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Studies on the electrical transport properties of carbon nanotube composites

Taylor Warren Tarlton

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STUDIES ON THE ELECTRICAL TRANSPORT PROPERTIES OF CARBON NANOTUBE COMPOSITES

by

Taylor Warren Tarlton, M.S. Applied Physics, B.S. Physics

A Dissertation Presented in Partial Fulfillment of the Requirements for the Degree Doctor of Philosophy in Engineering: Engineering Physics

COLLEGE OF ENGINEERING AND SCIENCE LOUISIANA TECH UNIVERSITY

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We hereby recommend that the dissertation prepared under our supervision by Taylor Warren Tarlton entitled Studies on the Electrical Transport Properties of Carbon Nanotube Composites be accepted in partial fulfillment of the requirements for the Degree of Ph.D. Engineering-Engineering Physics.

Supervisor of Dissertation Research

Head of Department

Department

Advisory Committee

Approved:

Director of Graduate Studies

Dean of the College

Approved:

Dean of the Graduate School
ABSTRACT

This work presents a probabilistic approach to model the electrical transport properties of carbon nanotube composite materials. A pseudo-random generation method is presented with the ability to generate 3-D samples with a variety of different configurations. Periodic boundary conditions are employed in the directions perpendicular to transport to minimize edge effects. Simulations produce values for drift velocity, carrier mobility, and conductivity in samples that account for geometrical features resembling those found in the lab. All results show an excellent agreement to the well-known power law characteristic of percolation processes, which is used to compare across simulations. The effect of sample morphology, like nanotube waviness and aspect ratio, and agglomeration on charge transport within CNT composites is evaluated within this model. This study determines the optimum simulation box-sizes that lead to minimize size-effects without rendering the simulation unaffordable. In addition, physical parameters within the model are characterized, involving various density functional theory calculations within Atomistix Toolkit. Finite element calculations have been performed to solve Maxwell’s Equations for static fields in the COMSOL Multiphysics software package in order to better understand the behavior of the electric field within the composite material to further improve the model within this work. The types of composites studied within this work are often studied for use in electromagnetic shielding, electrostatic reduction, or even monitoring structural changes due to
compression, stretching, or damage through their effect on the conductivity. However, experimental works have shown that based on various processing techniques the electrical properties of specific composites can vary widely. Therefore, the goal of this work has been to form a model with the ability to accurately predict the conductive properties as a function physical characteristics of the composite material in order to aid in the design of these composites.
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Author

Date

July 13, 2016
DEDICATION

This work is dedicated first to my Lord Jesus Christ for his many blessings and secondly to my beautiful wife for her steadfast love and encouragement throughout this process. I would also like to thank my wonderful parents for how they have raised me to always strive to be my best at whatever I pursue in this life.

- Taylor Warren Tarlton
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CHAPTER 1

INTRODUCTION

1.1 Background

Carbon Nanotubes (CNTs) were first discovered and described in 1991 by Sumio Iijima of the NEC Corporation, Fundamentals Research Laboratories in Japan. The discovery came by happenstance through an attempt to create $C_{60}$ molecules through arc discharge evaporation.[1] Since then, CNTs have been intensively studied as they are of interest for a number of technological applications due to their exceptional electrical, mechanical, and thermal properties.[2]–[8] CNTs can be blended into other materials to improve certain characteristics of those materials. When these nanocomposite materials are created, the overall composite can acquire some of the quality characteristics inherent in CNTs, improving thermal properties, electrical conductivity, and/or mechanical characteristics depending on the specific application.

CNTs can be either metallic or semiconducting based on the chirality of the CNT, which is essentially a quantification how a sheet of graphene (hexagonally arranged carbon atoms) could be rolled to obtain the specific CNT. CNTs can consist of one rolled sheet in single-walled CNTs (SWCNT), or many concentric sheets in multi-walled CNTs (MWCNT). The chiral vector, $(m,n)$, determines how the sheet of graphene would be rolled to form a CNT and is defined based on the unit vectors the hexagonal lattice. It has been found that when $m$ is equal to $n$, or $m-n$ is a multiple of 3, the CNT will have a
minimized gap between the valence and conduction bands and behave as a metal. Otherwise, the arrangement of carbon atoms induces a larger gap, causing the CNT to become semi-metallic in nature. CNTs are comprised of conjugated carbon atoms, which allow electrons to disassociate from specific atoms and occupy delocalized electronic states. The semiconducting or metallic nature of each CNT then determines the availability of those delocalized electronic states. In addition, the lengths of CNTs are on the order of the mean free path of electrons, leading to ballistic charge transport along the axial direction of the CNT.[2], [5], [7] Conductivity values of single CNTs have been reported to be on the order of $10^5$ S/m to $10^6$ S/m.[9], [10]

