**Manual for**

**Generalized Monte Carlo Tool for Investigating Low-Field**

**and High Field Properties of Materials Using**

**Non-parabolic Band Structure Model**

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**ABSTRACT**

In semiconductor physics, many properties or phenomena of materials can be brought to light through certain changes in the materials. Having a tool to define new material properties so as to highlight certain phenomena greatly increases the ability to understand that phenomena. The generalized Monte Carlo tool allows the user to do that by keeping every parameter used to define a material, within the non-parabolic band approximation, a variable in the control of the user. A material is defined by defining its valleys, energies, valley effective masses and their directions. The types of scattering to be included can also be chosen. The non-parabolic band structure model is used. The versatility of the tool has been investigated and has produced results closely matching the experimental values for some common materials.

**Table of Contents**

 Abstract

1. Introduction
2. The Monte Carlo Method
	1. Single Particle Monte Carlo Method
	2. Ensemble Monte Carlo Method
	3. Fermi’s Golden Rule
	4. Non-parabolic Bands
3. The Generalized Monte Carlo Code
4. Results
5. Conclusions

References

# INTRODUCTION

Semiconductors have been the focal point of study for electrical transport from the 20th century onwards. The main attraction towards these materials was the ability to change the conductivity of the semiconductor by introducing dopants and also by applying an electric field. Although the main use of semiconductor materials in devices started with the invention of the transistor, it was not the first device to use semiconductors. Metal rectifiers and detectors in radios called ‘Cat Whiskers’, which were primitive forms of modern day Schottky diodes, were quite common in the beginning of the 20th century [1]. The investigation of semiconductor materials could be said to have started with Russell Ohl of Bell Laboratories when he tried to grow pure crystals of these semiconductors and analyze their properties. These tests led to the realization of a diode structure to alter the electrical properties of a material. Building on the knowledge of how those diodes work, William Schockley, John Bardeen and Walter Brattain sandwiched two diodes together to create the first transistor in 1947 at Bell Labs. Since then the number of different semiconductor materials has grown immensely to produce a variety of devices exploiting the individual, unique advantages of these materials. The experimental success of the semiconductor industry would not have been as successful without the corresponding success in the understanding of the physical properties such as the electrical and optical properties of these devices. These properties were investigated in detail in the 1950’s itself [2], leading to the understanding of the energy band structures. From a greater understanding of band structures and other transport properties of these materials, the electrical properties of the devices created closely followed the theories proposed at that time.

 The study of charge transport in semiconductors is of fundamental importance both from the point of view of the basic physics and for its applications to electrical devices [3]. As the need for electrical appliances grew the need for smaller and faster devices grew as well. As can be seen in Figure 1.1 the number of transistors in a device grew at an amazing rate closely following Moore’s law [4].



Figure 1.1 CPU Transistor counts (from www.intel.com)

This was made possible by the reduction in the size of the devices. Since the voltage applied across the devices did not reduce at the same rate, the field applied across the device increased. Soon after the invention of the transistor in 1947 it was recognized that electric field strengths, so high so as to cause the devices to no longer obey Ohm’s Law, were encountered in semiconductor samples [5-6]. As the requirement of having such high electric fields in commercial transistors became a possibility, the need for new physics to tackle the working of these devices arose. The field of nonlinear transport which had been initiated by Landau and Kompanejez [7] entered a period of rapid development soon after the invention of the transistor and a number of researchers devoted their efforts to improving the scientific knowledge of this subject. The surge in research in this field gave way to the realization of new phenomena like the Gunn Effect [8] which then led to the invention of new types of devices like the transit-time device.

 Analyzing charge transport at high electric fields in devices operated in the on-state is a difficult problem both from the mathematical and physical point of view. The Boltzmann equation which defines the transport phenomena for semi-classical cases is a complicated integro-differential equation. Analytic solutions of the Boltzmann transport equation can only be obtained for very few cases and are usually not applicable to real systems. In order to get an analytic result it is necessary to use such drastic approximations that it can no longer be considered appropriate to describe real device operation. In 1966 Kurosawa proposed the Monte Carlo technique [9] and Budd proposed the iterative technique [10]. With these techniques it became clear that, with the use of modern computers, it would be possible to exactly solve the Boltzmann transport equation numerically for physical models of considerable complexity. These two techniques were then developed further by Price [11], Rees [12] and Fawcett [13]. The Monte Carlo method became the more popular technique because it is easier to use and gives more physically interpretable results.

