up</td>
<td>shift up, d, down --&gt; shift down, +</td>
</tr>
<tr>
<td>f, pgup</td>
<td>shift front, b, pgdn --&gt; shift back, +</td>
</tr>
<tr>
<td>i, *</td>
<td>zoom in, o, - --&gt; zoom out, +</td>
</tr>
<tr>
<td>h</td>
<td>actual size, h, ? --&gt; help, +</td>
</tr>
<tr>
<td>blank, enter</td>
<td>repeat last command, q --&gt; quit. +</td>
</tr>
</tbody>
</table>

When running the GVIEV2D program, you will be asked to give the path+name of the geometry definition file and the coordinates \((X_C, Y_C, Z_C)\) of the centre of the window (relative to the laboratory frame) in \(\text{cm}\). The window may appear black (the colour for void regions) if no material bodies are intersected. In this case, use the one-character viewer commands to reach the bodies or, more conveniently, start again and place the window centre near or within a filled body.

GVIEV3D generates three-dimensional pictures of the geometry by using a simple ray-tracing algorithm, with the source light and the camera at the same position. Bodies are displayed with the same colour code used by GVIEV2D and the intensity of each pixel is determined by the angle between the vision line and the normal to the limiting surface. This method does not produce shadows and disregards light diffusion, but makes fairly realistic three-dimensional images. The camera is assumed to be outside the system (placing the camera inside a body would stop the program). To reveal the inner structure of the system, the program can eliminate a wedge (limited by two vertical planes that intersect in the \(z\)-axis). The position and size of the system can be modified by means of one-character commands entered from the graphics window. The command keys and actions are similar to those of GVIEV2D. It is worth noting that GVIEV3D generates the image pixel by pixel, whereas GVIEV2D does it by drawing straight lines on the window; as a result, GVIEV2D is much faster.

GVIEV2D and GVIEV3D produce an output file named \(\text{GEOMETRY.REP}\) (which is generated by subroutine \(\text{GEOMIN}\)) in the working directory. The programs are stopped either when an input format is incorrect (reading error) or when a clear inconsistency in the definition file is found (e.g., when the element that is being defined and the furnished information do not match). The wrong datum appears in the last printed lines of the \(\text{GEOMETRY.REP}\) file, usually in the last one. Error messages are also written on that file, so that the identification of inconsistencies is normally very easy. When the structure of the input file is correct, the codes do not stop and the geometry is displayed for further analysis. Most of the possible errors in the input file can only be revealed by direct inspection of the images generated by GVIEV2D and GVIEV3D.

The file \(\text{GEOMETRY.REP}\) is a duplicate of the input definition file. The only differ-
ences between the two files are the labels assigned to the different surfaces, bodies and modules; in \texttt{GEOMETRY.REP}, these elements are numbered in strictly increasing order. It is important to bear in mind that PENGEOM internally uses this sequential labelling to identify bodies and surfaces. Knowing the internal label assigned to each element is necessary for scoring purposes, e.g. to determine the distribution of energy deposited within a particular body.

5.7 A short tutorial

To prepare a new geometry definition file, it is useful to start from a file that contains a model of each data set with default values of their parameters. Placing the end-line at the beginning of the model group discontinues the geometry reading; so that the model group can be kept in the geometry file, even when this one is operative. The starting file should look like this

\begin{verbatim}
END   0000000000000000000000000000000000000000000000000000000000000000
0000000000000000000000000000000000000000000000000000000000000000
SURFACE ( ) REDUCED FORM
INDICES= (1, 1, 1, 1, 1)
X-SCALE= (+1.00000000000000E+00, 0) (DEFAULT=1.0)
Y-SCALE= (+1.00000000000000E+00, 0) (DEFAULT=1.0)
Z-SCALE= (+1.00000000000000E+00, 0) (DEFAULT=1.0)
OMEGA= (+0.00000000000000E+00, 0) DEG (DEFAULT=0.0)
THETA= (+0.00000000000000E+00, 0) DEG (DEFAULT=0.0)
PHI= (+0.00000000000000E+00, 0) RAD (DEFAULT=0.0)
X-SHIFT= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
Y-SHIFT= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
Z-SHIFT= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
0000000000000000000000000000000000000000000000000000000000000000
SURFACE ( ) IMPLICIT FORM
INDICES= (0, 0, 0, 0, 0)
AXX= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AXY= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AXZ= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AYX= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AYY= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AYZ= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AZX= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AZY= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AZZ= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AX= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AY= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AZ= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
AO= (+0.00000000000000E+00, 0) (DEFAULT=0.0)
0000000000000000000000000000000000000000000000000000000000000000
BODY ( ) TEXT
MATERIAL ( )
SURFACE ( ), SIDE POINTER= (1)
BODY ( )
0000000000000000000000000000000000000000000000000000000000000000
MODULE ( ) TEXT
MATERIAL ( )
SURFACE ( ), SIDE POINTER= (1)
BODY ( )
\end{verbatim}
5.7. A short tutorial

MODULE ( )

OMEGA=(+0.0000000000000000E+00, 0) DEG (DEFAULT=0.0)
THETA=(+0.0000000000000000E+00, 0) DEG (DEFAULT=0.0)
PHI=(+0.0000000000000000E+00, 0) RAD (DEFAULT=0.0)
X-SHIFT=(+0.0000000000000000E+00, 0) (DEFAULT=0.0)
Y-SHIFT=(+0.0000000000000000E+00, 0) (DEFAULT=0.0)
Z-SHIFT=(+0.0000000000000000E+00, 0) (DEFAULT=0.0)

Then, to generate a new element, we just duplicate the corresponding data set, modify
the parameter values and eliminate the lines that are unnecessary (i.e. those of param-
eters that take their default values). Of course, the defining data set must be placed
before the end-line. The progressing geometry can be visualized with GVIE2D as
soon as the first complete body has been defined. If GVIE2D stops before entering
the graphics mode, the geometry definition is incorrect and we should have a look at
the GEOMETRY.REP file to identify the problem. Normally, the conflicting parameter or
element appears in the last line of this file.

The basic elements of the geometry definition are quadric surfaces. These can be
visualized by using the following simple file, which defines the inner volume of a reduced
quadric as a single body,

Visualization of reduced quadric surfaces.
Define the desired quadric (surface 1) by entering its indices.
The region with side pointer -1 (inside the quadric) corresponds
to MATERIAL=1.

SURFACE ( 1) Reduced quadric. One sheet hyperboloid.
INDICES=( 1, 1,-1, 0,-1)
BODY ( 1) The interior of the quadric.
MATERIAL( 1)

SURFACE ( 1), SIDE POINTER=(-1)
END

Notice that, in this case, the body is infinite in extent. There is no objection to using
infinite bodies, as long as the enclosure contains all material bodies. When only a single
body is defined, PENGEOFINAL identifies it as the enclosure, and this requirement is met.
Otherwise, we must define a proper enclosure (since all bodies and modules must have
a common ancestor).

The following example describes a sphere with an inner arrow (fig. 5.2):

XX.XXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXX

Sphere of 5 cm radius with an arrow.
SURFACE ( 1) PLANE Z=4.25
INDICES=( 0, 0, 0, 1,-1)
Z-SCALE=( 4.2500000000000000E+00, 0)
SURFACE ( 2) PLANE Z=1.5
INDICES=( 0, 0, 0, 1, -1)
Z-SCALE=( 1.500000000000000000E+00, 0)
0000000000000000000000000000000000000000000000000000000000000000
SURFACE ( 3) PLAN E Z=-4.0
INDICES=( 0, 0, 0, 1, 1)
Z-SCALE=( 4.000000000000000000E+00, 0)
0000000000000000000000000000000000000000000000000000000000000000
SURFACE ( 4) CONE
INDICES=( 1, 1, -1, 0, 0)
X-SCALE=( 5.000000000000000000E-01, 0)
Y-SCALE=( 5.000000000000000000E-01, 0)
Z-SHIFT=( 4.250000000000000000E+00, 0)
0000000000000000000000000000000000000000000000000000000000000000
SURFACE ( 5) CYLINDER
INDICES=( 1, 1, 0, 0, -1)
X-SCALE=( 7.250000000000000000E-01, 0)
Y-SCALE=( 7.250000000000000000E-01, 0)
0000000000000000000000000000000000000000000000000000000000000000
BODY ( 1) ARROW HEAD
MATERIAL( 2)
SURFACE ( 1), SIDE POINTER=( -1)
SURFACE ( 2), SIDE POINTER=( 1)
SURFACE ( 4), SIDE POINTER=( -1)
0000000000000000000000000000000000000000000000000000000000000000
BODY ( 2) ARROW STICK
MATERIAL( 2)
SURFACE ( 5), SIDE POINTER=( -1)
SURFACE ( 2), SIDE POINTER=( -1)
SURFACE ( 3), SIDE POINTER=( 1)
0000000000000000000000000000000000000000000000000000000000000000
SURFACE ( 6) SPHERE. R=5
INDICES=( 1, 1, 1, 0, -1)
X-SCALE=( 5.000000000000000000E+00, 0)
Y-SCALE=( 5.000000000000000000E+00, 0)
Z-SCALE=( 5.000000000000000000E+00, 0)
0000000000000000000000000000000000000000000000000000000000000000
MODULE ( 3) SPHERE WITH INNER ARROW
MATERIAL( 1)
SURFACE ( 6), SIDE POINTER=( -1)
BODY ( 1)
BODY ( 2)
1111111111111111111111111111111111111111111111111111111111111111
OMEGA=( 0.000000000000000000E+00, 0) DEG
THETA=(-90.000000000000000000E+00, 0) DEG
PHI=( 90.000000000000000000E+00, 0) DEG
0000000000000000000000000000000000000000000000000000000000000000
END
0000000000000000000000000000000000000000000000000000000000000000

We have defined the entire system as a single module, so that you may rotate and/or displace it arbitrarily. Notice that the initial arrow points in the positive direction of the z-axis. It is instructive to try various rotations and use GVIEW2D or GVIEW3D (with a sector excluded to make the inner arrow visible) for visualizing the rotated system.

Writing a geometry file is nothing more than routine work. After a little practice, you can define quite complex systems by using only surfaces and bodies. You will soon
realize that the visualization programs (as well as the actual simulations!) slow down when the number of elements in the geometry increases. The only way of speeding up the programs is to group the bodies into modules. The best strategy for improving the calculation speed is to build relatively simple modules and combine them into larger parent modules to obtain a genealogical tree where the number of daughters of each module is not too large (say 4 or 5).

You may save a lot of time by defining each body separately (and checking it carefully) and then inserting it into the progressing module that, once finished, will be added to the file. Notice that the input element labels are arbitrary (as long as they are not repeated for elements of the same kind) and that we can insert new elements anywhere in the file. Once the geometry definition is complete, we can generate an equivalent file, with elements labelled according to their input order, by simply editing the `GEOMETRY.REP` file.

The previous examples of geometry files (`QUADRIC` and `ARROW`) together with several other files of more complex geometries are included in the distribution package. They can be directly visualized by running `GVIEW2D` and `GVIEW3D`. The file `GLASS` (a glass of champagne) shows that common objects can be described quite precisely with only quadric surfaces; in this case, we do not use modules, which are useful only to accelerate the calculations. `WELL` defines a scintillation well detector with much detail; we have set an enclosure for the system, so that you can rotate the entire detector by editing the definition file. Notice that, when the detector is tilted, it is very difficult to get an idea of its geometry from the images generated by `GVIEW2D`. `SATURN` describes the head of an electron accelerator, quite a complicated geometry with 96 surfaces and 44 bodies. The structure `MALE`, which corresponds to a mathematical anthropomorphic phantom, consists of 174 surfaces and 108 bodies, grouped into 11 modules.

We cannot finish without a word of caution about the use of `PENGEOM`, and other general-purpose geometry packages. For simple geometries, they tend to waste a lot of time. It is always advisable to consider the possibility of handling geometric aspects directly; this may enable substantial reduction of the number of operations by taking full advantage of the peculiarities of the material system.
Chapter 6

Structure and operation of the code system

In this chapter we describe the structure of the PENEOPE code system and its operation. The kernel of the system is the FORTRAN77 subroutine package PENEOPE, which performs “analogue” simulation of electron-photon showers (i.e. the simulated showers are intended to be replicas of actual showers) in infinite (unbounded) media of various compositions. Photon histories are generated by using the detailed simulation method (see section 1.4), i.e. all interaction events are simulated in chronological succession. The generation of electron and positron tracks is performed by using the mixed procedure described in chapter 4. Secondary particles emitted with initial energy larger than the absorption energy—see below—are stored, and simulated after completion of each primary track. Secondary particles are produced in direct interactions (hard inelastic collisions, hard bremsstrahlung emission, positron annihilation, Compton scattering, photoelectric absorption and pair production) and as fluorescent radiation (characteristic x-rays and Auger electrons). PENEOPE simulates fluorescent radiation that results from vacancies produced in K-shells and L-subshells by photoelectric absorption and Compton scattering of photons and by electron/positron impact. The relaxation of these vacancies is followed until the K- and L-shells are filled up, i.e. until the vacancies have migrated to M and outer shells.

Being a subroutine package, PENEOPE cannot operate by itself. The user must provide a steering MAIN program for his/her particular problem. Nevertheless, this MAIN program is normally fairly simple, since it only has to control the evolution of the tracks simulated by PENEOPE and keep score of relevant quantities. PENEOPE is devised to do the largest part of the simulation work. It allows the user to write his or her own simulation program, with arbitrary geometry and scoring, without previous knowledge of the intricate theoretical aspects of scattering and transport theories. In the case of material systems with quadric geometries, the geometrical operations can be done automatically by using the package PENGEOM (see chapter 5). The distribution package also includes various examples of MAIN programs for simple geometries (slab and cylindrical) and for general quadric geometries with limited scoring. Although
they are mostly intended to illustrate the use of the simulation routines, they do allow studying many cases of practical interest. The complete program system is written in FORTRAN77 (ANSI/ISO standard form) and, therefore, it should run on any platform with a FORTRAN77 or FORTRAN90 compiler.

6.1 PENELOE

PENELOE simulates coupled electron-photon transport in arbitrary material systems consisting of a number of homogeneous regions (bodies) limited by sharp (and passive) interfaces. Initially, it was devised to simulate the PENetration and Energy LOss of Positrons and Electrons in matter; photons were introduced later. The adopted interaction models (chapters 2-4), and the associated databases, allow the simulation of electron/positron and photon transport in the energy range from 100 eV to 1 GeV.

It should be borne in mind that our approximate interaction models become less accurate when the energy of the transported radiation decreases. Actually, for energies below ~1 keV, the DCSs are not well known, mostly because they are strongly affected by the state of aggregation. On the other hand, for electrons and positrons, the trajectory picture ceases to be applicable (because coherent scattering from multiple centers becomes appreciable) when the de Broglie wavelength, \( \lambda_B = (150 \text{ eV} / E)^{1/2} \) Å, is similar to or greater than the interatomic spacing (\( \sim 1 \) Å). Therefore, results from simulations with PENELOE (or with any other Monte Carlo trajectory code) for energies below 1 keV or so, should be considered to have only a qualitative (or, at most, semi-quantitative) value. We recall also that, for elements with intermediate and high atomic numbers, secondary characteristic photons with energies less than the M-shell absorption edge are not simulated by PENELOE. This sets a lower limit to the energy range for which the simulation is faithful.

The source file \texttt{PENELOE.F} (about 8000 lines of FORTRAN code) consists of four blocks of subprograms, namely, preparatory calculations and I/O routines, interaction simulation procedures, numerical routines and transport routines. Only the latter are invoked from the \texttt{MAIN} program. The interaction simulation routines implement the theory and algorithms described in chapters 2 and 3. Although the interaction routines are not called from the \texttt{MAIN} program, there are good reasons to have them properly identified. Firstly, these are the code pieces to be modified to incorporate better physics (when available) and, secondly, some of these subroutines deliver numerical values of the DCSs (which can be useful to apply certain variance reduction techniques). To have these routines organized, we have named them according to the following convention:

- The first letter indicates the particle (E for electrons, P for positrons, G for photons).
- The second and third letters denote the interaction mechanism (EL for elastic, IN for inelastic, BR for bremsstrahlung, AN for annihilation, RA for Rayleigh, CO for Compton, PH for photoelectric and PP for pair production).
- The random sampling routines have three-letter names. Auxiliary routines, which perform specific calculations, have longer names, with the fourth and subsequent letters
and/or numbers indicating the kind of calculation (TX for total x-section, DX for differential x-section) or action (W for write data on a file, R for read data from a file, I for initialization of simulation algorithm).

Thus, for instance, subroutine EEL simulates elastic collisions of electrons while subroutine EINTX computes total (integrated) cross sections for inelastic scattering of electrons.