Thermal conductivity of CNTs is exceptional along its axial direction due to the ability of phonons to move along that direction. Values up to 6000 W/mK for thermal conductivity has been reported in the literature for CNTs. However, many factors such as atom arrangement, chirality, diameter, and surface defects play an important role in determining thermal conductivity of individual nanotubes (NTs).[4], [8]

CNTs also have quality mechanical properties along the axial direction of the NT, lending them exceptional tension strength. The Young’s Modulus of CNTs has been measured to be 1 – 5 TPa along the axial direction.[11] They also possess a high degree of flexibility, with the critical bending angle for SWCNTs being measured around 110°.[12]

CNTs have been widely studied for various purposes in order to take advantage of their exceptional electrical, mechanical, and thermal properties. Numerous studies have sought to use them as components in transistors as the semiconducting material separating the source and drain components in attempts to further scale down the size of
modern field effect transistors (FETs). Some studies have been quite successful and have even led to small scale computers consisting of only CNT-FETs. Much of the work involving CNTs has focused on how the inclusion of CNTs into various polymers, ceramics, and epoxies can produce enhancements in electrical properties while also often times improving upon the mechanical and thermal properties of the original material. In these studies, CNTs are used as conductive fillers in a specific type of matrix, and then the electrical properties of the overall composite are characterized versus the volume or weight fraction of CNTs. These conductive composites can make matrix materials that are insulating without the CNTs useful for a wide array of applications ranging from electromagnetic interference shielding, electrostatic reduction, and strain sensing for structural health monitoring.

Many of the experimental studies on CNT composite materials cite the importance of proper CNT dispersion for the qualities of the CNTs to be fully exhibited through the composite materials. CNTs interact with other CNTs via Van-der-Waals interactions, and therefore tend to coagulate, or agglomerate. These bundles can, and in early studies did, degrade the electrical conductivity in these composite materials. Many studies have gone through repeated steps of sonication of samples, functionalization of CNTs, shear mixing, and other processing methods in order to better disperse CNTs throughout the host matrix materials.
The main reason quality dispersion is desired for various electrical applications is that CNT composites consisting of electrically conductive CNTs and an insulating matrix generally behave in close agreement with percolation theory. [41], [66], [80]–[83] At a critical CNT concentration (percolation threshold), conductive networks begin to form throughout the host matrix, allowing charge transport through the previously non-conductive material and increasing the overall conductivity by several orders of magnitude. Better dispersion of CNTs can enable more conduction pathways to form throughout the composite material, allowing for either heat or charge transport to occur. Materials previously not useful for electronic applications have exhibited conductivities ranging from $10^{-7} - 10$ S/m when mixed with CNTs, an increase of around 10 orders of magnitude in some cases when compared to the bare matrix. [18], [37], [84]–[88]

A percolating material will typically follow the well-known power-law given by Eq. (1.1).

$$\sigma = \begin{cases} 0 & \phi < \phi_c \\ \sigma_c (\phi - \phi_c)^{t} & \phi > \phi_c \end{cases} \quad (1.1)$$

Here, $\phi$ is the filler concentration, $\phi_c$ is the critical concentration at which percolation occurs, $\sigma_c$ is the conductivity of the nanoinsert, and $t$ is the critical exponent. Given a particular system, its conductivity curve can be fit to Eq. (1.1) with $\sigma_c$, $\phi_c$, and $t$ to be used as fitting parameters. [81], [82] The value of $t$ found from various experimental works ranges between 1 and 6, with the majority of the values being less than 4. See for instance the review by Bauhoffer et al., certainly not the only source, for a more detailed discussion. [19]

The various experimental studies have shown that functionalization[32], [41], [56], [73], [89], type of CNT[56], [90], aspect ratio[25], [56], [80], host matrix
material, dispersion, alignment, and other processing variables can all have significant impacts on the electrical properties of the composite. Since so many variables can play a role in determining the conductive properties of CNT composites, accurate and predictive modeling is needed to better understand specifically how each variable contributes to the nanocomposite properties for widespread application to be realized.

There have been various models presented in the literature attempting to understand the mechanisms behind charge transport within CNT composites, a majority of which reduce the composite with randomly dispersed NTs to a resistor network. These models typically include a resistor for each CNT within the sample, resistors for CNT junctions, and resistors for the gaps separating CNTs, which can include quantum approximations. The goal is then to find an equivalent resistance for the entire sample. This approach is somewhat convenient as the extremely difficult problem of finding the voltage drop in each CNT-CNT junction is avoided. However, its implementation is limited to samples with a relatively small number of NTs and junctions. The work by Bao et al., which used a resistor model to study the electrical properties of CNT composites, studied samples that were 5μm in length. However, the NTs themselves were 4.5 μm, which led to a sample with relatively few NTs.