 Low field properties of the semiconductor can be investigated using the relaxation time approximation (RTA) for the case when the relevant scattering processes are either elastic or isotropic. If that is not the case, then Rode’s iterative procedure has to be used [14].

The Monte Carlo method, which can be used for calculation of low-field and high-field properties of a semiconductor, uses a different methodology. In fact, in the long-time limit the Monte Carlo method gives the solution of the Boltzmann Transport Equation (BTE). In the short-time limit, the Monte Carlo method gives the solution of the Prigogine equation. Monte Carlo techniques are statistical numerical methods, which are applied to the simulation of random processes. In fact, the Monte Carlo method as a statistical numerical method was born well before its application to transport problems [15] and has been applied to a number of scientific fields [16-17]. In case of the charge transport, however, the solution of the Boltzmann transport equation is a direct simulation of the dynamics of the carriers in the material. This means that while the simulation is being run, and the solution is being built up, any physical information can be easily extracted. Therefore, even though the result of the Monte Carlo simulation requires a correct physical interpretation, the method is a very useful tool to achieve real solutions. It permits the simulation of particular physical situations unattainable in experiments, or even investigation of nonexistent materials in order to emphasize special features of the phenomenon under study. This use of the Monte Carlo technique makes it similar to an experimental technique and can be compared with analytically formulated theory.

1. **THE MONTE CARLO METHOD**

The purpose of device modeling is to be able to predict the electrical properties of materials and devices. This would then permit changing certain parameters to improve performance. To obtain these electrical properties, one needs to know the behavior of the particles in the devices, or more specifically in the materials used in those devices. The Boltzmann transport equation [18-19]

|  |  |  |
| --- | --- | --- |
|  |  | (2.1) |

is used to obtain this behavior. It governs the carrier transport in materials under the semi-classical approximation. This equation is essentially a conservation of volumes in phase space. The left hand side of equation (2.1) consists of three terms, the first term describes the temporal variation of the distribution function, the second term describes the spatial variation of the distribution function which may arise due to temperature or concentration gradients, and finally the third term describes the effect on the distribution function due to applied fields (electric or magnetic). On the right hand side we have two terms, the first term describes the recombination and generation processes and the second term is the collision integral which describes the scattering processes. As can be seen, the Boltzmann transport equation is a complicated integro-differential equation which, if needs to be solved analytically, requires many simplifying assumptions which may not hold in real devices as was mentioned earlier.

The Monte Carlo method is a stochastic method used to solve the Boltzmann transport equation. In order to develop this approach we first write the Boltzmann equation as in [18],

|  |  |  |
| --- | --- | --- |
|  |  | (2.2) |

where

|  |  |  |
| --- | --- | --- |
|  |  | (2.3) |

is the total scattering rate out of state **p** for all scattering processes. This motion of the distribution function is described in six plus one-dimensional phase space, three in momentum, three in real space and one in time. It is therefore convenient to describe the motion of the distribution function along a trajectory in phase space. The variable along this trajectory is taken to be *s* and each coordinate can be parameterized as a function of this variable as

|  |  |  |
| --- | --- | --- |
|  |  | (2.4) |

and the partial derivatives are constrained by the relationships

|  |  |  |
| --- | --- | --- |
|  |  | (2.5) |

Applying these changes to equation (2.2) we get

|  |  |  |
| --- | --- | --- |
|  |  | (2.6) |

Equation (2.6) is a standard differential equation which can be solved using an integrating factor which gives

|  |  |
| --- | --- |
|  | (2.7) |

By a change of variables from , the above equation becomes

|  |  |
| --- | --- |
|   | (2.8) |

The above equation is the Chamber-Rees path integral [20] and is the form of the Botlzmann transport equation which can be iteratively solved. In order to make the above equation solvable, a useful mathematical trick introduced by Rees [21] is used which makes the complicated energy dependent function into an energy independent term, thereby making the term inside the integral in equation (1.7) trivially solvable. This is done by introducing a scattering term called self-scattering (. Self-scattering does not change the momentum or the energy of the particle and therefore does not change the physics of the particle. What this term does, however, is to convert the energy dependent function  into an energy independent term by defining