6.1.1 Database and input material data file

PENELOPE reads the required physical information about each material (which includes tables of physical properties, interaction cross sections, relaxation data, etc.) from the input material data file (identified as UNIT=IRD in the code source listing). The material data file is created by means of the auxiliary program MATERIAL, which extracts atomic interaction data from the database. This program runs interactively and is self-explanatory. Basic information about the considered material is supplied by the user from the keyboard, in response to prompts from the program. The required information is: 1) chemical composition (i.e. elements present and stoichiometric index of each element), 2) mass density, 3) mean excitation energy and 4) energy and oscillator strength of plasmon excitations. Alternatively, for a set of 279 prepared materials, the program MATERIAL can read data directly from the PDCOMPOS.TAB file (see below).

For compounds and mixtures, the additivity approximation is adopted to define the material’s cross sections, i.e. the corresponding “molecular” cross section is set equal to the sum of atomic cross sections weighted with the stoichiometric index of the element. Alloys and mixtures are treated as compounds, with stoichiometric indices equal, or proportional, to the percent number of atoms of the elements.

The PENELOPE database consists of the following 373 ASCII files,

PDCONFIG.TAB ... Atomic ground-state configurations, ionization energies (Lederer and Shirley, 1978) and central values, J_i(p_z = 0), of the one-electron shell Compton profiles (Biggs et al., 1975) for the elements, from hydrogen to uranium.

PDCOMPOS.TAB ... This file contains composition data, densities and mean excitation energies for 279 materials, adapted from the database of the ESTAR program of Berger (1992). The first 98 entries are the elements Z = 1 – 98, ordered by atomic number Z. Materials 99 to 279 are compounds and mixtures, in alphabetical order. Notice that PENELOPE does not work for elements with atomic number Z > 92.

PDEFLIST.TAB ... List of materials predefined in file PDCOMPOS.TAB, with their identification numbers.

PDRELAX.TAB ... Data on atomic relaxation, extracted from the LLNL Evaluated Atomic Data Library (Perkins et al., 1991)

92 files named PDEELZZ.TAB with ZZ=atomic number (01–92). These files contain integrated cross sections for elastic scattering of electrons and positrons by neutral
atoms, calculated by using the partial-wave methods described in section 3.1 (Salvat, 2000). The first line in each file gives the atomic number \( ZZ \); each subsequent line has 7 columns with the following data:
1st column: kinetic energy (eV), in increasing order.
2nd column: total cross section for electrons.
3rd column: first transport cross section for electrons.
4th column: second transport cross section for electrons.
5th column: total cross section for positrons.
6th column: first transport cross section for positrons.
7th column: second transport cross section for positrons.
The grid of energies is approximately logarithmic, with 15 points per decade, and
is the same for all elements. All cross sections are in cm\(^2\).

92 files named PDEBRZZ.TAB with \( ZZ \)=atomic number (01–92). They contain the
atomic bremsstrahlung scaled cross sections (energy loss spectra) and total inte-
grated radiative cross sections of electrons, for a grid of electron kinetic energies \( E \)
and reduced photon energies \( W/E \) that is dense enough to allow the use of cubic
spline log-log interpolation in \( E \) and linear interpolation in \( W/E \). The data in
these files is from a database, with 32 reduced photon energies, which was pro-
vided to the authors by Steven Seltzer (a brief description of the methods used
to compute the database and a reduced tabulation is given in Seltzer and Berger,
1986). The format of the bremsstrahlung database files is the following,
1) The first line contains the atomic number \( ZZ \).
2) Each four-lines block contains the electron kinetic energy \( E \), the scaled energy-
loss differential cross section at the 32 fixed reduced photon energies and the value
of the integrated radiative cross section.
Energies are in eV and the values of the scaled energy-loss cross section are in
millibarn (\( 10^{-27} \) cm\(^2\)).

PDBRANG.TAB ... Gives the parameters of the analytical shape function (angular
distribution) of bremsstrahlung photons, which is expressed as a statistical mix-
ture of two Lorentz-boosted dipole distributions, eq. (3.146). The distribution
parameters were obtained by fitting the benchmark partial-wave shapes tabulated
by Kissel et al. (1983).

92 files named PDGPPZZ.TAB with \( ZZ \)=atomic number (01–92). Total cross sections
for electron-positron pair production by photons with energies up to 100 GeV
in the field of neutral atoms. The data were generated by means of the XCOM
program of Berger and Hubbell (1987). The first line of each file gives the atomic
number \( ZZ \); each subsequent line gives,
1st column: photon energy, in eV. The same energy grid for all elements,
2nd column: total cross section for pair+triplet production in barn (\( 10^{-24} \) cm\(^2\)).

92 files named PDGPHZZ.TAB with \( ZZ \)=atomic number (01–92), containing photo-
electric total atomic cross sections and partial cross sections for photoionization of
inner shells (K shell and L subshells) for the elements and photon energies in the
range from 100 eV to 1 TeV. The data were extracted from the LLNL Evaluated Photon Data Library EPDL97 (Cullen et al., 1997). The format is the following, 1) the first line contains the atomic number \(Z\) and the number \(N\) of shells for which the partial cross section is tabulated. 2) each of the following lines contains a value of the photon energy (in eV) and the corresponding total cross section and partial cross sections of the shells K, L1, L2 and L3, respectively (all cross sections in barn). For low-\(Z\) elements, L-subshells are empty and, therefore, they do not appear in the table. The grid of energies for each element was obtained by merging a generic grid (the same for all elements, covering the energy range from 100 eV to 100 GeV) with the grid of absorption edges of the element, and adding additional points (where needed) to ensure that linear log-log interpolation will never introduce relative errors larger than 0.02.

Atomic cross sections for coherent and incoherent scattering of photons, inelastic scattering of electrons and positrons, and positron annihilation are evaluated directly from the analytical DCSs described in chapters 2 and 3.

In the material definition file generated by the program MATERIAL, mean free paths, transport mean free paths and stopping powers of electrons and positrons are given in mass-thickness units (1 mtu \(\equiv 1\) g/cm\(^2\)) and eV/mtu, respectively. Photon mass attenuation coefficients are expressed in cm\(^2\)/g. These quantities are practically independent of the material density; the only exception is the collision stopping power for electrons and positrons with kinetic energies larger than about 0.5 MeV, for which the density effect correction may be appreciable.

The energy-dependent quantities tabulated in the input material data file determine the most relevant characteristics of the scattering model. Thus, the MW differential cross section for electron and positron elastic scattering is completely defined by the mean free paths and transport mean free paths. Collision and radiative stopping powers read from the input file are used to renormalize the built-in analytical differential cross sections, i.e. these are multiplied by an energy-dependent factor such that the input stopping powers are exactly reproduced. The mean free paths used in the simulation of photon transport are directly obtained from the input total cross sections. Natural cubic spline log-log interpolation is used to interpolate the tabulated energy-dependent quantities, except for the photoelectric attenuation coefficient, which is obtained by simple linear log-log interpolation in the intervals between consecutive absorption edges.

To simulate geometrical structures with several materials, the corresponding material data files generated by the program MATERIAL must be catenated in a single input file. PENELope labels the \(M\)-th material in this file with the index \(\text{MAT}=M\), which is used during the simulation to identify the material where the particle moves. The maximum number of different materials that PENELope can handle simultaneously is fixed by the parameter \(\text{MAXMAT}\), which in the present version is set equal to 10. The required memory storage is roughly proportional to the value of this parameter. The user can increase \(\text{MAXMAT}\) by editing the program source files. Notice that the value of \(\text{MAXMAT} \) must be
the same in all subprograms.

6.1.2 Structure of the MAIN program

As mentioned above, PENELCOPE must be complemented with a steering MAIN program, which controls the geometry and the evolution of tracks, keeps score of the relevant quantities and performs the required averages at the end of the simulation.

The connection of PENELCOPE and the MAIN program is done via the named common block

\[
\text{COMMON/TRACK/E}, \text{X}, \text{Y}, \text{Z}, \text{U}, \text{V}, \text{W}, \text{WGHT, KPAR, IBODY, MAT, ILB(5)}
\]

that contains the following particle state variables:

- \text{KPAR} \ ... \ kind of particle (1: electron, 2: photon, 3: positron).
- \text{E} \ ... \ current particle energy (eV) (kinetic energy for electrons and positrons).
- \text{X}, \text{Y}, \text{Z} \ ... \ position coordinates (cm).
- \text{U}, \text{V}, \text{W} \ ... \ direction cosines of the direction of movement.
- \text{WGHT} \ ... \ in analogue simulations, this is a dummy variable. When using variance reduction methods, the particle weight can be stored here.
- \text{IBODY} \ ... \ this auxiliary flag serves to identify different bodies in complex material structures.
- \text{MAT} \ ... \ material where the particle moves (i.e. the one in the body labelled \text{IBODY}).
- \text{ILB(5)} \ ... \ an auxiliary array of 5 labels that describe the origin of secondary particles (see below). It is useful e.g. to study partial contributions from particles originated by a given process.

The position coordinates \( \mathbf{r} = (\text{X}, \text{Y}, \text{Z}) \) and the direction cosines \( \hat{\mathbf{d}} = (\text{U}, \text{V}, \text{W}) \) of the direction of movement are referred to a fixed rectangular coordinate system, the “laboratory” system, which can be arbitrarily defined. During the simulation, all energies and lengths are expressed in eV and cm, respectively.

The label \text{KPAR} identifies the kind of particle: \text{KPAR=1}, electron; \text{KPAR=2}, photon; \text{KPAR=3}, positron. A particle that moves in material \( \text{M} \) is assumed to be absorbed when its energy becomes less than a value \( \text{EABS(KPAR, M)} \) (in eV) specified by the user. Positrons are assumed to annihilate, by emission of two photons, when absorbed. In dose calculations, \( \text{EABS(KPAR, M)} \) should be determined so that the residual range of particles with this energy is smaller than the dimensions of the volume bins used to tally the spatial dose distribution. As the interaction database is limited to energies above 100 eV, absorption energies \( \text{EABS(KPAR, M)} \) must be larger than this value.

The transport algorithm for electrons and positrons in each material \( \text{M} \) is controlled by the following simulation parameters,
6.1. PENELPOPE

C1(M) ... Average angular deflection, \( C_1 \approx 1 - \langle \cos \theta \rangle \) [eq. (4.11)], produced by multiple elastic scattering along a path length equal to the mean free path between consecutive hard elastic events [see eq. (4.1)]. \( C_1(M) \) should be of the order of 0.05; its maximum allowed value is 0.2.

C2(M) ... Maximum average fractional energy loss, \( C_2 \) [eq. (4.85)], between consecutive hard elastic events. Usually, a value of the order of 0.05 is adequate. The maximum allowed value of \( C_2(M) \) is 0.2.

WCC(M) ... Cutoff energy loss, \( W_{cc} \) (in eV), for hard inelastic collisions in the \( M \)th material.

WCR(M) ... Cutoff energy loss, \( W_{cr} \) (in eV), for hard bremsstrahlung emission in material \( M \).

These parameters determine the accuracy and speed of the simulation. To ensure accuracy, \( C_1(M) \) and \( C_2(M) \) should have small values (of the order of 0.01 or so). With larger values of \( C_1(M) \) and \( C_2(M) \) the simulation gets faster, at the expense of a certain loss in accuracy. The cutoff energies \( WCC(M) \) and \( WCR(M) \) mainly influence the simulated energy distributions. The simulation speeds up by using larger cutoff energies, but if these are too large, the simulated energy distributions may be somewhat distorted. In practice, simulated energy distributions are found to be insensitive to the adopted values of \( WCC(M) \) and \( WCR(M) \) when these are less than the bin width used to tally the energy distributions. Thus, the desired energy resolution determines the maximum allowed cutoff energies. The reliability of the whole simulation rests on a single condition: the number of steps (or random hinges) per primary track must be “statistically sufficient”, i.e. larger than 10 or so.

The simulation package is initialized from the \texttt{MAIN} program with the statement

\begin{verbatim}
CALL PElIT(EPMAX,NMAT,IRD,IWR,INFO)
\end{verbatim}

Subroutine \texttt{PEINIT} reads the data files of the different materials, evaluates relevant scattering properties and prepares look-up tables of energy-dependent quantities that are used during the simulation. Its input arguments are:

\texttt{EPMAX} ... Maximum energy (in eV) of the simulated particles. Notice that if the primary particles are positrons with initial kinetic energy \( E_P \), the maximum energy of annihilation photons may be close to (but less than) \( EPMAX = 1.21(E_P + m_e c^2) \); in this special case, the maximum energy is larger than the initial kinetic energy.

\texttt{NMAT} ... Number of different materials (less than or equal to \texttt{MAXMAT}).

\texttt{IRD} ... Input unit.

\texttt{IWR} ... Output unit.

\texttt{INFO} ... Determines the amount of information that is written on the output unit. Minimal for \texttt{INFO} = 0 and increasingly detailed for \texttt{INFO} = 1, 2, etc.
For the preliminary computations, **PEINIT** needs to know the absorption energies \( E_{ABS}(KPAR, M) \) and the simulation parameters \( C_1(M), C_2(M), \ WCC(M) \) and \( \ WCR(M) \). This information is introduced through the named common block

\[
\rightarrow \text{COMMON/CSIMPA/EABS(3, MAXMAT), C1(MAXMAT), C2(MAXMAT), WCC(MAXMAT), WCR(MAXMAT)}
\]

that has to be loaded before invoking subroutine **PEINIT**. Notice that we can employ different values of the simulation parameters for different materials. This possibility can be used to speed up the simulation in regions of lesser interest.

**PENELOPE** has been structured in such a way that a particle track is generated as a sequence of track segments (free flights or “jumps”); at the end of each segment, the particle suffers an interaction with the medium (a “knock”) where it loses energy, changes its direction of movement and, in certain cases, produces secondary particles. Electron-photon showers are simulated by successively calling the following subroutines:

**SUBROUTINE CLEANS** ... Initiates the secondary stack.

**SUBROUTINE START** ... For electrons and positrons, this subroutine forces the following interaction event to be a soft artificial one. It must be called before starting a new primary or secondary event and also when a track crosses an interface.

Calling **START** is strictly necessary only for electrons and positrons; for photons this subroutine has no physical effect. However, it is advisable to call **START** for any kind of particle since it checks whether the energy is within the expected range, and can thus help to detect “bugs” in the **MAIN** program.

**SUBROUTINE JUMP(DSMAX, DS)** ... Determines the length \( DS \) of the track segment to the following interaction event.

The input parameter \( D_{\text{MAX}} \) defines the maximum allowed step length for electrons/positrons; for photons, it has no effect. As mentioned above, to limit the step length, **PENELOPE** places delta interactions along the particle track. These are fictitious interactions that do not alter the physical state of the particle. Their only effect is to interrupt the sequence of simulation operations (which requires altering the values of inner control variables to permit resuming the simulation in a consistent way). The combined effect of the soft interactions that occur along the step preceding the delta interaction is simulated by the usual random hinge method.

As mentioned above, to ensure the reliability of the mixed simulation algorithm, the number of artificial soft events per particle track in each body should be larger than, say, 10. For relatively thick bodies (say, thicker than 10 times the mean free path between hard interactions), this condition is automatically satisfied. In this case we can switch off the step-length control by setting \( D_{\text{MAX}}=1.0 \ D_{35} \) (or any other very large value). On the other hand, when the particle moves in a thin body, \( D_{\text{MAX}} \) should be given a value of the order of one tenth of the “thickness” of that body. Limiting the step length is also necessary to simulate particle transport in external electromagnetic fields.
6.1. PENEOPE

SUBROUTINE KNOCK(DE,ICOL) ... Simulates an interaction event, computes new energy and direction of movement, and stores the initial states of the generated secondary particles, if any. On output, the arguments are:
DE ... deposited energy in the course of the event,
ICOL ... kind of event that has been simulated, according to the following convention,
- Electrons (KPAR=1)
  ICOL=1, artificial soft event (random hinge).
  -2, hard elastic collision.
  -3, hard inelastic collision.
  -4, hard bremsstrahlung emission.
- Photons (KPAR=2)
  ICOL=1, coherent (Rayleigh) scattering.
  -2, incoherent (Compton) scattering.
  -3, photoelectric absorption.
  -4, electron-positron pair production.
- Positrons (KPAR=3)
  ICOL=1, artificial soft event (random hinge).
  -2, hard elastic collision.
  -3, hard inelastic collision.
  -4, hard bremsstrahlung emission.
  -5, annihilation.

For electrons and positrons ICOL=7 corresponds to delta interactions. The value ICOL=6 is used for the “auxiliary” interactions (an additional mechanism that may be defined by the user, e.g. to simulate photonuclear interactions, see the source file PENEOPE.F).

SUBROUTINE SECPAR(LEFT) ... Sets the initial state of a secondary particle and removes it from the secondary stack. The output value LEFT is the number of secondary particles that remained in the stack at the calling time.