|  |  |  |
| --- | --- | --- |
|  |  | (2.9) |

Therefore, equation (2.8) becomes

|  |  |  |
| --- | --- | --- |
|  |  | (2.10) |

where

|  |  |  |
| --- | --- | --- |
|  |  | (2.11) |

The first term of equation (2.10) is a transient term while the second term is the term which can be iteratively solved. If we look at the second term closely the first integral over represents the scattering of the distribution function out of state to state . The second integral represents the integration along the trajectory *s* and the exponential is just the probability that no scattering takes place during the time it moves a distance *s*. Thus, if we look at how the electrons move physically, it consists of a scattering event determined by the first integral and then there is a free-flight motion (no scattering) for a time interval *ts*. Rees showed that the time steps *ts* correlate to 1/ΓT. This free-flight scatter sequence is the basis of every Monte-Carlo method used in device simulations.

 The Monte Carlo method is mainly used in three different styles, the one-particle Monte Carlo, the ensemble Monte Carlo and the self-consistent ensemble Monte Carlo. In the one-particle Monte Carlo method a single carrier’s motion is tracked for a certain period of time until the particle has reached steady state. This method is mostly useful to study bulk properties, like the steady state drift velocity as a function of field.

In the ensemble Monte Carlo method a large ensemble of carriers are simulated at the same time. This method can be sped up using parallelization and is useful for super-computation. This method is mostly useful for transient analysis as ensemble averages can be taken at certain time steps during the simulation.

 In the self-consistent ensemble Monte Carlo method, the ensemble Monte Carlo method is coupled with a Poisson solver or also to a Schrödinger solver and is the most suitable method for device simulations.

## 2.1 Single Particle Monte Carlo Method

 As was mentioned earlier, there are free flight times (drift times) and then scatter sequences in the Monte Carlo method. If is the probability that an electron in state *k* suffers a collision during the time interval *dt*, then the probability that an electron which has had a collision at time t=0 has not yet undergone another collision after time *t* is [22],

|  |  |  |
| --- | --- | --- |
|  |  | (2.12) |

Therefore, the probability that the electron will suffer a collision during around is,

|  |  |  |
| --- | --- | --- |
|  |  | (2.13) |

If is the maximum of in the region of k-space then,

|  |  |  |
| --- | --- | --- |
|  |  | (2.14) |

Using a random variable transformation and integrating equation (2.14) on both sides we obtain,

|  |  |  |
| --- | --- | --- |
|  |  | (2.15) |

where *r* is a random number between 0 and 1. As can be seen the value of *ti* will have a higher probability of being a value around 1/ΓT, and if we take a large number of particles, the average *ti* will be around 1/ΓT,

|  |  |  |
| --- | --- | --- |
|  |  | (2.16) |

The total scattering rate is calculated by adding up all the scattering rates of individual types of scattering as well as the self-scattering rate which are all energy dependent.

|  |  |  |
| --- | --- | --- |
|  |  | (2.17) |

where *n* is the total number of scattering types considered for a particular material (e.g. acoustic phonon scattering, optical phonon scattering etc.) As one can see, the value of has no upper limit, only a lower limit. It obviously cannot be lower than . But as is energy dependent, it is important that be greater than , where is the energy of the scattering table at which the cumulative scattering rate is the maximum.

 Once the particle has drifted, it is time to scatter it. The type of scattering to be used is chosen from the scattering tables. The usual method is to store the scattering values for each type as a fraction of the total scattering rate . The figure below better explains this method.



Figure 2.1 Construction of scattering tables (left panel) and scattering tables renormalization (right panel)

A random number (R) is chosen between 0 and 1 and if

|  |  |  |
| --- | --- | --- |
|  |  | (2.18) |

then scattering type *j+1* is chosen. Here, of course, . If self scattering is chosen, we do nothing and move on. Once enough time has elapsed, the average carrier velocity is calculated using

|  |  |  |
| --- | --- | --- |
|  |  | (2.19) |

This average is only valid as long as the distribution function is in steady state. This prevents the analysis of transient behavior, like velocity overshoot effects, or any other non-ergodic process. But using the single particle Monte Carlo method the steady state velocity, steady-state energy and other parameters can be calculated.

## 2.2 Ensemble Monte Carlo Method

 A different approach than what is described in the previous section is commonly used by most, and that is the ensemble Monte Carlo Method. Instead of following a single particle for hundreds of thousands of iterations, thousands of particles can be followed for a much lesser number of iterations. In this manual the ensemble Monte Carlo method is adopted. The time coordinate of each electron must be maintained during the simulation.

 The physical quantities such as velocity and energy are averaged over the whole ensemble at frequent time intervals so as to obtain the time evolution of these quantities. For example,

|  |  |  |
| --- | --- | --- |
|  |  | (2.20) |

where *N* is the number of particles in the ensemble and *t* is one of the time intervals at which the ensemble averages are taken. The general block diagram of an ensemble Monte Carlo code is shown in figure 2.2. In figure 2.2 the initial distribution of carriers is a Maxwellian distribution at the given temperature.

 The ensemble Monte Carlo method does not require steady state conditions to calculate the ensemble averages and therefore can be used to investigate transients in bulk systems and devices. Equation (2.21) and equation (2.22) represent an estimate of the true velocity and energy which has a standard error of where is the variance that is estimated from [23],

|  |  |  |
| --- | --- | --- |
|  |  | (2.21) |

and

|  |  |  |
| --- | --- | --- |
|  |  | (2.22) |

for the velocity and energy calculations respectively. Typically the value of *N* is of the order of 104 or 105.



Figure 2.2 Block Diagram of the Ensemble Monte Carlo Code

To obtain the time evolution of certain physical quantities, the need to ‘freeze’ the simulation comes up. The time steps *∆t* at which the simulation is paused and the ensemble averages taken should not be much larger than the maximum frequency of scattering. If it is, it will cause a coarsening of the time evolution and a loss of information. If the time step is too small then it will create noise in the output. Therefore a balance is needed for the time steps. One usually keeps it at a few femtoseconds. Therefore there are two different time scales used in this method, the free-flight duration time and the sampling time. The different time coordinates are shown in the figure below.



Figure 2.3 Free-Flight Scatter representation of the Monte Carlo Method

In the above figure, *N* is the total number of particles in the simulation while *ts* is the total simulation time and is the free-flight duration time for the ith particle. The simulation is paused and the ensemble averages are taken at every as shown in figure 2.3.

## 2.3 Fermi’s Golden Rule

 The scattering processes which interrupt the carrier free-flights are calculated quantum mechanically. The scattering event is treated by defining a scattering potential, which is calculated for each type of scattering process. Each of the different processes, or interactions leads to a different “matrix element” form in terms of its dependence on the initial wave vector, the final wave vector and their corresponding energies. The matrix element is given by,

|  |  |  |
| --- | --- | --- |
|  |  | (2.23) |

For three dimensional cases the matrix element usually contains the momentum conservation condition which arises due to the overlap of the normal Bloch functions of the electrons.

 Solving the time-dependent Schrödinger equation using first-order perturbation theory leads to the equation for the scattering rate from a state to *k* as,

|  |  |  |
| --- | --- | --- |
|  |  | (2.24) |

Equation (2.24) is called Fermi’s Golden Rule, where and are the initial and final states of the carrier, and are the corresponding kinetic energies and is the phonon energy and describes the conservation of energy during the scattering process. The conservation of energy is only valid in the long-time limit, i.e. when the scattering events are infrequent. The top sign is for absorption and the bottom sign is for the phonon emission process.

 The total scattering rate out of a state defined by wave vector and the energy is obtained by summing over all states in equation (2.24).

|  |  |  |
| --- | --- | --- |
|  |  | (2.25) |

In equation (2.25) the sum over all states can be converted to an integral over giving,

|  |  |  |
| --- | --- | --- |
|  |  | (2.26) |

where is the total volume of the crystal and is given by equation (2.24). Equation (2.26) is used to calculate scattering rates as a function of energy.