SUBROUTINE STORES(E,X,Y,Z,U,V,W,WGHT,KPAR,ILB) ... Stores a particle in the secondary stack. Arguments have the same meaning as in COMMON/TRACK/, but refer to the particle that is being stored. The variables IBODY and MAT are set equal to the current values in COMMON/TRACK/.

Calling STORES from the MAIN program is useful e.g. to store particles produced by splitting, a variance reduction method (see section 1.6.2).

The sequence of calls to generate a random track is independent of the kind of particle that is being simulated. The generation of random showers proceeds as follows (see fig. 6.1):

(i) Set the initial state of the primary particle, i.e. assign values to the state variables KPAR, E, position coordinates \( \mathbf{r} = (X,Y,Z) \) and direction of movement \( \hat{d} = (U,V,W) \).

Specify the body and material where the particle moves by defining the values of IBODY and MAT, respectively. Optionally, set the values of WGHT and ILB(1:5).
Figure 6.1: Flow diagram of the **MAIN** program for simulating electron-photon showers with PENELOPE.
(ii) **CALL CLEAN**S to initialize the secondary stack.

(iii) **CALL START** to initiate the simulation of the track.

(iv) **CALL JUMP(\(DS_{\text{MAX}}\), DS)** to determine the length DS of the next track segment (for electrons and positrons, DS will never exceed the input value \(DS_{\text{MAX}}\)).

(v) Compute the position of the following event:

- If the track has crossed an interface, stop the particle at the position where the track intersects the interface, and shorten the step length DS accordingly. Change to the new material (the one behind the interface) by redefining the variables \(IBODY\) and \(MAT\).
  
  When the particle escapes from the system, the simulation of the track has been finished; increment counters and go to step (vii).

  Go to step (iii).

(vi) **CALL KNOCK(DE, ICOL)** to simulate the following event.

- If the energy is less than \(EABS(KPAR, MAT)\), end the track, increment counters and go to step (vii).

- Go to step (iv).

(vii) **CALL SECPAR(LEFT)** to start the track of a particle in the secondary stack (this particle is then automatically removed from the stack).

- If \(\text{LEFT} > 0\), go to step (iii). The initial state of a secondary particle has already been set.

- If \(\text{LEFT} = 0\), the simulation of the shower produced by the primary particle has been completed. Go to step (i) to generate a new primary particle (or leave the simulation loop after simulating a sufficiently large number of showers).

Notice that subroutines **JUMP** and **KNOCK** keep the position coordinates unaltered; the positions of successive events have to be followed by the **MAIN** program (simply by performing a displacement of length DS along the direction of movement after each call to **JUMP**). The energy of the particle is automatically reduced by subroutine **KNOCK**, after generating the energy loss from the relevant probability distribution. **KNOCK** also modifies the direction of movement according to the scattering angles of the simulated event. Thus, at the output of **KNOCK**, the values of the energy \(E\), the position \(r = (X, Y, Z)\) and the direction of movement \(\mathbf{d} = (u, v, w)\) define the particle state immediately after the interaction event.

In order to avoid problems related with possible overflows of the secondary stack, when a secondary particle is produced its energy is temporarily assumed as locally deposited. Hence, the energy \(E\) of a secondary must be subtracted from the corresponding dose counter when the secondary track is started. Occasional overflows of the secondary stack are remedied by eliminating the less energetic secondary electron or photon in the stack (positrons are not eliminated since they will eventually produce quite energetic annihilation radiation). As the main effect of secondary particles is to spread out the
energy deposited by the primary one, the elimination of the less energetic secondary electrons and photons should not invalidate local dose calculations.

It is the responsibility of the user to avoid calling subroutines \texttt{JUMP} and \texttt{KNOCK} with energies outside the interval \((\text{EABS(KPAR, M), EMAX})\). This could cause improper interpolation of the cross sections. The simulation is aborted (and an error message is printed in unit 6) if the conditions \(\text{EABS(KPAR) < E < EMAX}\) are not satisfied when a primary or secondary track is started (whenever subroutine \texttt{START} is called at the beginning of the track).

Pseudo-random numbers uniformly distributed in the interval \((0, 1)\) are supplied by function \texttt{RAND(DUMMY)} that implements a 32-bit generator due to L’Ecuyer (see table 1.1). The seeds of the generator (two integers) are transferred from the \texttt{MAIN} program through the named common block \texttt{RSEED} (see below). The random number generator can be changed by merely replacing that \texttt{FUNCTION} subprogram (the new one has to have a single dummy argument). Some compilers incorporate an intrinsic random number generator with the same name (but with different argument lists). To avoid conflict, \texttt{RAND} should be declared as an external function in all subprograms that call it.

Notice that

(1) In the simulation routines, real and integer variables are declared as \texttt{DOUBLE PRECISION} and \texttt{INTEGER*4}, respectively. To prevent type mismatches, it is prudent to use the following \texttt{IMPLICIT} statement

\[
\rightarrow \text{IMPLICIT DOUBLE PRECISION (A-H, O-Z), INTEGER*4 (I-N)}
\]

in the \texttt{MAIN} program and other user program units.

(2) The \texttt{MAIN} program \texttt{must} include the following three common blocks:

\[
\rightarrow \text{COMMON/TRACK/E,X,Y,Z,U,V,W,WGHT,KPAR,IBODY, MAT,ILB(5)}
\]

\[
\rightarrow \text{COMMON/CSIMPA/EABS(3, MAXMAT), C1(MAXMAT), C2(MAXMAT), WCC(MAXMAT),}
\]

\[
1 \text{ WC1(MAXMAT)} \quad \text{! Simulation parameters.}
\]

\[
\rightarrow \text{COMMON/RSEED/ISEED1, ISEED2} \quad \text{! Random number generator seeds.}
\]

As mentioned above, \texttt{ILB(5)} is an array of labels that describe the origin of secondary particles. It is assumed that the user has set \texttt{ILB(1)} equal to 1 (one) when a primary (source) particle history is initiated. Then, \texttt{PENEOLOPE} assigns the following labels to each particle in a shower:

\textbf{ILB(1)}: generation of the particle. 1 for primary particles, 2 for their direct descendants, etc.

\textbf{ILB(2)}: kind \texttt{KPAR} of the parent particle, only if \texttt{ILB(1)}>1 (secondary particles).

\textbf{ILB(3)}: interaction mechanism \texttt{ICOL} (see above) that originated the particle, only when \texttt{ILB(1)}>1.

\textbf{ILB(4)}: a non-zero value identifies particles emitted from atomic relaxation events and describes the atomic transition where the particle was released. The numerical value is \(Z \cdot 10^6 + IS1 \cdot 10^4 + IS2 \cdot 100 + IS3\), where \(Z\) is the atomic number of the parent atom and \(IS1, IS3\) and \(IS3\) are the active atomic electron shells.
6.2. Examples of \texttt{MAIN} programs

\textbf{ILB(5)}: this label can be defined by the user; it is transferred to all descendants of the particle.

The \texttt{ILB} label values are delivered by subroutine \texttt{SECPAR}, through common \texttt{TRACK}, and remain unaltered during the simulation of the track.

Owing to the long execution time, the code will usually be run in batch mode. It is advisable to limit the simulation time rather than the number of tracks to be simulated, since the time required to follow each track is difficult to predict. To this end, one can link a clock routine to the simulation code and stop the computation after exhausting the allotted time; examples of clock routines for two different compilers are included in the PENEOPE distribution package.

\subsection{Variance reduction}

The subroutine package \texttt{PENEOPE.F} is intended to perform analogue simulation and, therefore, does not include any variance reduction methods. The source file \texttt{PENVARED.F} contains subroutines to perform splitting (\texttt{SPLIT}), Russian roulette (\texttt{VKILL}) and interaction forcing (\texttt{JUMPF, KNOCKF}) in an automatic way. Splitting and Russian roulette (see section 1.6.2) do not require changes in PENEOPE; the necessary manipulations on the numbers and weights \texttt{WGHT} of particles could be done directly in the main program. Particles resulting from splitting are stored in the secondary stack by calling subroutine \texttt{STORES}.

Interaction forcing (section 1.6.1) implies changing the mean free paths of the forced interactions and, at the same time, redefining the weights of the generated secondary particles. In principle, it is possible to apply interaction forcing from the \texttt{MAIN} program by manipulating the interaction probabilities, that are made available through the named common block \texttt{CJUMPO}. These manipulations are performed automatically by calling the subroutines \texttt{JUMPF} and \texttt{KNOCKF} instead of \texttt{JUMP} and \texttt{KNOCK}.

Although these subroutines operate like "black boxes", they should be invoked with care. In general, it is advisable to prevent particle weights from reaching very large or very small values. In the first case, a very "heavy" particle can completely hide the information collected from many lighter particles. Conversely, it is not convenient to spend time simulating particles with very small weights, which contribute insignificant amounts to the scores. Notice also that repeated splitting and interaction forcing may easily lead to saturation of the secondary stack (the default stack size is 1000 particles). Hence, we usually apply interaction forcing only to primary particles.

\section{Examples of \texttt{MAIN} programs}

In general, the user must provide the \texttt{MAIN} program for each specific geometry. The distribution package of PENEOPE includes various examples of \texttt{MAIN} programs for simple geometries (slab and cylindrical) and for general quadric geometries with limited scoring.
In these examples, we assume that a single kind of particles is emitted from the radiation source. The programs can be easily generalized to the case of multi-particle sources with continuous (or discrete) energy spectra. For details on the operation of these codes, see section 6.2.4 below and the heading comments in the corresponding source files.

### 6.2.1 Program PENSLAB

The program PENSLAB simulates electron/photon showers within a material slab (see fig. 6.2). It illustrates the use of the simulation routines for the simplest geometry (as geometry operations are very simple, this program is faster than the ones described below). The slab is limited by the planes $z = 0$ and $z = t$, the thickness. The lateral extension of the slab is assumed to be infinite, i.e. much larger than the maximum range of the particles). Primary particles start with a given energy $E_0$ from a point source at a given “height” $z_0$ (positive or negative) on the $z$-axis, and moving in directions distributed uniformly in a spherical “sector” defined by its limiting polar angles, say $\theta_1$ and $\theta_2$, which is indicated by the hatched wedge in fig. 6.2. That is, to generate the initial direction, the polar cosine $\hat{\mathbf{w}} = \cos \theta$ is sampled uniformly in the interval from $\cos \theta_1$ to $\cos \theta_2$ and the azimuthal angle $\phi$ is sampled uniformly in $(0, 2\pi)$. Thus, the case $\theta_1 = 0$ and $\theta_2 = 180$ deg corresponds to an isotropic source, whereas $\theta_1 = \theta_2 = 0$ defines a beam parallel to the $z$-axis. Notice that the complete arrangement has rotational invariance about the $z$-axis.

![Diagram of PENSLAB](image.png)

**Figure 6.2:** General planar geometry considered in PENSLAB.

**PENSLAB** generates detailed information on many quantities and distributions of physical interest. The output files contain a self-explanatory report of the simulation results, which consist of:

(i) Fractions of primary particles that are transmitted, backscattered and absorbed and a number of average quantities (track length within the sample; number of events of each kind per particle; energy, direction and lateral displacement of particles that leave the sample, etc.).

(ii) Energy distributions of transmitted and backscattered primary particles.
(iii) Angular distributions of transmitted and backscattered particles.
(iv) Depth-dose distribution (i.e. deposited energy per unit depth).
(v) Depth-distribution of deposited charge.
(vi) Distribution of energy deposited into the slab.

Each simulated continuous distribution is printed on a separate file (as a histogram), with a heading describing its content and in a format ready for visualization with a plotting program. The code computes and delivers the statistical uncertainties ($3\sigma$) of all evaluated quantities and distributions. Many authors quote these uncertainties as one standard deviation, which means that the probability for the actual value to lie outside the error bar is 0.317. We prefer to be more conservative and stay at the $3\sigma$ level, for which the probability of “missing the real value” is only 0.003.

The program **PENSILAB** and its predecessors have been intensively used during the last years to analyze the reliability of **PENELOPE**. They have been applied to a variety of experimental situations, covering a wide energy range. Benchmark comparisons with experimental data have been published elsewhere (Baró et al., 1995; Sempau et al., 1997).

**WARNING:** In the output files of **PENSILAB** (and also in those of the program **PENCYL** described below), the terms “transmitted” and “backscattered” are used to denote particles that leave the material system moving upwards ($\dot{W} > 0$) and downwards ($\dot{W} < 0$), respectively. Notice that this agrees with the usual meaning of these terms only when primary particles impinge on the system coming from below (i.e. with $\dot{W} > 0$).

### 6.2.2 Program **PENNCYL**

The program **PENNCYL** simulates electron-photon showers in multilayered cylindrical structures. The material system consists of one or several layers of given thicknesses. Each layer contains a number of concentric homogeneous rings of given compositions and radii (and thickness equal to that of the layer). The layers are perpendicular to the z-axis and the centre of the rings in each layer is specified by giving its $x$ and $y$ coordinates. When all the centres are on the z-axis, the geometrical structure is symmetrical under rotations about the z-axis (see fig. 6.3).

Primary particles of a given kind, **KPARP**, are emitted from a point or extense (cylindrical) source, with its centre at the point ($sx0, sy0, sz0$), either with fixed energy $se0$ or with a specified (piecewise constant) energy spectrum. The initial direction of the primary particles is sampled uniformly inside a cone of (semi)aperture $SA\,PH\,A$ and with central axis in the direction ($s\theta, s\phi$). Thus, the case $SA\,PH\,A = 0$ defines a monodirectional source and $SA\,PH\,A = 180$ deg corresponds to an isotropic source. When $sx0 = sy0 = 0$ and $s\theta = 0$ or 180 deg, the source is axially symmetrical about the z-axis.

In the distributed form of the program, we assume that both the source and the
material structure are symmetrical about the z-axis, because this eliminates the dependence on the azimuthal angle \( \phi \). The program takes advantage of this symmetry to tally 3D dose distributions. It is possible to consider geometries that are not axially symmetrical, but then the program only delivers values averaged over \( \phi \). To obtain the dependence of the angular distributions on the azimuthal angle, we need to increase the value of the parameter \texttt{NBPHM} (the maximum number of bins for \( \phi \), which is set equal to 1 in the distributed source file) and, in the input data file, set \texttt{NBPH} equal to \texttt{NBPHM}.

The source file \texttt{PENCYL.F} includes a (self-contained) set of geometry routines for tracking particles through multilayered cylindrical structures. These routines can be used for simulation even when the source is off-axis. Cylindrical geometries can be viewed with the program \texttt{GVIEWC} (which is similar to \texttt{GVIEW2D} and runs only under Microsoft Windows 9x). This program reads the geometry definition list from a file and displays a two-dimensional map of the materials intersected by the window plane. It is useful for debugging the geometry definition list.

\texttt{PENCYL} delivers detailed information on the transport and energy deposition, which includes energy and angular distributions of emerging particles, depth-dose distribution, depth-distribution of deposited charge, distributions of deposited energy in selected materials and 2D (depth-radius) dose and deposited charge distributions in selected bodies (cylinders). \texttt{PENCYL} can be directly used to study radiation transport in a wide variety of practical systems, e.g. planar ionization chambers, cylindrical scintillation detectors, solid-state detectors and multilayered structures.

![Diagram of cylindrical geometry](image)

**Figure 6.3:** An example of cylindrical geometry, a cavity (C) with walls, with a point off-axis source. In this case, the material structure is symmetrical about the z-axis, but the radiation flux and other three-dimensional quantities (e.g. dose and deposited charge distributions) depend on the azimuthal angle \( \phi \).
6.2.3 Program PENDOSES

This MAIN program provides a practical example of simulation with complex material structures (quadric geometry only). It assumes a point source of primary particles at a given position \( r_0 = (X_0, Y_0, Z_0) \) which emits particles in directions uniformly distributed in a cone with (semi)aperture \( \text{SALPHA} \) and central axis in the direction \( \text{STHETA}, \text{SPHI} \) [the same direction distribution assumed in the code \text{PENCYL}]. The geometry of the material system is described by means of the package PENGEOM (chapter 5).

PENDOSES computes only the average energy deposited in each body per primary particle. With minor modifications, it also provides the probability distribution of the energy deposited in selected bodies or groups of bodies. It is a simple exercise to introduce a spatial grid, and the corresponding counters, and tally spatial dose distributions. Any future user of PENELPOE should become familiar with the programming details of PENDOSES before attempting her/his own application of PENELPOE.

6.2.4 Running the PENCYL program

The programs PENSFLAB, PENCYL and PENDOSES operate in a similar way. They all read data from a corresponding input file and output the results in a number of files with fixed names\(^1\). The input files also have similar structures and formats. For concreteness, here we describe that of PENCYL, which is the most versatile of the example programs.