## 2.4 Non-Parabolic bands

The equation mapping the energy of an electron above the valley minima to its wave vector k using the parabolic band approximation is,

|  |  |  |
| --- | --- | --- |
|  |  | (2.27) |

This approximation is only valid for energies slightly greater than the energy of the valley minima. For Monte Carlo simulations, in which high field transport is essential, this approximation is not accurate enough. To improve accuracy we need a function that better maps the energy of an electron to its wave vector. A full band calculation will give us a more accurate mapping of energy and momentum of an electron but as it is not an analytic function it becomes hard to switch freely between the energy and momentum which is essential to do in a Monte-Carlo simulation. Therefore, the full band simulations are very computer intensive. In order to use an analytic approach and still improve upon the accuracy the **k.p** method is used to obtain the non-parabolic equation [24],

|  |  |  |
| --- | --- | --- |
|  |  | (2.28) |

 Here, α is a term coming from the **k.p** method which depends on the material as,

|  |  |  |
| --- | --- | --- |
|  |  | (2.29) |

where is the energy difference between the conduction band and the valence band at the point, is the electron rest mass and is the conductivity mass. The above equation is also an approximation valid as long as,

|  |  |  |
| --- | --- | --- |
|  |  | (2.30) |

 For electron energies in which ~ , the above equation fails and a full band calculation is required to more accurately simulate transport in the material. Assuming that the above assumptions are valid it is important to note the changes that the non-parabolic band approximation introduces. The density of states and the conductivity effective masses are given by,

|  |  |  |
| --- | --- | --- |
|  |  | (2.31) |

For parabolic bands both masses turn out to be equal to *m*, but when non-parabolic bands are used we get,

|  |  |  |
| --- | --- | --- |
|  |  | (2.32) |

 It is interesting to note that in this method, as the electron gains more and more energy its effective mass increases. This means that the electron becomes ‘heavier’ or reacts slower to the electric field as its energy increases. The change in density of states effective mass causes the scattering rates to get modified slightly as compared to the parabolic bands case. The final expressions for the scattering rates of most common scattering processes in semiconductors are listed in Table 2.1 below.

Table 2.1 Scattering rates of most common semiconductors.

|  |
| --- |
| 1. Acoustic Phonon Scattering |
|  |
| 2. Intervalley Phonon Scattering |
|  |
| 3. Ionized Impurity Scattering |
|  |
| 4. Polar Optical Phonon Scattering |
|     |

|  |
| --- |
| 5. Piezoelectric Scattering |
|  |
| 6. Dislocation Scattering (e.g. GaN) |
| where *n’* is the effective screening concentration Ndis is the Line dislocation density |
| 7. Alloy Disorder Scattering (AlxGa1-xAs) |
| Where: *d* is the lattice disorder (0≤d≤1) *Dalloy* is the alloy disorder scattering potential |

# THE GENERALIZED MONTE CARLO CODE

The purpose of the generalized Monte Carlo code is to give users the option of defining their own material or modifying the definition of an existing material. Technically any number of different materials can be simulated within the non-parabolic band approximation. This makes the code very versatile and necessitates a very general method of implementing the code which results in a large set of input parameters which, in turn, increases the complexity of the code.

 In the generalized Monte Carlo code there is a wide range of input parameters. The input parameters are either loaded from a file or from the Rappture interface. In order to be able to define a material the user can input the number of valleys to be used in the simulation, the number of sub-valleys (equivalent valleys) within each valley, the direction of those sub-valleys, the effective masses of electrons in those sub-valley directions and the energy difference between the bottom of the valleys. In addition to choosing deformation potential scattering (acoustic and optical), ionized impurity scattering, polar-optical phonon scattering and piezoelectric scattering the user can also specify whether the optical phonon scattering within a valley are umklapp processes like it is needed in Silicon for the case of g-phonon intervalley scattering.

 The scattering table has to be created and stored separately for each valley as the scattering parameters depend on the effective mass of the electron which differs from valley to valley. Also some scattering processes might exist in one valley but not in the other (for e.g. in Germanium f and g type scattering is considered in the X valley but not in the L valley). Therefore the number of scattering processes can also vary from one valley to another. The scattering tables are normalized to the maximum value of the total scattering rate within the energy range specified by the user which will be different for different valleys.

 

Figure 3.1 Flow chart for the generalized Monte Carlo code

 The electrons are initialized to a Maxwell distribution at the temperature *T*. The formulae used to initialize the electron’s energy and momentum are:

 (3.1)

The electrons, whose number is defined by the user, are initially placed in the lowest energy valley which is also defined by the user and is equally distributed to all the sub-valleys present in that valley. Here *m* is the drift (conductivity) effective mass used to make the bands isotropic.

 In between scattering events, the electron is drifted under the applied electric field. The equation describing the dispersion relation of the electron for a general sub-valley for non-parabolic bands is,

|  |  |  |
| --- | --- | --- |
|  |  | (3.2) |

where are the wave-vectors along the three mutually perpendicular directions that define the sub-valley and are the effective masses of the electrons along those directions.