Each line in the input data file of PENCYL consists of a 6-character keyword (columns 1-6) followed either by numerical data (in free format) or by a character string, which start at the 8th column. Keywords are explicitly used/verified by the program (which is case sensitive!). Notice also that the order of the data lines is important. The keyword “------” (6 blanks, which we have denoted by “-”) indicates comment lines, these can be placed anywhere after the end of the geometry definition list. The program ignores any text following the first blank after the last numerical datum, or after the character string, in each line (thus, in the table given below, the comments in square brackets are ignored by the program). Lines with certain keywords (e.g., “SPECTR”) can appear an arbitrary number of times, limited only by the allocated amount of memory. The program assigns default values to many input variables; lines that declare default values may be removed from the input file. Notice that the default source is a pencil beam that moves upwards along the \( z \)-axis.

- Structure of the PENCYL input file (the 72-column rulers are just for visual aid, they do not form part of the input file)

    ....+....1....+....2....+....3....+....4....+....5....+....6....+....7...
    TITLE  Title of the job, up to 65 characters.
    GSTART >>>>>>>> Beginning of the geometry definition list.

\(^1\)Warning: The programs overwrite older output files that are left in the working directory. You should save all result files on a separate directory before running the program again.
layer ZLOW,ZHIG [Z_lower and Z_higher]
centre XCEN,YCEN [X_centre and Y_centre]
cylind M,RIN,ROUT [Material, R_inner and R_outer]
gend

>>>>>>>> Source definition.

skpar KPARP [Primary particles: 1=electron, 2=photon, 3=positron]
SENERG SENERG [Initial energy (monoenergetic sources only)]
spectr Ei,pi [E bin: lower-end and total probability]
STHICK STHICK [Source thickness]
SRADII SRIN,SROUT [Source inner and outer radii]
SPOSIT SXO,SYO,SZ0 [Coordinates of the source centre]
SDIREC STHETA,SPHI [Beam axis direction angles, in deg]
SAPERT SALPHA [Beam aperture, in deg]

>>>>>>> Material data and simulation parameters.

nmat NMAT [Number of different materials, .le.10]
SIMPAR M,EABS(1:3,M),C1,C2,WCC,WCR [Sim. parameters for material M]
PFNAME filename_0.ext [Material definition file, 18 characters]

>>>>>> Counter array dimensions and pdf ranges.

NBE NBE,EMIN,EMAX [No. of energy bins, and E-interval]
NBTH NBTH [No. of bins for the polar angle THETA]
NBPH NBPH [No. of bins for the azimuthal angle PHI]
NBZ NBZ [No. of bins for the Z-coordinate]
NBR NBR [No. of radial bins]
NBTL NBTL,TLMIN,TLMAX [No. of track-length bins and TL-interval]

>>>>>> Additional distributions to be tallied.

ABSEN MAT [Tally the distr. of absorbed E in material MAT]
DOSE2D KL,KC [Tally 2D dose and charge distribns. in body KL,KC]

>>>>>> Interaction forcing.

IFORCE KL,KC,KPAR,ICOL,FORCE,WL,WH [Interaction forcing parameters]

>>>>>> Job properties.

RESUME filename1.ext [Resume from this dump file, 18 characters]
DUMPTO filename2.ext [Generate this dump file, 18 characters]
NSIMSH NTOT [Desired number of simulated showers, max=2**31-1]
RSEED ISSEED1,ISSEED2 [Seeds of the random number generator]
TIME ITIME [Allotted simulation time, in sec]

The following listing describes the function of each of the keywords, the accompanying
6.2. Examples of MAIN programs

... title of the job (up to 65 characters).
Default: none (the input file must start with this line)

Geometry definition list ... begins with the line “GSTART” and ends with the line “GEND—” (notice the two blanks). The only allowed keywords in the geometry list are “GSTART”, “LAYER—”, “CENTRE”, “CYLIND” and “GEND—”. The line after “GSTART” must be a “LAYER—” line. Each “LAYER—” line contains the z-coordinates of its lower and upper limiting planes and is followed by a “CENTRE” line (optional) and by one or several “CYLIND” lines, which contain the inner and outer radii of the various concentric rings in the layer; empty layers are disregarded.

Layers must be defined in increasing order of heights, from bottom to top of the structure. If the “CENTRE” line is not entered, cylinders are assumed to be centered on the z-axis (Xcen = Ycen = 0.0). Cylinders have to be defined in increasing radial order, from the centre to the periphery. The two lengths in each “LAYER—” and “CYLIND” line must be entered in increasing order. The geometry definition list can be debugged/visualized with the code GVIEWC (operable only under Microsoft Windows 9x).

SKPAR ... kind of primary particle (1=electrons, 2=photons or 3=positrons).
Default: KPARP=1

SENERG ... for monoenergetic sources: initial energy SE0 of primary particles.
Default: SE0=1.0E6

SPECTR ... For sources with continuous (stepwise constant) energy spectra. Each “SPECTR” line gives the lower end-point of an energy bin of the source spectrum and the associated relative probability, integrated over the bin. Up to NSEM = 200 lines, in arbitrary order. The upper end of the spectrum is defined by entering a line with the upper energy value and null probability.
Default: none

STHICK ... thickness of the active volume of the source (cylinder).
Default: STHICK=0.0

SRADII ... inner and outer radii of the active source volume.
 Defaults: SRIN=0.0, SROUT=0.0

SPOSIT ... coordinates of the centre of the source volume.
 Defaults: SX0=SY0=0, SZ0=-1.0E15

SDIREC ... polar angle θ and azimuthal angle φ of the source beam axis direction, in deg.
 Default: STHETA=0.0, SPHI=0.0

SAPERT ... angular aperture α of the source beam, in deg.
 Default: SALPHA=0.0

NMAT ... number of different materials (up to 10 with the original program dimensions). Materials are identified by their ordering in PENEOLE’s input material
data file.
Default: NMAT=1

SIMPAR ... set of simulation parameters for the M-th material: absorption energies, 
\( E_{ABS}(1:3,M) \), elastic scattering parameters, \( C_1(M) \) and \( C_2(M) \), and cutoff energy 
losses for inelastic collisions and bremsstrahlung emission, \( WCC(M) \) and \( WCR(M) \). 
One line for each material.
Defaults: \( E_{ABS}(1,M)=E_{ABS}(3,M)=0.01\times E_{MAX}, \) \( E_{ABS}(2,M)=0.001\times E_{MAX} \)
\( C_1(M)=C_2(M)=0.1, \) \( WCC=E_{ABS}(1,M), \) \( WCR=E_{ABS}(2,M) \)

\( E_{MAX} \) is the maximum energy of all particles found in the simulation (depends on 
the source energies).

PFNAME ... name of PENELOPE’s input material data file (18 characters).
Default: ‘material.mat’

NBE ... number of energy bins and limits of the interval where energy distributions 
are tallied.
Defaults: NBE=100, EMIN=0.0, EMAX=E_{MAX}

NBTH ... number of bins for the polar angle \( \theta \).
Default: NBTH=90

NBPH ... number of bins for the azimuthal angle \( \phi \).
Default: NBPH=1 (azimuthal average).

NBZ ... number of bins for the z-coordinate.
Default: NBZ=100

NBR ... number of bins for the radial variable, \( r = (x^2 + y^2)^{1/2} \).
Default: NBR=100

NBTL ... number of bins for track-length distributions of primary particles. Limits 
of the interval where these distributions are tallied.
Defaults: NBTL=100, TLMIN=0, TLMAX=2.5*RANGE(E_{MAX},KPARP,1)

ABSEN ... indicates a material M for which we require the code to tally the distribution 
of absorbed energy (up to three different materials can be selected, a separate line 
for each).
Default: off

DOSE2D ... the program will tally 2D, depth-radius, dose and deposited charge dist-
tributions in the cylinder \( KC \) of layer \( KL \). Up to three different cylinders (bodies) 
can be selected, a line for each body).
Default: off
Note: The labels \( KL, KC \) that identify a given cylinder are defined by the order-
ing in the input geometry list. These labels are written on the output geometry report.

IFORCE ... activates forcing of interactions of type ICOL of particles KPAR in cylinder 
\( KC \) of layer \( KL \). FORCE is the forcing factor and WLOW, WHIG are the limits of the 
weight window where interaction forcing is active.
Default: no interaction forcing
6.2. Examples of MAIN programs

RESUME ... the program will read the dump file named filename1.ext (18 characters) and resume the simulation from the point where it was left. Use this option very, very carefully. Make sure that the input data file is fully consistent with the one used to generate the dump file.
Default: off

DUMPTO ... generate a dump file named filename2.ext (18 characters) after completing the simulation run. This allows resuming the simulation to improve statistics.
Default: off

NSIMSH ... desired number of simulated showers. Notice that NTOT is an INTEGER*4 value and, hence, it cannot exceed $2^{31} - 1$.
Default: NTOT=2147 million

RSEED ... seeds of the random number generator.
Default: ISEED1=12345; ISEED2=54321

TIME ... allotted simulation time, in sec.
Default: ITIME=100

The program PENCYL (as well as the other example MAIN programs) is aborted when an incorrect input datum is found. The conflicting quantity usually appears in the last line of the output file. If the trouble is with arrays having dimensions smaller than required, the program indicates how the problem can be solved (this usually requires editing the source file, be careful).

The example of input file given below belongs to the PENCYL file set included in the distribution package. It corresponds to the simulation of a narrow photon beam with $E_0 = 1.25$ MeV (roughly the average energy of gamma rays from $^{60}$Co) entering a $3'' \times 3''$ NaI scintillation detector in an Al case, whose inner surface is partially covered by a layer of Al$_2$O$_3$, which diffuses scintillation light back to the crystal and the photomultiplier. In the material data file NAIAL.MAT, the order of the materials is NaI (MAT=1), Al$_2$O$_3$ (MAT=2) and Al (MAT=3). The incident beam photons move along the z-axis with $\theta = 0$ deg (i.e. upwards) and impinge normally on the surface of the detector. The geometry is shown schematically in the insets of fig. 6.4, which displays two of the distributions generated by PENCYL. The plotted distributions were obtained from 5 million random showers; the error bars represent statistical uncertainties ($3\sigma$), which are pretty small in this case.

* Example input file of the PENCYL code.

...+...1......+...2......+...3......+...4......+...5......+...6......+...7...
TITLE Example from the distribution package...
GSTART NaI detector with Al cover and Al2O3 reflecting foil
LAYER -0.24 -0.16 1
CENTRE 0.00 0.00
CYLIND 3 0.00 4.05
LAYER -0.16  0.00 2
Figure 6.4: Partial results from PENCYL for the NaI photon detector described in the text. Top: distribution of energy deposited in the NaI crystal (MAT=1). Bottom: energy distribution of backscattered photons.
6.3. Selecting the simulation parameters

|
| CYLIND | 2 | 0.00 | 3.97 |
| CYLIND | 3 | 3.97 | 4.05 |
| LAYER  | 0.00 | 7.72 | 3 |
| CYLIND | 1 | 0.00 | 3.81 |
| CYLIND | 2 | 3.81 | 3.97 |
| CYLIND | 3 | 3.97 | 4.05 |
| LAYER  | 7.72 | 9.72 | 4 |
| CYLIND | 3 | 0.00 | 4.05 |

END

SKPAR 2
[Primary particles: 1=electron, 2=photon, 3=positron]
SENEG 1.25E6
[Initial energy (monoenergetic sources only)]
SPOST 0.0 0.0 -10.0
[Coordinates of the source centre]
NMAT 3
[Number of different materials, .1e.10]
SIMPAR 1 1.0E5 1000 1.0E5 0.1 0.1 1.0E4 1000 [M,EABS,C1,C2,WCC,WCR]
SIMPAR 2 1.0E5 1000 1.0E5 0.1 0.1 1.0E4 1000 [M,EABS,C1,C2,WCC,WCR]
SIMPAR 3 1.0E5 1000 1.0E5 0.1 0.1 1.0E4 1000 [M,EABS,C1,C2,WCC,WCR]
PFNAME naial.mat
[Material definition file, 18 characters]
ABSE 1
[Tally the distr. of absorbed E in this material]
DOSE2D 3 1
[Tally 2D dose and charge distr. in this body]
RESUME lastdump.dat
[Read from this dump file, 18 characters]
DUMPPD lastdump.dat
[Generate this dump file, 18 characters]
NSIMSH 1000
[Desired number of simulated showers, max=2**31-1]
TIME 300
[Allotted simulation time, in sec]
....+....1....+....2....+....3....+....4....+....5....+....6....+....7..

The upper plot in fig. 6.4 shows the distribution of energy $E_d$ deposited into the NaI crystal volume (per primary photon). The lower figure displays the distribution (per primary photon) of the energy $E_b$ of “backscattered” photons, i.e. photons that emerge from the system pointing “downwards”, with $W = \cos \theta < 0$. These distributions show three conspicuous structures that arise from backscattering of incident photons in the crystal volume or in the Al backing (B), escape of one of the $\sim 511$ keV x-rays resulting from positron annihilation (A) and escape of $\sim 30$ keV iodine K x-rays (C). The peak A is so small because pair production is a relatively unlikely process for 1.25 MeV photons (the energy is too close to the threshold).

6.3 Selecting the simulation parameters

The speed and accuracy of the simulation of electrons and positrons is determined by the values of the simulation parameters $E_{abs}$, $C_1$, $C_2$, $W_{cc}$, $W_{cr}$ and $s_{max}$, which are selected by the user for each material in the simulated structure\(^2\). Here we summarize the rules for assigning “safe” values to these parameters.

\(^2\)To specify simulation parameters for a single body we can simply assign a specific material to this body, different from that of other bodies of the same composition.
The absorption energies \( E_{\text{abs}} \) are determined either by the characteristics of the experiment or by the required space resolution. If we want to tally dose or deposited-charge distributions, \( E_{\text{abs}} \) should be such that the residual range \( R(\theta, E_{\text{abs}}) \) of electrons/positrons is less than the typical dimensions of the volume bins used to tally these distributions\(^3\). In other cases, it is advisable to run short simulations (for the considered body alone) with increasing values of \( E_{\text{abs}} \) (starting from 100 eV) to study the effect of this parameter on the results.

The allowed values of the elastic scattering parameters \( C_1 \) and \( C_2 \) are limited to the interval \((0, 0.2)\). For the present version 2001 of PENelope, these parameters have a very weak influence on the results, weaker than for previous versions of the code. As discussed in section 4.4.1, this is mostly due to the improved modelling of soft energy losses and to the consideration of the energy dependence of the hard mean free paths (see sections 4.2 and 4.3). Our recommended practice is to set \( C_1 = C_2 = 0.05 \), which is fairly conservative, as shown by the example given below. Before increasing the value of any of these parameters, it is advisable to perform short test simulations to verify that with the augmented parameter value the results remain essentially unaltered (and that the simulation runs faster; if there is no gain in speed, keep the conservative values).

We have already indicated that the cutoff energies \( W_{\text{cc}} \) and \( W_{\text{cr}} \) have a very weak influence on the accuracy of the results provided only that they are both smaller than the width of the bins used to tally energy distributions. When energy distributions are of no interest, our recommendation is setting these cutoff energies equal to one hundredth of the typical energy of primary particles.

The maximum allowed step length \( s_{\text{max}} \) (denoted by \texttt{DSMAX} in the FORTRAN source files) should be less than one tenth of the characteristic thickness of the body where the particle moves. This ensures that, on average, there will be more than 20 soft events\(^4\) (hinges) along a typical electron/positron track within that body, which is enough to “wash out” the details of the artificial distributions used to sample these events. Notice however that PENelope internally forces the step length to be less than \( \sim 3 \lambda_T^{(b)} \) (see section 4.4). Therefore, for thick bodies (thicker than \( \sim 20 \lambda_T^{(b)} \)), we can set \( s_{\text{max}} = 10^{-5} \), or some other very large value, to switch off the external step-length control.

The \texttt{MAIN} program PENSLAB can be readily used to study the effect of the simulation parameters for a material body of a given characteristic thickness. As an example, figs. 6.5 and 6.6 display partial results from a \texttt{PENSLAB} simulation for a parallel electron beam of 500 keV impinging normally on the surface of a 200-\( \mu \text{m} \)-thick aluminium slab. The absorption energies were set equal to 10 keV (for all kinds of particles) and \( W_{\text{cr}} \) was given a negative value, which compels PENelope to set \( W_{\text{cr}} = 10 \) eV and to disregard emission of soft bremsstrahlung (with \( W < 10 \) eV). We ran \texttt{PENSLAB} using \( W_{\text{cc}} = 0 \) and \( C_1 = C_2 = 0 \); in this case, PENelope performs purely detailed, collision by

---

\(^3\)PENelope prints tables of electron and positron ranges if subroutine \texttt{PINIT} is invoked with \texttt{INFO=3} or larger.