Equation (3.2) represents the dispersion relation for an anisotropic band, which is the most general case. To calculate the drift velocity equations, equation (3.2) must first be converted to an isotropic dispersion relation by changing the wave-vectors to where,

|  |  |  |
| --- | --- | --- |
|  |  . | (3.3) |

Here is the conductivity effective mass used for all . Substituting equation (3.3) into equation (3.2) we get,

|  |  |  |
| --- | --- | --- |
|  |  | (3.4) |

The above equation now represents the dispersion relation of an electron for a spherical band with effective mass in all directions. According to Newton’s second law of motion - the rate of change of momentum is equal to the force applied to the electron giving,

|  |  |  |
| --- | --- | --- |
|  |  | (3.5) |

where are the electric field magnitudes along the mutually perpendicular directions that define the sub-valley. is the drift time selected by equation (2.15). Substituting equation (3.5) into (3.6) we get,

|  |  |  |
| --- | --- | --- |
|  |  | (3.6) |

The electric field applied to the device is defined by the user before the simulation starts along the (x,y,z) coordinate system. In order to drift the electron according to equation (3.6) we need the electric field magnitudes along the three mutually perpendicular directions (1,2,3) which define the sub-valley in k-space. In general, the coordinate system (1,2,3) is completely different from the (x,y,z) coordinate system. Therefore, before every electron is drifted, the sub-valley in which the electron currently exists is identified and the coordinate system is transformed from the (x,y,z) system to the (1,2,3) system which defines that sub-valley in k-space. The electric field acting on the electron is then given by,

|  |  |
| --- | --- |
|  | (3.7) |

where the three mutually perpendicular directions that describe the sub-valley are [a1,b1,c1], [a2,b2,c2] and [a3,b3,c3]. According to equation (3.6) we also need to have the wave vectors along the directions [a1,b1,c1], [a2,b2,c2] and [a3,b3,c3]. This is obtained by performing another transformation which gives,

|  |  |
| --- | --- |
|  | (3.8) |

The electron is then drifted according to equation (3.6) and the coordinate system is transformed back to the (x,y,z) coordinate system using,

|  |  |  |
| --- | --- | --- |
|  |  | (3.9) |

The energy of the electron is then calculated by using,

|  |  |  |
| --- | --- | --- |
|  |  | (3.10) |

where

|  |  |  |
| --- | --- | --- |
|  |  | (3.11) |

where is the mass used in equation (3.3) to make the sub-valley spherically symmetric. Therefore, whenever a change from energy to momentum or vice-versa is required the mass for that particular sub-valley must be used. In this code the mass is always the ‘drift (conductivity) mass’ of the sub-valley or,

|  |  |  |
| --- | --- | --- |
|  |  | (3.12) |

where are the effective masses of the electron along the three mutually perpendicular directions that define the sub-valley.

 The scattering type is chosen by the method mentioned in the previous section. It would depend on the valley the electron is in at the time the scattering takes place as the scattering tables are different for different valleys. Therefore, it is necessary to keep track of which valley and which sub-valley the electron is in at all times. If the scattering type chosen is non-polar optical phonon scattering from valley 1 to valley 2 then the final sub-valley of the electron is randomly chosen from all the sub-valleys present in valley 2 as they are all at the same energy level and should therefore have equal probability of being scattered into. If there is non-polar optical phonon scattering within a valley then the final sub-valley of the electron will depend on whether f and g type scattering occurs or not. If there isn’t any f and g type scattering, then the final sub-valley is randomly chosen from the remaining sub-valleys in that valley. If f and g type scattering is present and if the f-type scattering process is chosen then the final sub-valley is randomly chosen from all remaining sub-valleys in the valley which are not in the same axis as the present sub-valley. In the case of g-type scattering the final sub-valley is the other sub-valley which lies on the same axis as the present sub-valley.

 At the end of every time step, the ensemble averages are calculated. This involves calculating the average drift velocity of the electrons, the average energy of the electrons and the number of electrons present in each valley and sub-valley. The average energy is calculated using,

|  |  |  |
| --- | --- | --- |
|  |  | (3.13) |

where is the energy of the ith electron. The average energies within a sub-valley ‘j’ in valley ‘i’ is,

|  |  |  |
| --- | --- | --- |
|  |  | (3.14) |

where is the energy of the kth electron in the jth sub-valley of the ith valley and is the number of electrons in the jth sub-valley of the ith valley.