\(^4\)PENelope randomizes \( s_{\text{max}} \) in such a way that the actual step lengths never exceeds the value \( s_{\text{max}} \) set by the user and, on average, is equal to \( s_{\text{max}} / 2 \).
6.3. Selecting the simulation parameters

Figure 6.5: Results from PENSLAB for a 500 keV electron beam impinging normally on the surface of a 200 μm Al slab (further details are given in the text). Top: depth-dose distribution within the slab. Bottom: angular distribution of emerging (transmitted and backscattered) electrons (primary and secondary).
Figure 6.6: Results from PENSLAB for a 500 keV electron beam impinging normally on the surface of a 200 μm Al slab (further details are given in the text). Top: energy distribution of transmitted electrons. Bottom: energy distribution of backscattered electrons. Secondary electrons are included in both cases.
collision, simulation and, therefore, it provides exact results (affected only by statistical uncertainties and by inaccuracies of the physical interaction model). Differences between these results and those from mixed simulation are then completely attributable to the approximations in our mixed transport algorithm. To our knowledge, no other high-energy transport code allows detailed simulation and this kind of direct validation of the electron/positron transport mechanics.

In figs. 6.5 and 6.6 we compare results from the detailed simulation (7.5 million showers) with those from a mixed simulation using $W_{\infty} = 1$ keV and $C_1 = C_2 = 0.15$ (20 million simulated showers); the error bars indicate statistical uncertainties (3σ). With these relatively high values of $C_1$ and $C_2$, mixed simulation is quite fast, the speed (generated showers per second) being about 45 times higher than that of the detailed simulation. As shown in the plots, mixed simulation results are practically equivalent to those from detailed simulation. It should be noted that backscattering, fig. 6.6b, is one of the most difficult cases to study, because it involves transport near and across an interface that is far from electronic equilibrium. The only visible artifact is a kind of singularity in the energy distribution of backscattered electrons at $\sim 250$ keV (which averages to the correct value and, therefore, would not be seen in a coarser energy grid). This artifact is also present in the energy distribution of transmitted electrons, but hardly visible in the scale of fig. 6.6a.

\section{The code SHOWER}

Monte Carlo simulation has proven to be a very valuable tool for education. In the past, radiation physics used to be considered as a tough subject, mostly because high-energy radiation is well outside the realm of daily experience. Nowadays, by simply running a transport simulation code on a personal computer we can learn more than from tens of obscure empirical formulas and numerical tables, and eventually “understand” many aspects of radiation transport (those for which we have run the simulation code and “digested” the results).

The PENEOPE distribution package includes a binary file named SHOWER that generates electron-photon showers within a slab (of one of the 279 materials defined in POCOMPOS.TAB) and displays them (projected) on the computer screen plane. The current version operates only under Microsoft Windows 9x. The program is self-explanatory, and requires only a small amount of information from the user, which is entered from the keyboard, in response to prompts from the program. Electron, photon and positron tracks are displayed in different colors and intensities that vary with the energy of the particle. It is worth mentioning that the maximum number of showers that can be plotted in a single shot is limited to 50, because the screen may become too messy. Generating this small number of showers takes a short time, of the order of a few seconds, even on modest personal computers (provided only that the absorption energies are sensibly chosen).

Once on the graphical screen, the view plane can be rotated about the horizontal
screen axis by typing “r” and the rotation angle in degrees; the screen plane can also be rotated progressively, by 15 deg steps, by pressing the “enter” key repeatedly. Entering the single-character command “n” erases the screen and displays a new shower. Observation of single showers projected on a revolving plane gives a truly three-dimensional perspective of the transport process.

6.5 Installation

The FORTRAN77 source files of PENELOPe, the examples and auxiliary programs and the database are distributed as a single ZIP compressed file named PENELOPe.ZIP. To extract the files, keeping the directory structure, create the directory PENELOPe in your hard disk, copy the distribution file PENELOPe.ZIP into this directory and, from there, inflate ( unzip) it. The directory structure and contents of the PENELOPe code system are the following:

```
  PENELOPe
    PROGRAMS
    PENDBASE
    OTHER
    GVIEW
    SHOWER
    EMFIELDS
    PLOTTER
```

**Figure 6.7:** Directory tree of the PENELOPe code system.

- Subdirectory **PROGRAMS**. It contains the following 17 files:

  - **MANUAL.TXT** ... abridged manual with general information.
  - **PENELOPe.F** ... simulation subroutine package.
  - **PENGEOM.F** ... modular quadric geometry subroutine package ( handles systems with up to 250 surfaces and 125 bodies).
  - **PENVARED.F** ... variance-reduction subroutines.
  - **MATERIAL.F** ... main program to generate material data files.
  - **PENSLAB.F** ... main program for particle transport in a slab.
  - **PENSLAB.IN** ... sample input data file of PENSLAB.
  - **AL.MAT** ... Material data file for PENSLAB.
  - **PENCYL.F** ... main program for multilayered cylindrical geometries and axially symmetric beams.
6.5. **Installation**

`PENCYL.IN` ... sample input data file of `PENCYL`. Describes the same geometry as `PENDOSES.GEO`.

`PENDOSES.F` ... main program for arbitrary quadric geometries.

`PENDOSES.IN` ... sample input data file of `PENDOSES`.

`PENDOSES.GEO` ... geometry definition file for `PENDOSES`.

`NAIAL.MAT` ... material data file for `PENCYL` and `PENDOSES`. Illustrates the use of multiple materials.

`TIMERMS.F` ... clock subroutine, based on the subroutine `GETTIM` of Microsoft FORTRAN (version 5.0). It gives the execution time in seconds.

`TIMERG77.F` ... clock subroutine for the g77 FORTRAN compiler from the Free Software Foundation. The compact G77 for Win32 (Windows 9x/NT) package can be downloaded from http://www.geocities.com/Athens/Olympus/5564

`g77` is the default FORTRAN compiler in most Linux distributions.

`TIMER1.F` ... fake clock subroutine. To be used with “unfamiliar” compilers for which a proper timing routine is not known/available.

To get the executable file of `MATERIAL`, compile and link the files `MATERIAL.F` and `PENEOLOPE.F`. This executable file must be placed and run in the same subdirectory as the database files (PENDBASE).

The executable files of `PENSLAB`, `PENCYL` and `PENDOSES` are obtained by compiling and linking the following groups of source files (here we use the clock subroutine of the g77 compiler):

- `PENSLAB`: `PENSLAB.F`, `PENEOLOPE.F`, `TIMERG77.F`
- `PENCYL`: `PENCYL.F`, `PENEOLOPE.F`, `PENVARED.F`, `TIMERG77.F`
- `PENDOSES`: `PENDOSES.F`, `PENEOLOPE.F`, `PENGEOM.F`, `TIMERG77.F`

The simulation programs are written in standard FORTRAN77 language, so that they should run on any computer. The only exception is the clock subroutine `TIMERxxx.F`, which must be adapted to your computer’s compiler.

- Subdirectory `PENDBASE`. PENEOLOPE’s database. 373 files with the extension “.TAB” and names beginning with the letters “PD” (for details, see section 6.1.1).

- Subdirectory `OTHER`. Consists of the following subdirectories,

`GVIEW` ... Contains the geometry viewers `GVIEW2D`, `GVIEW3D` and `GVIEWC`, that are operable only under Microsoft Windows 9x, and several examples of geometry definition files.

`EMFIELDS` ... Contains the subroutine package `PENFIELD.F`, which does simulation of electron/positron transport under external static magnetic and electric fields (see appendix C), and examples of programs that use it.
SHOWER ... Contains a single binary file named SHOWER.EXE, which operates only under Microsoft Windows 9x. This code generates electron-photon showers within a slab and displays them projected on the screen. To use the SHOWER viewer, just copy the file SHOWER.EXE into the directory PENDBASE and run it from there. This little tool is particularly useful for teaching purposes, it makes radiation physics “visible”.

PLOTTER ... The programs PENSLAB and PENCYL generate multiple files with simulated probability distribution functions. Each output file has a heading describing its content, which is in a format ready for visualization with a plotting program. We use GNUPLOT, which is small in size, available for various platforms (including Linux and Microsoft Windows) and free (distribution sites are listed at the Gnuplot Central site, http://www.gnuplot.org). The directory PLOTTER contains GNUPLOT scripts that plot the probability distributions evaluated by the simulation codes on your terminal. For instance, after running PENSLAB you can visualize the results by simply 1) copying the file PENSLAB.GNU from the directory PLOTTER to the directory that contains the results and 2) entering the command “GNUPLOT PENSLAB.GNU” (or clicking the icon).
Appendix A

Collision kinematics

To cover the complete energy range of interest in radiation transport studies we use relativistic kinematics. Let $\vec{P}$ denote the energy-momentum 4-vector of a particle, i.e.

$$\vec{P} = (\mathcal{W}c^{-1}, \mathbf{p}),$$

(A.1)

where $\mathcal{W}$ and $\mathbf{p}$ are the total energy (including the rest energy) and momentum respectively and $c$ is the velocity of light in vacuum. The product of 4-vectors, defined by

$$(\vec{P} \vec{P}') = \mathcal{W}\mathcal{W}'c^{-2} - \mathbf{p} \cdot \mathbf{p'},$$

(A.2)

is invariant under Lorentz transformations. The rest mass $m$ of a particle determines the invariant length of its energy-momentum,

$$(\vec{P} \vec{P}) = \mathcal{W}^2c^{-2} - \mathbf{p}^2 = (mc)^2.$$  

(A.3)

The kinetic energy $E$ of a massive particle ($m \neq 0$) is defined as

$$E = \mathcal{W} - mc^2,$$

(A.4)

where $mc^2$ is the rest energy. The magnitude of the momentum is given by

$$(cp)^2 = E(E + 2mc^2).$$

(A.5)

In terms of the velocity $\mathbf{v}$ of the particle, we have

$$E = (\gamma - 1)mc^2 \quad \text{and} \quad \mathbf{p} = \beta \gamma mc\hat{\mathbf{v}},$$

(A.6)

where

$$\beta \equiv \frac{v}{c} = \sqrt{\frac{\gamma^2 - 1}{\gamma^2}} = \sqrt{\frac{E + 2mc^2}{(E + mc^2)^2}}$$

(A.7)

is the velocity of the particle in units of $c$ and

$$\gamma \equiv \sqrt{\frac{1}{1 - \beta^2}} = \frac{E + mc^2}{mc^2}$$

(A.8)
is the total energy in units of the rest energy. From the relation (A.5), it follows that

\[ E = \sqrt{(cp)^2 + m^2c^4 - mc^2} \]  
(A.9)

and

\[ \frac{dp}{dE} = \frac{1}{v} = \frac{1}{c\beta}. \]  
(A.10)

For a photon (and any other particle with \( m = 0 \)), the energy and momentum are related by

\[ E = cp. \]  
(A.11)

A.1 Two-body reactions

Consider a reaction in which a projectile "1" collides with a target "2" initially at rest in the laboratory frame of reference. We limit our study to the important case of two-body reactions in which the final products are two particles, "3" and "4". The kinematics of such reactions is governed by energy and momentum conservation.

We take the direction of movement of the projectile to be the \( z \)-axis, and set the \( x \)-axis in such a way that the reaction plane (i.e. the plane determined by the momenta of particles "1", "3" and "4") is the \( x-z \) plane. The energy-momentum 4-vectors of the projectile, the target and the reaction products are then (see fig. A.1)

\[ \vec{P}_1 = (W_1c^{-1}, 0, 0, p_1) \]  
(A.12a)
\[ \vec{P}_2 = (m_2c, 0, 0, 0) \]  
(A.12b)
\[ \vec{P}_3 = (W_3c^{-1}, p_3 \sin \theta_3, 0, p_3 \cos \theta_3) \]  
(A.12c)
\[ \vec{P}_4 = (W_4c^{-1}, -p_4 \sin \theta_4, 0, p_4 \cos \theta_4) \]  
(A.12d)

Energy and momentum conservation is expressed by the 4-vector equation

\[ \vec{P}_1 + \vec{P}_2 = \vec{P}_3 + \vec{P}_4. \]  
(A.13)

From this equation, the angles of emergence of the final particles, \( \theta_3 \) and \( \theta_4 \), are uniquely determined by their energies, \( W_3 \) and \( W_4 \). Thus,

\[ m_1^2c^2 = (\vec{P}_4 \cdot \vec{P}_4) = (\vec{P}_1 + \vec{P}_2 - \vec{P}_3)(\vec{P}_1 + \vec{P}_2 - \vec{P}_3) \]
\[ = (\vec{P}_1 \cdot \vec{P}_1) + (\vec{P}_2 \cdot \vec{P}_2) + (\vec{P}_3 \cdot \vec{P}_3) + 2(\vec{P}_1 \cdot \vec{P}_2) - 2(\vec{P}_1 \cdot \vec{P}_3) - 2(\vec{P}_2 \cdot \vec{P}_3) \]
\[ = m_1^2c^2 + m_2^2c^2 + m_3^2c^2 + 2W_1W_2c^{-2} \]
\[ - 2(W_1W_3c^{-2} - p_1p_3 \cos \theta_3) - 2W_2W_3c^{-2}, \]  
(A.14)

and it follows that

\[ \cos \theta_3 = \frac{m_1^2c^4 - m_2^2c^4 - m_3^2c^4 + 2W_1(W_3 - W_2) + 2W_2W_3}{2(W_1^2 - m_1^2c^4)^{1/2} (W_2^2 - m_2^2c^4)^{1/2}}. \]  
(A.15)
A.1. Two-body reactions

Clearly, by symmetry, we can obtain a corresponding expression for \( \cos \theta_4 \) by interchanging the indices 3 and 4

\[
\cos \theta_4 = \frac{m_2^2 c^4 - m_1^2 c^4 - m_3^2 c^4 - m_4^2 c^4 + 2W_1(W_4 - W_2) + 2W_2W_4}{2(W_1^2 - m_1^2 c^4)^{1/2}(W_4^2 - m_4^2 c^4)^{1/2}}.
\]  

(A.16)

\[
\cos \theta_3 = \left( \frac{E - W}{E} \frac{E + 2m_e c^2}{E - W + 2m_e c^2} \right)^{1/2},
\]

\[
\cos \theta_4 = \left( \frac{W}{E} \frac{E + 2m_e c^2}{W + 2m_e c^2} \right)^{1/2}.
\]

(A.17) (A.18)

- Compton scattering of photons by free electrons at rest.

  - Projectile: Photon \( m_1 = 0, \quad W_1 = E \equiv \kappa m_e c^2 \).
  - Target: Electron \( m_2 = m_e, \quad W_2 = m_e c^2 \).
  - Scattered photon: \( m_3 = 0, \quad W_3 \equiv \tau E \).
  - Recoil electron: \( m_4 = m_e, \quad W_4 = m_e c^2 + (1 - \tau)E \).
Appendix A. Collision kinematics

\[ \cos \theta_3 = \frac{1}{\kappa} \left( \kappa + 1 - \frac{1}{\tau} \right), \]  \hspace{1cm} (A.19)  

\[ \cos \theta_4 = (\kappa + 1) \left( \frac{1 - \tau}{\kappa [2 + \kappa(1 - \tau)]} \right)^{1/2}. \]  \hspace{1cm} (A.20)

- Annihilation of positrons with free electrons at rest.

  Projectile: \hspace{1cm} Positron \hspace{0.5cm} m_1 = m_e, \hspace{0.5cm} W_1 = E + m_e c^2 = \gamma m_e c^2.

  Target: \hspace{1cm} Electron \hspace{0.5cm} m_2 = m_e, \hspace{0.5cm} W_2 = m_e c^2.

  Annihilation photons: \hspace{1cm} m_3 = 0, \hspace{0.5cm} W_3 = \zeta (E + 2m_e c^2).

  m_4 = 0, \hspace{0.5cm} W_4 = (1 - \zeta) (E + 2m_e c^2).

\[ \cos \theta_3 = \left( \gamma^2 - 1 \right)^{-1/2} (\gamma + 1 - 1/\zeta), \]  \hspace{1cm} (A.21)  

\[ \cos \theta_4 = \left( \gamma^2 - 1 \right)^{-1/2} \left( \gamma + 1 - \frac{1}{1 - \zeta} \right). \]  \hspace{1cm} (A.22)

A.1.1 Elastic scattering

By definition, elastic collisions keep the internal structure (i.e. the mass) of the projectile and target particles unaltered. Let us consider the kinematics of elastic collisions of a projectile of mass \( m = m_1 = m_3 \) and kinetic energy \( E \) with a target particle of mass \( M = m_2 = m_4 \) at rest (see fig. A.2). After the interaction, the target recoils with a certain kinetic energy \( W \) and the kinetic energy of the projectile is reduced to \( E' = E - W \). The angular deflection of the projectile \( \cos \theta \) and the energy transfer \( W \) are related through eq. (A.15), which now reads

\[ \cos \theta = \frac{E (E + 2mc^2) - W (E + mc^2 + Mc^2)}{\sqrt{E (E + 2mc^2) (E - W) (E - W + 2mc^2)}}. \]  \hspace{1cm} (A.23)

The target recoil direction is given by eq. (A.16),

\[ \cos \theta_r = \frac{(E + mc^2 + Mc^2) W}{\sqrt{E (E + 2mc^2) W (W + 2mc^2)}}. \]  \hspace{1cm} (A.24)

Solving eq. (A.23), we obtain the following expression for the energy transfer \( W \) corresponding to a given scattering angle \( \theta \),

\[ W = \left[ (E + mc^2) \sin^2 \theta + Mc^2 - \cos \theta \sqrt{M^2 c^4 - m^2 c^4 \sin^2 \theta} \right] \]

\[ \times \frac{E (E + 2mc^2)}{(E + mc^2 + Mc^2)^2 - E (E + 2mc^2) \cos^2 \theta}. \]  \hspace{1cm} (A.25)
In the case of collisions of particles with equal mass, \( m = M \), this expression simplifies to

\[
W = \frac{E(E + 2mc^2) \sin^2 \theta}{E \sin^2 \theta + 2mc^2} \quad \text{if } M = m. \quad (A.26)
\]

In this case, \( \theta \) can only take values less than 90 deg. For \( \theta = 90 \) deg, we have \( W = E \) (i.e. the full energy and momentum of the projectile are transferred to the target). Notice that for binary collisions of electrons and positrons \( (m = m_e) \), the relation (A.26) becomes identical to (A.17).