 In most semiconductors, in order to properly simulate high field transport, it is necessary to consider more than 1 conduction band valley. To calculate the drift velocity along any direction, the effective mass along that particular direction is required. This makes it a little complicated to calculate the average drift velocity as different valleys are orientated differently in k-space and we only have the effective mass values along specific directions within each sub-valley. For example, in GaAs, a sub-valley of the L valley lies along the [111] direction. We know the effective masses along the transverse and longitudinal directions of the sub-valley but we do not know the effective mass along the [100] direction of that sub-valley. In order to calculate the drift velocity along the [100] direction, we transform the coordinate system to a system along which we know the effective masses and then switch back to the original coordinate system.

 Assume the Monte-Carlo simulation is run on the x,y and z coordinate system where the x-direction is [100] , y-direction is [010] and z-direction is [001]. Each sub-valley can be completely described by three mutually perpendicular axes. Let the three mutually perpendicular directions that describe the sub-valley be [a1,b1,c1], [a2,b2,c2] and [a3,b3,c3]. The electrons in the Monte-Carlo simulation will be drifted according to the x,y and z coordinate system so there will be kx , the component of the wave vector along [100], ky , the component of the wave vector along [010] and kz , the component of the wave vector along [001]. Therefore the total wave vector of the electron can be written as,

|  |  |  |
| --- | --- | --- |
|  |  | (3.15) |

The drift velocity along [a1,b1,c1] under the non-parabolic band approximation is calculated using,

|  |  |  |
| --- | --- | --- |
|  |  | (3.16) |

where is the electron dispersion relation for the electrons in the valley. Equation (3.16) is valid only for spherical valleys, therefore we have to convert the anisotropic valley to an isotropic valley by using the method described earlier. After making the valley isotropic the drift velocity is,

|  |  |  |
| --- | --- | --- |
|  |  | (3.17) |

where is the component of the wave vector along [a1,b1,c1] and is the effective mass of the electron along [a1,b1,c1], is the conductivity effective mass used to make the valley isotropic in equation (3.3) and is the energy of the electron in that valley. Similarly the drift velocity along [a2,b2,c2] and [a3,b3,c3] is given by,

|  |  |  |
| --- | --- | --- |
|  |  | (3.18) |

Using a simple transformation of coordinates from x,y and z coordinate system to the [a1,b1,c1], [a2,b2,c2] and [a3,b3,c3] coordinate system we get,

|  |  |  |
| --- | --- | --- |
|  |  | (3.19) |

The coordinates system is then transformed once again back to the x,y and z coordinate system to get the drift velocities along the x,y and z directions.

|  |  |  |
| --- | --- | --- |
|  |  | (3.20) |

As [a1,b1,c1], [a2,b2,c2] and [a3,b3,c3] are mutually perpendicular direction we have,

|  |  |  |
| --- | --- | --- |
|  |  | (3.21) |

For *N* electrons in the simulation the average drift velocity is then calculated as

|  |  |  |
| --- | --- | --- |
|  |  | (3.22) |

where , and are the drift velocities of the *i*th electron in the x,y and z directions and will depend on the sub-valley the electron is in at the time of the calculation.

 In a typical Monte Carlo simulation, there is always a certain amount of error in the final velocities even when steady state is reached. Therefore, simply taking the last value of the steady state velocity is inaccurate. Similar conclusion holds for the average energies as well. Therefore, an average is taken in the last ‘t’ seconds of the simulation over all the quantities and it is this average over time ‘t’ that is used to plot the velocity versus electric field plots to extract the mobility or the energy versus electric field plots. The amount of time ‘t’ used to take the average is a user defined value. It is usually one or two picoseconds after steady state is reached.

# RESULTS

The generalized Monte Carlo code was used to reproduce the characteristic results of certain materials to test its capability. There are a set of parameters for each material which are fitting parameters used to best fit the simulated data to the experimental data.

## Silicon





Figure 4.1 Energy of electrons in eV versus time in seconds (bottom panel) Drift velocity the of electrons in m/s versus time in seconds (top panel)

In figure 4.1 the usual plots of energy versus time and velocity versus time are shown. As can be seen steady state is achieved quite fast at around 2ps. In figure 4.1 the saturation of the velocity with the increase in electric field can be seen with the last two values of electric field plotted.