For elastic collisions of electrons by atoms and ions, the mass of the target is much larger than that of the projectile and eq. (A.25) becomes

\[
W = \frac{[(E + mc^2) \sin^2 \theta + Mc^2(1 - \cos \theta)] E(E + 2mc^2)}{(E + Mc^2)^2 - E(E + 2mc^2) \cos^2 \theta} \quad \text{if } M \gg m. \quad (A.27)
\]

The non-relativistic limit \( (c \to \infty) \) of this expression is

\[
W = \frac{2m}{M}(1 - \cos \theta)E \quad \text{if } M \gg m \text{ and } E \ll mc^2. \quad (A.28)
\]

### A.2 Inelastic collisions of charged particles

We consider here the kinematics of inelastic collisions of charged particles of mass \( m \) and velocity \( \mathbf{v} \) as seen from a frame of reference where the stopping medium is at rest (laboratory frame). Let \( \mathbf{p} \) and \( E \) be the momentum and the kinetic energy of the projectile just before an inelastic collision, the corresponding quantities after the collision are denoted by \( \mathbf{p}' \) and \( E' = E - W \), respectively. Evidently, for positrons the maximum energy loss is \( W_{\text{max}} = E \). In the case of ionization by electron impact, owing to the indistinguishability between the projectile and the ejected electron, the maximum energy loss is \( W_{\text{max}} \approx E/2 \) (see section 3.2). The momentum transfer in the collision is \( \mathbf{q} \equiv \mathbf{p} - \mathbf{p}' \). It is customary to introduce the recoil energy \( Q \) defined by

\[
Q(Q + 2m_e c^2) = (eq)^2 = c^2 \left( p^2 + p'^2 - 2pp' \cos \theta \right), \quad (A.29)
\]
where \(m_e\) is the electron rest mass and \(\theta = \arccos(\mathbf{p} \cdot \mathbf{p}')\) is the scattering angle. Equivalently, we can write
\[
Q = \sqrt{(cp)^2 + m_e^2c^4} - m_e c^2.
\] (A.30)

Notice that, when the collision is with a free electron at rest, the energy loss is completely transformed into kinetic energy of the recoiling electron, i.e. \(Q = W\). For collisions with bound electrons, the relation \(Q \approx W\) still holds for hard ionizing collisions (that is, when the energy transfer \(W\) is much larger than the ionization energy of the target electron so that binding effects are negligible).

The kinematically allowed recoil energies lie in the interval \(Q_- < Q < Q_+\), with end points given by eq. (A.29) with \(\cos \theta = +1\) and \(-1\), respectively. That is
\[
Q_\pm = \sqrt{(cp \pm cp')^2 + m_e^2c^4} - m_e c^2
\]
\[
= \sqrt{[E(E + 2mc^2)]^2 + (E - W)(E - W + 2mc^2)]^2} + m_e^2c^4 - m_e c^2.\] (A.31)

Notice that, for \(W < E\), \(Q_+\) is larger than \(W\) and \(Q_- < W\). When \(W \ll E\), expression (A.31) is not suitable for evaluating \(Q_-\) since it involves the subtraction of two similar quantities. In this case, it is more convenient to use the approximate relation
\[
 cp - cp' \simeq c \left( -\frac{dp}{dE} W + \frac{1}{2} \frac{dp^2}{dE^2} W^2 \right) = \frac{W}{\beta} \left( 1 + \frac{1}{2} \frac{1}{\gamma^2 - 1} \frac{W}{E + mc^2} \right)\] (A.32)

and calculate \(Q_-\) as
\[
Q_- \simeq \sqrt{(cp - cp')^2 + m_e^2c^4} - m_e c^2\] (A.33)
or, if \(cp - cp' \ll m_e c^2\),
\[
Q_- \simeq \frac{1}{2} \frac{(cp - cp')^2}{m_e c^2} - \frac{1}{8} \frac{(cp - cp')^4}{(m_e c^2)^3}.\] (A.34)

Thus, for \(E \gg W\),
\[
Q_-(Q_- + 2m_e c^2) \approx W^2 / \beta^2.\] (A.35)

In the non-relativistic limit,
\[
Q \equiv q^2/2m_e, \quad Q_\pm = \left[ E^{1/2} \pm (E - W)^{1/2} \right]^2.\] (A.36)

From (A.31), it is clear that the curves \(Q = Q_- (W)\) and \(Q = Q_+ (W)\) vary monotonously with \(W\) and intersect at \(W = E\). Thus, they define a single continuous function \(W = W_m(Q)\) in the interval \(0 < Q < Q_+(0)\). By solving the eqs. \(Q = Q_\pm (W_m)\) we obtain
\[
W_m(Q) = E + mc^2 - \sqrt{\left[ \sqrt{E(E + 2mc^2)} - \sqrt{Q(Q + 2m_e c^2)} \right]^2 + m_e^2 c^4},\] (A.37)
Figure A.3: Domains of kinematically allowed transitions in the \((Q,W)\) plane for electrons/positrons. The curves represent the maximum allowed energy loss \(W_m(Q)\), given by eq. (A.37), for electrons with the indicated kinetic energies (in eV). When \(E\) increases, \(W_m(Q)\) approaches the vacuum photon line, \(W = [Q(Q + 2m_e c^2)]^{1/2}\), which is an absolute upper bound for the allowed energy losses.

which, when \(W \ll E\), reduces to

\[
W_m(Q) \simeq \beta \sqrt{Q(Q + 2m_e c^2)}. \quad (A.38)
\]

Now it follows that, for given values of \(E\) and \(Q < Q_+(0)\), the only kinematically allowed values of the energy loss are those in the interval \(0 < W < W_m(Q)\) (see fig. A.3).

For a given energy loss \(W\), the quantity

\[
q_{\text{min}} \equiv c^{-1} \sqrt{Q_-(Q_- + 2m_e c^2)}, \quad (A.39)
\]

is the minimum value of the momentum transfer in an inelastic collision, which occurs when \(\theta = 0\). \(q_{\text{min}}\) is always larger than \(W/c\). When the energy of the projectile increases, \(\beta \to 1\) and \(q_{\text{min}}\) decreases approaching (but never reaching) the value \(W/c\). It is worth recalling that a photon of energy \(W\) in vacuum has a linear momentum \(q = W/c\) and, hence, interactions consisting of emission of bare photons would be located on the line \(Q(Q + 2m_e c^2) = W^2\) of the \((Q,W)\) plane, the so-called vacuum photon line. This line, lies outside the kinematically allowed region, i.e. the “recoil” energy of the photon is less than \(Q_-\) (see fig. A.3). Therefore, when the target is a single atom, the emission of
Appendix A. Collision kinematics

photons by the projectile is not possible\(^1\). When the energy \(E\) of the projectile increases, \(Q_-\) decreases and tends to the photon line when \(\beta\) tends to unity. Hence, emission of photons by ultrarelativistic projectiles in low-density media is barely prevented by energy and momentum conservation. Generally speaking, as the interaction involves the exchange of a virtual photon, the DCS increases as the photon becomes more real, that is as we approach the photon line. For a dilute gas, this causes a gradual increase of the cross section with the projectile energy when \(\beta \to 1\).

The scattering angle \(\theta\) is related to the energy loss through

\[
\cos \theta = \frac{(cp)^2 + (cp')^2 - Q(Q + 2me^2)}{2(cp)(cp')}. \tag{A.40}
\]

The recoil angle \(\theta_r\) between \(p\) and \(q\) is given by

\[
\cos \theta_r = \frac{(cp)^2 - (cp')^2 + (cq)^2}{2(cp)(cq)}, \tag{A.41}
\]

which can also be written in the form

\[
\cos^2 \theta_r = \frac{W^2/\beta^2}{Q(Q + 2me^2)} \left( 1 + \frac{Q(Q + 2me^2) - W^2}{2W(E + me^2)} \right)^2. \tag{A.42}
\]

For heavy \((m \gg m_e)\) high-energy projectiles and collisions such that \(Q \ll E\) and \(W \ll E\),

\[
\cos^2 \theta_r \sim \frac{W^2/\beta^2}{Q(Q + 2me^2)} \sim \frac{Q_-(Q_+ + 2me^2)}{Q(Q + 2me^2)} \tag{A.43}
\]

\(^1\)In a condensed medium, ultrarelativistic projectiles can emit real photons (Cerenkov radiation) under certain, quite restricting circumstances (see e.g. Jackson, 1975).
Appendix B

Numerical tools

B.1 Cubic spline interpolation

In this section we follow the presentation of Maron (1982). Suppose that a function \( f(x) \) is given in numerical form, i.e. as a table of values

\[
    f_i = f(x_i) \quad (i = 1, \ldots, N).
\]  

(B.1)

The points (knots) \( x_i \) do not need to be equispaced, but we assume that they are in (strictly) increasing order

\[
    x_1 < x_2 < \cdots < x_N.
\]  

(B.2)

A function \( \varphi(x) \) is said to be an interpolating cubic spline if

1) It reduces to a cubic polynomial within each interval \([x_i, x_{i+1}]\), i.e. if \( x_i \leq x \leq x_{i+1} \)

\[
    \varphi(x) = a_i + b_i x + c_i x^2 + d_i x^3 \equiv p_i(x) \quad (i = 1, \ldots, N - 1).
\]  

(B.3)

2) The polynomial \( p_i(x) \) matches the values of \( f(x) \) at the endpoints of the \( i \)-th interval,

\[
    p_i(x_i) = f_i, \quad p_i(x_{i+1}) = f_{i+1} \quad (i = 1, \ldots, N - 1),
\]  

(B.4)

so that \( \varphi(x) \) is continuous in \([x_1, x_N]\).

3) The first and second derivatives of \( \varphi(x) \) are continuous in \([x_1, x_N]\)

\[
    p_i'(x_{i+1}) = p_{i+1}'(x_{i+1}) \quad (i = 1, \ldots, N - 2),
\]  

(B.5)

\[
    p_i''(x_{i+1}) = p_{i+1}''(x_{i+1}) \quad (i = 1, \ldots, N - 2).
\]  

(B.6)

Consequently, the curve \( y = \varphi(x) \) interpolates the table (B.1) and has a continuously turning tangent.
To obtain the spline coefficients $a_i$, $b_i$, $c_i$, $d_i$ ($i = 1, \ldots, N - 1$) we start from the fact that $\varphi''(x)$ is linear in $[x_i, x_{i+1}]$. Introducing the quantities
\[ h_i \equiv x_{i+1} - x_i \quad (i = 1, \ldots, N - 1) \] (B.7)
and
\[ \sigma_i \equiv \varphi''(x_i) \quad (i = 1, \ldots, N), \] (B.8)
we can write the obvious identity
\[ p_i''(x) = \sigma_i \frac{x_{i+1} - x}{h_i} + \sigma_{i+1} \frac{x - x_i}{h_i} \quad (i = 1, \ldots, N - 1). \] (B.9)
Notice that $x_{i+1}$ must be larger than $x_i$ in order to have $h_i > 0$. Integrating eq. (B.9) twice with respect to $x$, gives for $i = 1, \ldots, N - 1$
\[ p_i(x) = \frac{\sigma_i (x_{i+1} - x)^3}{6h_i} + \frac{\sigma_{i+1} (x - x_i)^3}{6h_i} + A_i (x - x_i) + B_i (x_{i+1} - x), \] (B.10)
where $A_i$ and $B_i$ are constants. These can be determined by introducing the expression (B.10) into eqs. (B.4), this gives the pair of eqs.
\[ \sigma_i \frac{h_i^2}{6} + B_i h_i = f_i \quad \text{and} \quad \sigma_{i+1} \frac{h_i^2}{6} + A_i h_i = f_{i+1}. \] (B.11)
Finally, solving for $A_i$ and $B_i$ and substituting the result in (B.10), we obtain
\[ p_i(x) = \frac{\sigma_i}{6} \left[ \frac{(x_{i+1} - x)^3}{h_i} - h_i (x_{i+1} - x) \right] + f_i \frac{x_{i+1} - x}{h_i} + \frac{\sigma_{i+1}}{6} \left[ \frac{(x - x_i)^3}{h_i} - h_i (x - x_i) \right] + f_{i+1} \frac{x - x_i}{h_i}. \] (B.12)

To be able to use $\varphi(x)$ to approximate $f(x)$, we must find the second derivatives $\sigma_i$ ($i = 1, \ldots, N$). To this end, we impose the conditions (B.5). Differentiating (B.12) gives
\[ p_i'(x) = \frac{\sigma_i}{6} \left[ -3(x_{i+1} - x)^2 + h_i \right] + \frac{\sigma_{i+1}}{6} \left[ 3(x - x_i)^2 - h_i \right] + \delta_i, \] (B.13)
where
\[ \delta_i = \frac{y_{i+1} - y_i}{h_i}. \] (B.14)
Hence,
\[ p_i'(x_{i+1}) = \frac{\sigma_i h_i}{6} + \frac{\sigma_{i+1} h_i}{3} + \delta_i, \] (B.15a)
\[ p_i'(x_i) = -\frac{\sigma_i h_i}{3} - \frac{\sigma_{i+1} h_i}{6} + \delta_i \] (B.15b)
and, similarly,
\[ p_{i+1}'(x_{i+1}) = -\sigma_{i+1}\frac{h_{i+1}}{3} - \sigma_{i+2}\frac{h_{i+1}}{6} + \delta_{i+1}. \]
(B.15c)

Replacing (B.15a) and (B.15c) in (B.5), we obtain
\[ h_i \sigma_i + 2(h_i + h_{i+1}) \sigma_{i+1} + h_{i+1} \sigma_{i+2} = 6(\delta_{i+1} - \delta_i) \quad (i = 1, \ldots, N - 2). \]
(B.16)

The system (B.16) is linear in the \( N \) unknowns \( \sigma_i \) \( (i = 1, \ldots, N) \). However, since it contains only \( N - 2 \) equations, it is underdetermined. This means that we need either to add two additional (independent) equations or to fix arbitrarily two of the \( N \) unknowns. The usual practice is to adopt endpoint strategies that introduce constraints on the behaviour of \( \varphi(x) \) near \( x_1 \) and \( x_N \). An endpoint strategy fixes the values of \( \sigma_1 \) and \( \sigma_N \), yielding an \( (N - 2) \times (N - 2) \) system in the variables \( \sigma_i \) \( (i = 2, \ldots, N - 1) \). The resulting system is, in matrix form,
\[
\begin{pmatrix}
H_2 & h_2 & 0 & \cdots & 0 & 0 & 0 \\
\vdots & h_3 & h_3 & \cdots & 0 & 0 & 0 \\
0 & h_3 & H_4 & \cdots & 0 & 0 & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\
0 & 0 & 0 & \cdots & h_{N-3} & h_{N-3} & 0 \\
0 & 0 & 0 & \cdots & h_{N-3} & h_{N-2} & h_{N-2} \\
0 & 0 & 0 & \cdots & h_{N-3} & H_{N-1} & h_{N-1} \\
0 & 0 & 0 & \cdots & 0 & h_{N-2} & H_{N-1}
\end{pmatrix}
\begin{pmatrix}
\sigma_2 \\
\sigma_3 \\
\sigma_4 \\
\vdots \\
\sigma_{N-3} \\
\sigma_{N-2} \\
\sigma_{N-1} \\
\sigma_N
\end{pmatrix}
= \begin{pmatrix}
D_2 \\
D_3 \\
D_4 \\
\vdots \\
D_{N-3} \\
D_{N-2} \\
D_{N-1}
\end{pmatrix},
\]
(B.17)

where
\[ H_i = 2(h_{i-1} + h_i) \quad (i = 2, \ldots, N - 1) \]
and
\begin{align*}
D_2 & = 6(\delta_2 - \delta_1) - h_1 \sigma_1 \\
D_i & = 6(\delta_i - \delta_{i-1}) \quad (i = 3, \ldots, N - 2) \\
D_{N-1} & = 6(\delta_{N-1} - \delta_{N-2}) - h_{N-1} \sigma_N.
\end{align*}
(B.19)

\( (\sigma_1 \text{ and } \sigma_N \text{ are removed from the first and last equations, respectively). The matrix of}}

coefficients is symmetric, tridiagonal and diagonally dominant (the larger coefficients are in the diagonal), so that the system (B.17) can be easily (and accurately) solved by Gauss elimination. The spline coefficients \( a_i, b_i, c_i, d_i \) \( (i = 1, \ldots, N - 1) \) —see eq. (B.3)— can then be obtained by expanding the expressions (B.12):

\begin{align*}
a_i & = \frac{1}{6h_i} \left[ \sigma_i x_{i+1}^3 - \sigma_{i+1} x_i^3 + 6\left( f_i x_{i+1} - f_{i+1} x_i \right) \right] + \frac{h_i}{6} (\sigma_{i+1} x_i - \sigma_i x_{i+1}), \\
b_i & = \frac{1}{2h_i} \left[ \sigma_i x_i^2 - \sigma_{i+1} x_{i+1}^2 + 2\left( f_{i+1} - f_i \right) \right] + \frac{h_i}{6} (\sigma_i - \sigma_{i+1}), \\
c_i & = \frac{1}{2h_i} (\sigma x_{i+1} - \sigma_{i+1} x_i), \\
d_i & = \frac{1}{6h_i} (\sigma_{i+1} - \sigma_i).
\end{align*}
(B.20)
When accurate values of \( f''(x) \) are known, the best strategy is to set \( \sigma_1 = f''(x_1) \) and \( \sigma_N = f''(x_N) \), since this will minimize the spline interpolation errors near the endpoints \( x_1 \) and \( x_N \). Unfortunately, the exact values \( f''(x_1) \) and \( f''(x_N) \) are not always available.