Figure 4.2 Energy of the electrons versus the applied electric field. Experimental data is taken from [29].

 In figure 4.2 the velocity for different electric fields is plotted and compared with experimental data. As can be seen there is very good agreement between the experimental values and the values obtained from the simulation.

 In figure 4.3 a similar plot is plotted between energy and electric field. In all these plots the electric field is applied in the [111] direction. As can be seen again there is a very good agreement between the experimental values and those obtained from the simulation.



Figure 4.3 Drift velocities of the electrons versus the applied electric field. Experimental data is taken from [29].

## Germanium





Figure 4.4 Drift velocity versus time for Germanium (top panel). Average energy vs. time (bottom panel).

The velocity of the electrons and the energy of the electrons are plotted against time in figure 4.4. In germanium the number of valleys chosen is 3, the L valley, Gamma valley and the X valley. Due to this the simulation takes longer to reach steady state as there will be transfer of electrons between the valleys and a transfer of energy between the valleys. This can be seen when these curves are compared with those obtained with silicon which uses just 1 valley (the X valley) in the previous section.



Figure 4.5 Population of Valleys versus time at 6kV/cm in Germanium

In figure 4.5 the number of electrons in each valley is plotted versus time. As can be seen there are almost no electrons in the gamma valley as its effective mass is really small. This causes a low density of states causing a low probability of scattering into that valley.



Figure 4.6 Energy versus Applied electric field in Germanium. Experimental data is taken from [30].

In figure 4.6 and figure 4.7 the energy of the electrons and the steady state drift velocity of the electrons are plotted against the applied electric field and compared with experimental values. As can be seen there is a good agreement between the experimental data and the simulated data. It is also possible to obtain even more accurate results by tweaking the fitting parameters further more.



Figure 4.7 Drift velocity versus applied electric field in Germanium. Experimental is data taken from [30].

## Gallium Arsenide

In figure 4.8 the drift velocity and the energy of the electrons are plotted against time. Just as is the case in germanium, in gallium arsenide there are 3 valleys. Therefore the simulation takes a longer time to reach steady state as can be seen in these two figures. In figure 4.9 and figure 4.10 the energy of the electrons and the steady state drift velocity of the electrons are plotted against the applied electric field and compared with experimental values. In figure 4.11 the fraction of electrons in the L valley is plotted against the applied electric field.



Figure 4.8 Energy versus time in Gallium Arsenide (top panel). Velocity vs. time (bottom panel).



Figure 4.9 Drift velocity of electrons versus applied electric field. Experimental data is taken from [31].



Figure 4.10 Energy of electrons in the gamma valley versus applied electric field. Experimental data is taken from [32].



Figure 4.11 Fraction of electrons in the L valley versus applied electric field. Experimental data is taken from [32].

# CONCLUSIONS

A research and educational tool has been created to solve the Boltzmann Transport Equation for bulk systems using the Ensemble Monte Carlo method. The tool gives the user the option and freedom to vary parameters within limits determined by experimental values of the measured coupling constants and effective masses.

 The tool uses a non-parabolic band approximation and incorporates most types of scattering rates. As of now it has deformation potential scattering (acoustic and optical), polar optical phonon scattering, piezoelectric scattering and ionized impurity scattering. Using non-parabolic bands makes the simulation as accurate as possible without considering a full band relation. Since it uses an analytical relation between energy and momentum of the electron the simulation time is low, making it more user friendly as an online tool.

 There are some improvements that can be added to this tool. As was mentioned earlier, the tool can be easily updated with regard to adding new materials or changing the input interface. There is also a plan to add the scattering rates plots to the output so that the user can see which scattering types dominate and which do not. A possible but not necessary extension of the tool is incorporating a full band simulation instead of the non-parabolic band approximation that was used here. The two have advantages and disadvantages. The non-parabolic band model is definitely not accurate for very high applied fields. But the full band calculation, being an equilibrium calculation is also inaccurate in representing conduction bands, in particular those that lie high in energy. This limitation of the full band models is not always clearly stated in the literature and amongst the scientific community. Modeling of holes with a full-band calculation is a must as holes are accurately represented with a full band structure. Thus adding hole transport with a full-band model is a possible extension of the tool.

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