The so-called natural spline corresponds to taking \( \sigma_1 = \sigma_N = 0 \). It results in a \( y = \varphi(x) \) curve with the shape that would be taken by a flexible rod (such as a draughtman’s spline) if it were bent around pegs at the knots but allowed to maintain its natural (straight) shape outside the interval \( [x_1, x_N] \). Since \( \sigma_1 = \sigma_N = 0 \), extrapolation of \( \varphi(x) \) outside the interval \( [x_1, x_N] \) by straight segments gives a continuous function with continuous first and second derivatives [i.e. a cubic spline in \((-\infty, \infty)\)].

The accuracy of the spline interpolation is mainly determined by the density of knots in the regions where \( f(x) \) has strong variations. For constant, linear, quadratic and cubic functions the interpolation errors can be reduced to zero by using the exact values of \( \sigma_1 \) and \( \sigma_N \) (in these cases, however, the natural spline may introduce appreciable errors near the endpoints). It is important to keep in mind that a cubic polynomial has, at most, one inflexion point. As a consequence, we should have at least a knot between each pair of inflexion points of \( f(x) \) to ensure proper interpolation. Special care must be taken when interpolating functions that have a practically constant value in a partial interval, since the spline tends to wiggle instead of staying constant. In this particular case, it may be more convenient to use linear interpolation.

Obviously, the interpolating cubic spline \( \varphi(x) \) can be used not only to obtain interpolated values of \( f(x) \) between the knots, but also to calculate integrals such as

\[
\int_a^b f(x) \, dx \simeq \int_a^b \varphi(x) \, dx, \quad x_1 \leq a \quad \text{and} \quad b \leq x_N,
\]

analytically. It is worth noting that derivatives of \( \varphi(x) \) other than the first one may differ significantly from those of \( f(x) \).

To obtain the interpolated value \( \varphi(x_c) \) —see eq. (B.3)— of \( f(x) \) at the point \( x_c \), we must first determine the interval \( [x_i, x_{i+1}] \) that contains the point \( x_c \). To reduce the effort to locate the point, we use the following binary search algorithm:

(i) Set \( i = 1 \) and \( j = N \).

(ii) Set \( k = [(i + j)/2] \).

(iii) If \( x_k < x_c \), set \( i = k \); otherwise set \( j = k \).

(iv) If \( j - i > 1 \), go to step (ii).

(v) Deliver \( i \).

Notice that the maximum delivered value of \( i \) is \( N - 1 \).
B.2 Numerical quadrature

In many cases, we need to calculate integrals of the form

$$\int_{A}^{B} f(z) \, dz,$$

(B.22)

where the integrand is coded as an external function subprogram, which gives nominally exact values. These integrals are evaluated by using the FORTRAN 77 external function SUMGA, which implements the twenty-point Gauss method with an adaptive bipartition scheme to allow for error control. This procedure is comparatively fast and is able to deal even with functions that have integrable singularities located at the endpoints of the interval \([A, B]\), a quite exceptional feature.

B.2.1 Gauss integration

We use the twenty-point Gauss formula (see e.g. Abramowitz and Stegun, 1974), given by

$$\int_{a}^{b} f(z) \, dz = \frac{b - a}{2} \sum_{i=1}^{20} w_i f(z_i)$$

(B.23)

with

$$z_i = \frac{b - a}{2} x_i + \frac{b + a}{2}.$$  

(B.24)

The abscissa \(x_i \ (-1 < x_i < 1)\) is the \(i\)-th zero of the Legendre polynomial \(P_{20}(x)\), the weights \(w_i\) are defined as

$$w_i = \frac{2}{(1 - x_i^2) [P'_{20}(x_i)]^2}.$$  

(B.25)

The numerical values of the abscissas and weights are given in table B.1. The difference between the exact value of the integral and the right-hand side of eq. (B.23) is

$$\Delta_{20} = \frac{(b - a)^4 (20)!^4}{41 \cdot 40!^3} f^{(40)}(\xi),$$

(B.26)

where \(\xi\) is a point in the interval \([a, b]\).

The Gauss method gives an estimate of the integral of \(f(z)\) over the interval \([a, b]\), which is obtained as a weighted sum of function values at fixed points inside the interval. We point out that (B.23) is an open formula, i.e. the value of the function at the endpoints of the interval is never required. Owing to this fact, function SUMGA can integrate functions that are singular at the endpoints. As an example, the integral of \(f(x) = x^{-1/2}\) over the interval \([0,1]\) is correctly evaluated. This would not be possible with a method based on a closed formula (i.e. one that uses the values of the integrand at the interval endpoints).
Table B.1: Abscissas and weights for twenty-point Gauss integration.

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<th>$\pm x_i$</th>
<th>$w_i$</th>
</tr>
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</table>

B.2.2 Adaptive bipartition

Function SUMGA exploits the fact that the error $\Delta_{2n}$, eq. (B.26), of the calculated integral decreases when the interval length is reduced. Thus, halving the interval and applying the Gauss method to each of the two subintervals gives a much better estimate of the integral, provided only that the function $f(x)$ is smooth enough over the initial interval. Notice that the error decreases by a factor of about $2^{-40(1)}$.

The algorithm implemented in SUMGA is as follows. The integration interval $(A, B)$ is successively halved so that each iteration gives a doubly finer partition of the initial interval. We use the term “$n$-subinterval” to denote the subintervals obtained in the $n$-th iteration. In each iteration, the integrals over the different $n$-subintervals are evaluated by the Gauss method, eq. (B.23). Consider that the integral over a given $n$-subinterval is $S_1$. In the following iteration, this $n$-subinterval is halved and the integrals over each of the two resulting $(n+1)$-subintervals are evaluated, giving values $S_{1a}$ and $S_{1b}$. If $S'_1 = S_{1a} + S_{1b}$ differs from $S_1$ in less than the selected tolerance, $S'_1$ is the sought value of the integral in the considered $n$-subinterval; the value $S'_1$ is then accumulated and this $n$-subinterval is no longer considered in subsequent iterations. Each iteration is likely to produce new holes (eliminated subintervals) in the regions where the function is smoother and, hence, the numerical effort progressively concentrates in the regions where $f(x)$ has stronger variations. The calculation terminates when the exploration of the interval $(A, B)$ has been successfully completed or when a clear indication of an anomalous behaviour of $f(x)$ is found (e.g. when there is a persistent increase of the number of remaining $n$-subintervals in each iteration). In the second case a warning message is printed in unit 6 and the control is returned to the calling program.
Appendix C

Electron/positron transport in electromagnetic fields

In this appendix, we consider the transport of electrons/positrons in static external electromagnetic (EM) fields, in vacuum and in condensed media. We assume that, in the region where particles move, there is an electric field $\mathbf{E}$ and a magnetic field $\mathbf{B}$, which are set up by external sources and do not vary with time. For practical purposes, we also consider that both $\mathbf{E}$ and $\mathbf{B}$ are continuous functions of the position vector $\mathbf{r}$.

The interactions with the medium will be described by means of PENEOPE. In each individual interaction event, the electron/positron looses a discrete amount of kinetic energy and changes its direction of motion. In the absence of EM fields, the electron travels freely between consecutive interaction events, i.e. following a straight trajectory segment at constant speed. To simulate electron transport with static external EM fields, we assume that the interaction properties of electrons with the medium are not substantially affected by the field. Consequently, to account for the effect of the EM field, we only need to consider that along each “free flight” the electron is driven by the EM force. With a proper selection of the simulation parameters (i.e. the energy loss and angular cutoff values), trajectory segments may have macroscopic lengths. Therefore, in material media it is appropriate to consider the macroscopic EM fields $\mathbf{D}$ and $\mathbf{H}$ rather than the microscopic fields $\mathbf{E}$ and $\mathbf{B}$.

It should be noted that, under the action of an electric field, the kinetic energy of the electron can vary substantially along a single trajectory segment. This conflicts with one of the basic assumptions in PENEOPE, namely that the energy of the particle stays practically constant along the segment. In practice, however, we can always limit the maximum segment length by means of the parameter $s_{\text{max}}$. Then, the effect of the EM field can be treated independently of that of the interactions with the medium. In other words, for simulation purposes, we only need an efficient method to generate particle trajectories in the EM field in vacuum. It is also important to recall that strong electric fields in material media accelerate unbound charged particles, even when they are at rest (i.e. electrons are never absorbed, simulated tracks can only terminate when
they leave the field volume). Injection of a single electron in the medium may give rise to a complex cascade of delta rays, that accelerate in the direction opposite to the electric field. To describe these cascades we need accurate cross sections for ionization by impact of low-energy electrons, much more accurate than the simple ones implemented in PENEOPE. Therefore, PENEOPE is not expected to yield a reliable description of this process. The simulation algorithm described here is applicable only to magnetic fields and, cautiously, to weak electric fields.

## C.1 Tracking particles in vacuum.

Let us begin by describing a “brute force” method to calculate trajectories of charged particles in arbitrary static electric and magnetic fields in vacuum. We start from the Lorentz force equation\(^1\) for an electron \((Z_0 = -1)\) or positron \((Z_0 = +1)\),

\[
\frac{dp}{dt} = Z_0 e (E + \mathbf{v} \times B), \tag{C.1}
\]

which we write as

\[
\frac{d(\gamma \beta \hat{v})}{dt} = \frac{Z_0 e}{m_e c} (E + c \beta \hat{v} \times B), \tag{C.2}
\]

with \(\hat{v} = \mathbf{v} / v\), \(\beta = v / c\) and \(\gamma = (1 - \beta^2)^{-1/2}\). We note that

\[
\frac{d(\gamma \beta \hat{v})}{dt} = \gamma^3 \frac{d\beta}{dt} \hat{v} + \gamma \beta \frac{d\hat{v}}{dt} \tag{C.3}
\]

where the vectors \(\hat{v}\) and \(d\hat{v}/dt\) are orthogonal. Then, projecting eq. (C.2) into the directions of these two vectors, we obtain

\[
\frac{d\beta}{dt} = \frac{Z_0 e c}{m_e c \gamma} (1 - \beta^2)(E \cdot \hat{v}) \tag{C.4}
\]

and

\[
\frac{d\hat{v}}{dt} = \frac{Z_0 e c}{m_e c \gamma} [E - (E \cdot \hat{v}) \hat{v} + c \beta \hat{v} \times B]. \tag{C.5}
\]

It then follows that

\[
\frac{d\beta \hat{v}}{dt} = \frac{d\beta}{dt} \hat{v} + \beta \frac{d\hat{v}}{dt} = \frac{Z_0 e c}{m_e c \gamma} \left[ E - \beta^2 (E \cdot \hat{v}) \hat{v} + c \beta \hat{v} \times B \right], \tag{C.6}
\]

which we cast in the form

\[
\frac{dv}{dt} = A, \quad A \equiv \frac{Z_0 e c}{m_e \gamma} \left[ E - \beta^2 (E \cdot \hat{v}) \hat{v} + c \beta \hat{v} \times B \right], \tag{C.7}
\]

\(^1\)In this appendix, electromagnetic quantities are expressed in SI units.
C.1. Tracking particles in vacuum.

Notice that, for arbitrary fields $\mathcal{E}$ and $\mathcal{B}$, the “acceleration” $\mathbf{A}$ is a function of the particle’s position $\mathbf{r}$, energy $E$, and direction of motion $\dot{\mathbf{v}}$.

Implicit integration of eq. (C.7) gives the equations of motion

$$\mathbf{v}(t) = \mathbf{v}_0 + \int_0^t \mathbf{A}(\mathbf{r}(t'), E(t'), \dot{\mathbf{v}}(t')) \, dt', \quad (C.8)$$

$$\mathbf{r}(t) = \mathbf{r}_0 + t \mathbf{v}_0 + \int_0^t \mathbf{v}(t') \, dt'. \quad (C.9)$$

Evidently, these equations are too complex for straight application in a simulation code and we must have recourse to approximate solution methods. We shall adopt the approach proposed by Bielajew (1988), which is well suited to transport simulations. The basic idea is to split the trajectory into a number of conveniently short steps such that the acceleration $\mathbf{A}$ does not change much over the course of a step. Along each step, we then have

$$\mathbf{v}(t) = \mathbf{v}_0 + t \mathbf{A}(\mathbf{r}_0, E_0, \dot{\mathbf{v}}_0) \quad (C.10)$$

$$\mathbf{r}(t) = \mathbf{r}_0 + t \mathbf{v}_0 + t^2 \frac{1}{2} \mathbf{A}(\mathbf{r}_0, E_0, \dot{\mathbf{v}}_0), \quad (C.11)$$

where the subscript “0” indicates values of the various quantities at the starting point ($t = 0$). The traveled path length $s$ and the flying time $t$ are related by

$$t = \int_0^s \frac{ds'}{v}, \quad (C.12)$$

which, to first order becomes

$$t = \frac{s}{v_0} \left(1 - \frac{v(t) - v_0}{2v(t)}\right). \quad (C.13)$$

Then, to first order in the electromagnetic force,

$$\mathbf{v}(s) = \mathbf{v}_0 + s \frac{\mathbf{A}(\mathbf{r}_0, E_0, \dot{\mathbf{v}}_0)}{c\beta_0}$$

$$\mathbf{r}(s) = \mathbf{r}_0 + s \dot{\mathbf{v}}_0 + s^2 \frac{1}{2} \frac{\mathbf{A}(\mathbf{r}_0, E_0, \dot{\mathbf{v}}_0)}{c^2\beta_0^2}.$$ 

That is,

$$\mathbf{r}(s) = \mathbf{r}_0 + s \dot{\mathbf{v}}_0 + s^2 \frac{1}{2} \frac{Z_0 e [\mathbf{E}_0 - \beta_0^2 (\mathbf{E}_0 \cdot \dot{\mathbf{v}}_0) \dot{\mathbf{v}}_0 + c \beta_0 \dot{\mathbf{v}}_0 \times \mathbf{B}_0]}{m_e c^2 \gamma_0 \beta_0^2}. \quad (C.14)$$

The particle’s velocity can be calculated directly from eq. (C.10), which to first order gives

$$\mathbf{v}(s) = \mathbf{v}_0 + \Delta \mathbf{v} \quad (C.15)$$

with

$$\Delta \mathbf{v} = s \frac{Z_0 e [\mathbf{E}_0 - \beta_0^2 (\mathbf{E}_0 \cdot \dot{\mathbf{v}}_0) \dot{\mathbf{v}}_0 + c \beta_0 \dot{\mathbf{v}}_0 \times \mathbf{B}_0]}{m_e c \gamma_0 \beta_0}. \quad (C.16)$$
In the tracking algorithm, the velocity is used to determine the direction vector at the end of the step,

$$\dot{\mathbf{v}}(s) = \frac{\mathbf{v}_0 + \Delta \mathbf{v}}{\| \mathbf{v}_0 + \Delta \mathbf{v} \|}$$

(C.17)

Owing to the action of the electromagnetic force, the kinetic energy $E$ of the particle varies along the step. As the trajectory is accurate only to first order, it is not advisable to compute the kinetic energy from the velocity of the particle. It is preferable to calculate $E(t)$ as

$$E(s) = E_0 + Z_0 e [\varphi(r_0) - \varphi(r(s))]$$

(C.18)

where $\varphi(r)$ is the electrostatic potential, $\mathbf{E} = -\nabla \varphi$. Notice that this ensures energy conservation, i.e. it gives the exact energy variation in going from the initial to the final position.

This tracking method is valid only if

1) the fields do not change too much along the step

$$\frac{\| \mathbf{E}(r(s)) - \mathbf{E}(r_0) \|}{\| \mathbf{E}(r_0) \|} < \delta \mathbf{E} \ll 1, \quad \frac{\| \mathbf{B}(r(s)) - \mathbf{B}(r_0) \|}{\| \mathbf{B}(r_0) \|} < \delta \mathbf{B} \ll 1$$

(C.19)

and

2) the relative changes in kinetic energy and velocity (or direction of motion) are small

$$\left| \frac{E(s) - E_0}{E_0} \right| < \delta E \ll 1, \quad \frac{\| \Delta \mathbf{v} \|}{\mathbf{v}_0} < \delta_v \ll 1.$$  

(C.20)

These conditions set an upper limit on the allowed step length, $s_{\text{max}}$, which depends on the local fields and on the energy and direction of the particle. The method is robust, in the sense that it converges to the exact trajectory when the maximum allowed step length tends to zero. In practical calculations, we shall specify the values of the $\delta$-parameters (which should be of the order of 0.05 or less) and consider step lengths consistent with the above conditions. Thus, the smallness of the $\delta$-parameters determines the accuracy of the generated trajectories.

To test the accuracy of a tracking algorithm, it is useful to consider the special cases of a uniform electric field (with $\mathbf{B} = 0$) and a uniform magnetic field (with $\mathbf{E} = 0$), which admit relatively simple analytical solutions of the equations of motion.

### C.1.1 Uniform electric fields

Let us study first the case of a uniform electric field $\mathbf{E}$. The equation of the trajectory of an electron/positron that starts at $t = 0$ from the point $\mathbf{r}_0$ with velocity $\mathbf{v}_0$ can be expressed in the form (adapted from Bielajew, 1988)

$$\mathbf{r}(t) = \mathbf{r}_0 + t \mathbf{v}_{0 \perp} + \frac{1}{a} \left[ \cosh (a c t) - 1 + \frac{v_{0 \parallel}}{c} \sinh (a c t) \right] \mathbf{\hat{E}},$$

(C.21)
where \( \mathbf{v}_{0\parallel} \) and \( \mathbf{v}_{0\perp} \) are the components of \( \mathbf{v}_0 \) parallel and perpendicular to the direction of the field,

\[
\mathbf{v}_{0\parallel} = (\mathbf{v}_0 \cdot \hat{\mathbf{E}}) \hat{\mathbf{E}}, \quad \mathbf{v}_{0\perp} = \mathbf{v}_0 - (\mathbf{v}_0 \cdot \hat{\mathbf{E}}) \hat{\mathbf{E}}
\]

and

\[
a = \frac{Z_0 e \mathcal{E}}{m_e c^2 \gamma_0} = \frac{Z_0 e \mathcal{E}}{E_0}
\]

The velocity of the particle is

\[
\mathbf{v}(t) = \mathbf{v}_{0\perp} + \left[ c \sinh(\alpha ct) + v_{0\parallel} \cosh(\alpha ct) \right] \hat{\mathbf{E}}
\]

\[
= \mathbf{v}_0 + \left\{ c \sinh(\alpha ct) + v_{0\parallel} \left[ \cosh(\alpha ct) - 1 \right] \right\} \hat{\mathbf{E}}.
\]

(C.24)

Since the scalar potential for the constant field is \( \varphi(\mathbf{r}) = -\mathcal{E} \cdot \mathbf{r} \), the kinetic energy of the particle varies with time and is given by

\[
E(t) = E_0 - Z_0 e \mathcal{E} \cdot [\mathbf{v}_0 - \mathbf{r}(t)].
\]

(C.25)

Figure C.1: Trajectories of electrons and positrons in a uniform electric field of 511 kV/cm. Continuous curves represent exact trajectories obtained from eq. (C.21). The dashed lines are obtained by using the first-order numerical tracking method described by eqs. (C.14)-(C.20) with \( \delta \mathcal{E} = \delta E = \delta \varphi = 0.02 \). The displayed trajectories correspond to the following cases. a: positrons, \( E_0 = 0.1 \) MeV, \( \theta = 135 \) deg. b: positrons, \( E_0 = 1 \) MeV, \( \theta = 135 \) deg. c: positrons, \( E_0 = 10 \) MeV, \( \theta = 135 \) deg. f: electrons, \( E_0 = 0.1 \) MeV, \( \theta = 90 \) deg. g: electrons, \( E_0 = 1 \) MeV, \( \theta = 90 \) deg. h: electrons, \( E_0 = 10 \) MeV, \( \theta = 90 \) deg.

Fig. C.1 displays trajectories of electrons and positrons with various initial energies and directions of motion in a uniform electric field of 511 kV/cm directed along the positive z-axis. Particles start from the origin \( (\mathbf{r}_0 = 0) \), with initial velocity in the \( xz \)-plane forming an angle \( \theta \) with the field, i.e. \( \mathbf{v}_0 = (\sin \theta, 0, \cos \theta) \), so that the whole
trajectories lie in the $xz$-plane. Continuous curves represent exact trajectories obtained from the analytical formula (C.21). The dashed curves are the results from the first-order tracking algorithm described above [eqs. (C.14)-(C.20)] with $\delta \mathbf{E} = \delta \mathbf{v} = 0.02$. We show three positron trajectories with initial energies of 0.1, 1 and 10 MeV, initially moving in the direction $\theta = 135$ deg. Three trajectories of electrons that initially move perpendicularly to the field ($\theta = 90$ deg) with energies of 0.1, 1 and 10 MeV are also depicted. We see that the tracking algorithm gives quite accurate results. The error can be further reduced, if required, by using shorter steps, i.e. smaller $\delta$-values.

### C.1.2 Uniform magnetic fields

We now consider the motion of an electron/positron, with initial position $\mathbf{r}_0$ and velocity $\mathbf{v}_0$, in a uniform magnetic field $\mathbf{B}$. Since the magnetic force is perpendicular to the velocity, the field does not alter the energy of the particle and the speed $v(t) = v_0$ is a constant of the motion. It is convenient to introduce the precession frequency vector $\mathbf{\omega}$, defined by (notice the sign)

$$\mathbf{\omega} \equiv -\frac{Z_0 e \mathbf{B}}{m_c \gamma} = -\frac{Z_0 e c^2 \mathbf{B}}{E_0},$$

(C.26)

and split the velocity $\mathbf{v}$ into its components parallel and perpendicular to $\mathbf{\omega}$,

$$\mathbf{v}_\parallel = (\mathbf{v} \cdot \mathbf{\omega})\mathbf{\omega}, \quad \mathbf{v}_\perp = \mathbf{v} - (\mathbf{v} \cdot \mathbf{\omega})\mathbf{\omega}.$$  

(C.27)

Then, the equation of motion (C.7) becomes

$$\frac{d\mathbf{v}_\parallel}{dt} = 0, \quad \frac{d\mathbf{v}_\perp}{dt} = \mathbf{\omega} \times \mathbf{v}_\perp.$$  

(C.28)

The first of these eqs. says that the particle moves with constant velocity $\mathbf{v}_0$ along the direction of the magnetic field. From the second eq. we see that, in the plane perpendicular to $\mathbf{B}$, the particle describes a circle with angular frequency $\omega$ and speed $v_0$ (which is a constant of the motion). The radius of the circle is $R_\perp = v_0/\omega$. That is, the trajectory is an helix with central axis along the $\mathbf{B}$ direction, radius $R_\perp$ and pitch angle $\alpha = \arctan(v_0/\omega)$. The helix is right-handed for electrons and left-handed for positrons (see fig. C.2).

In terms of the path length $s = tv_0$, the equation of motion takes the form

$$\mathbf{r}(s) = \mathbf{r}_0 + \frac{s}{v_0} \mathbf{v}_0 + R_\perp [\cos(s/\omega - 1)] (\mathbf{v}_0 \times \mathbf{\omega}) + R_\perp \sin(s/\omega) \mathbf{v}_0,$$

(C.29)

where $\mathbf{v}_0 = \mathbf{v}_0/\omega$ and $s = st_0/v_0$. Equivalently,

$$\mathbf{r}(s) = \mathbf{r}_0 + s \mathbf{\dot{v}}_0 - \frac{s}{v_0} \mathbf{v}_0 + \frac{1}{\omega} [\cos(s\omega/v_0 - 1)] (\mathbf{v}_0 \times \mathbf{\omega}) + \frac{1}{\omega} \sin(s\omega/v_0) \mathbf{v}_0.$$  

(C.30)

After the path length $s$, the particle velocity is

$$\mathbf{v}(s) = v_0 \frac{d\mathbf{r}}{ds} = \mathbf{v}_0 + [\cos(s\omega/v_0 - 1)] \mathbf{v}_0 - \sin(s\omega/v_0) (\mathbf{v}_0 \times \mathbf{\omega}).$$  

(C.31)
C.1. Tracking particles in vacuum.

\[ \text{Figure C.2: Trajectories of electrons and positrons in a uniform magnetic field. The two particles start from the base plane with equal initial velocities.} \]

In fig. C.3 we compare exact trajectories of electrons and positrons in a uniform magnetic field obtained from the analytical formula (C.30) with results from the first-order tracking algorithm [eqs. (C.14)-(C.20)] with \( \delta_B = \delta_E = \delta_\nu = 0.02 \). The field strength is 0.2 tesla. The depicted trajectories correspond to 0.5 MeV electrons (a) and 3 MeV positrons (b) that initially move in a direction forming an angle of 45 deg with the field. We see that the numerical algorithm is quite accurate for small path lengths, but it deteriorates rapidly for increasing \( s \). In principle, the accuracy of the algorithm can be improved by reducing the value of \( \delta_\nu \), i.e. the length of the step length. In practice, however, this is not convenient because it implies a considerable increase of numerical work, which can be easily avoided.

\[ \text{Figure C.3: Trajectories of electrons and positrons in a uniform magnetic field of 0.2 tesla. Continuous curves are exact trajectories calculated from eq. (C.30). The short-dashed lines are obtained by using the numerical tracking method described in the text with } \delta_\nu = 0.02. \text{ Long-dashed curves are the results from the tracking algorithm with } \delta_\nu = 0.005. \text{ a: electrons, } E_0 = 0.5 \text{ MeV, } \theta = 45 \text{ deg. b: positrons, } E_0 = 3 \text{ MeV, } \theta = 45 \text{ deg.} \]
C.2 Exact tracking in homogeneous magnetic fields

In our first-order tracking algorithm [see eqs. (C.14) and (C.16)], the effects of the electric and magnetic fields are decoupled, i.e. they can be evaluated separately. For uniform electric fields, the algorithm offers a satisfactory solution since it usually admits relatively large step lengths. In the case of uniform magnetic fields (with $E = 0$), the kinetic energy is a constant of the motion and the only effective constraint on the step length is that the change in direction $|Δv|/v_0$ has to be small. Since the particle trajectories on the plane perpendicular to the field $B$ are circles and the first-order algorithm generates each step as a parabolic segment, we need to move in sub-steps of length much less than the radius $R_\perp$ (i.e. $δ_\perp$ must be given a very small value) and this makes the calculation slow. On the other hand, the action of the uniform magnetic field is described by simple analytical expressions [eqs. (C.30) and (C.31)], that are amenable for direct use in the simulation code. These arguments suggest the following obvious modification of the tracking algorithm.

As before, we assume that the fields are essentially constant along each trajectory step and write

$$\mathbf{r}(s) = \mathbf{r}_0 + s \dot{\mathbf{r}}_0 + (\Delta \mathbf{r})_E + (\Delta \mathbf{r})_B,$$  \hspace{1cm} (C.32)

where $(\Delta \mathbf{r})_E$ and $(\Delta \mathbf{r})_B$ are the displacements caused by the electric and magnetic fields, respectively. For $(\Delta \mathbf{r})_E$ we use the first-order approximation [see eq. (C.14)],

$$ (\Delta \mathbf{r})_E = \frac{s}{2} \frac{Z_0 e [\mathbf{E}_0 - \beta_0^2 (\mathbf{E}_0 \cdot \dot{\mathbf{v}}_0) \dot{\mathbf{v}}_0]}{m_e \gamma_0 \beta_0^2}. $$  \hspace{1cm} (C.33)

The displacement caused by the magnetic field is evaluated using the result (C.30), i.e.

$$ (\Delta \mathbf{r})_B = -\frac{s}{v_0} \mathbf{v}_{0\perp} + \frac{1}{\omega} [\cos(s\omega/v_0) - 1] (\mathbf{v}_{0\perp} \times \dot{\omega}) + \frac{1}{\omega} \sin(s\omega/v_0) \mathbf{v}_{0\perp} $$  \hspace{1cm} (C.34)

with

$$ \omega \equiv -\frac{Z_0 e c^2 B_0}{E_0}, \quad \text{and} \quad \mathbf{v}_{0\perp} = \mathbf{v}_0 - (\mathbf{v}_0 \cdot \dot{\omega}) \dot{\omega}. $$  \hspace{1cm} (C.35)

Similarly, the particle velocity along the step is expressed as

$$ \mathbf{v}(s) = \mathbf{v}_0 + (\Delta \mathbf{v})_E + (\Delta \mathbf{v})_B $$  \hspace{1cm} (C.36)

with [see eqs. (C.16) and (C.31)]

$$ (\Delta \mathbf{v})_E = s \frac{Z_0 e [\mathbf{E}_0 - \beta_0^2 (\mathbf{E}_0 \cdot \dot{\mathbf{v}}_0) \dot{\mathbf{v}}_0]}{m_e \gamma_0 \beta_0} $$  \hspace{1cm} (C.37)

and

$$ (\Delta \mathbf{v})_B = [\cos(s\omega/v_0) - 1] \mathbf{v}_{0\perp} - \sin(s\omega/v_0) (\mathbf{v}_{0\perp} \times \dot{\omega}). $$  \hspace{1cm} (C.38)
In our implementation of this tracking algorithm, the allowed step lengths \( s \) are limited by the following constraints [see eqs. (C.19) and (C.20)]

\[
\frac{|\mathbf{E}(\mathbf{r}(s)) - \mathbf{E}(\mathbf{r}_0)|}{|\mathbf{E}(\mathbf{r}_0)|} < \delta_E \ll 1, \quad \frac{|\mathbf{B}(\mathbf{r}(s)) - \mathbf{B}(\mathbf{r}_0)|}{|\mathbf{B}(\mathbf{r}_0)|} < \delta_B \ll 1 \tag{C.39}
\]

and

\[
\left| \frac{E(s) - E_0}{E_0} \right| < \delta_E \ll 1, \quad \left| \frac{(\Delta v) \mathbf{E} + (\Delta v) \mathbf{B}}{v_0} \right| < \delta_v \ll 1. \tag{C.40}
\]

The algorithm is robust, i.e. the accuracy of the generated trajectories increases when the \( \delta \)-parameters are reduced. In many practical cases, a good compromise between accuracy and simulation speed is obtained by setting \( \delta_E = \delta_B = \delta_E = \delta_v = 0.02 \). Notice that, in the case of a uniform magnetic field, the tracking algorithm is now exact, irrespective of the step length.

This tracking algorithm has been implemented in the subroutine package PENFIELD, which is devised to work linked to PENELPOE and PENGEO. To simulate radiation transport in a given field/material configuration, the user must provide the steering main program as well as specific routines that define the EM field (see the examples and comments in the source file \textsc{PENFIELD.F}).
Bibliography


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Radiation is used in many applications of modern technology. Its proper handling requires competent knowledge of the basic physical laws governing its interaction with matter. To ensure its safe use, appropriate tools for predicting radiation fields and doses, as well as pertinent regulations, are required.

One area of radiation physics that has received much attention concerns electron-photon transport in matter. PENELOPE is a modern, general-purpose Monte Carlo tool for simulating the transport of electrons and photons, which is applicable for arbitrary materials and in a wide energy range. PENELOPE provides quantitative guidance for many practical situations and techniques, including electron and x-ray spectroscopies, electron microscopy and microanalysis, biophysics, dosimetry, medical diagnostics and radiotherapy, and radiation damage and shielding.

These proceedings contain the teaching notes of a recent workshop/training course on PENELOPE, with a detailed description of the physics, numerical algorithms and structure of the code system.