The modeling work by Xu et al. took a different approach to studying these materials. In that study, CNT composites were generated into a representative epoxy matrix, but modeled by solving Maxwell’s electrodynamics equations using a finite element method approach. The current density was then found for the system, but under
the assumption that tunneling is modeled as a resistor, which is very similar to the resistor models presented above.[98]

Other groups have employed an excluded volume approach to model these CNT composites. This type of model is a more mathematical approach towards understanding percolation theory and how the inclusions being studied lead to percolation. In works by Berhan et al.[99] and Sun et al.[100], the fillers were given an excluded volume, meaning a volume that cannot be overlapped by any other exclusion. This excluded volume is then used to find the critical volume concentration necessary for percolation of the CNT network to occur. Each of these studies incorporated various parameters describing the characteristics of the filler material, such as NT aspect ratio and waviness. Berhan et al. employed a soft shell approach to incorporate the possibility of tunneling. This soft shell approach, which they state had been used in other studies modeling percolation, incorporates tunneling by allowing the inclusions to have a soft outer shell that can be penetrated by the other inclusions.[99]

The types of models presented thus far only account for complete percolation paths and are unable to describe some important phenomena such as charge traps, which are locations where charges become trapped within the sample and do not reach the electrodes. Certainly those paths do not contribute to the conductivity of the sample, but because their presence negatively affects the conductive properties of these materials, studying their effect is important to help in the design of nanocomposites. These models, however, have been beneficial in helping understand how NT waviness, aspect ratio, tunneling barrier height, and agglomeration affect conductivity in composite samples.[25], [54], [92]–[100]
An experimental work performed in 2011 by Shang et al. studied the conductivity of stretchable multi-walled CNT/Polyurethane composites for various temperature ranges. The study concluded that the temperature relationship with conductivity indicated that charge transport within that composite was in fact dominated by a three-dimensional hopping mechanism, where charges hop between localized sites, which in the case of that study would have been the CNTs inside the polyurethane.[101] It is therefore sensible that one could study charge transport within these materials via a hopping process where delocalized electrons are able to hop between conductive CNTs within an insulating matrix. The main task with employing any hopping algorithm then becomes appropriately modeling the fundamental physics underlying the hopping transport, which is captured through transition rates that in turn determine the rate for charge transport.[102]–[105]

Charge-hopping models have been implemented for various semiconducting and conductive polymer materials.[104]–[107] These hopping models study charge transport by allowing charge carriers to hop between sites within the material. Historically, charge hopping algorithms have sought to explain charge transport in systems where there are donor and acceptor sites in the system, as in a doped semiconductor.

One of the difficulties with a model such as the one introduced in this dissertation, is the determination of the local electric field inside the composite material. The local electric field depends on many factors such as the dielectric nature of the materials in the composite as well as the shape and special distribution of the filler materials.[108] There have been many studies focused on combining the properties of the matrix materials to determine the effective properties of the composite material using effective medium theory, which has been used to determine thermal, mechanical, and electrical properties
A study in 1988 by Zeng et al. used Eq. (1.2) as the effective medium approximation for the dielectric properties of a composite with two materials having different dielectric properties.

\[
\rho_1 \frac{\varepsilon_1^{(0)} - \varepsilon_e^{(0)}}{\varepsilon_e^{(0)} + g(\varepsilon_1^{(0)} - \varepsilon_e^{(0)})} + (1 - \rho_1) \frac{\varepsilon_2^{(0)} - \varepsilon_e^{(0)}}{\varepsilon_e^{(0)} + g(\varepsilon_2^{(0)} - \varepsilon_e^{(0)})} = 0
\]  

(1.2)

Here, \( \rho_1 \) is the volume fraction of material one, \( \varepsilon_1 \) and \( \varepsilon_2 \) are the dielectric functions of materials 1 and 2 respectively, \( \varepsilon_e \) is the effective dielectric function of the composite material, and \( g \) is a geometrical factor dependent on the shape of the inclusions. This effective dielectric function is obtained by solving the above quadratic equation.

Other studies have also shown various mathematical formulations for other effective medium approximations. In terms of electrical properties of the composite, the various formulations attempt to take into account the polarizability of each material, along with some type of factor for the geometry of the inclusions. Each method is also dependent on the volume fraction of the inclusion material. Effective medium approximations lead to a simple usable quantity for describing how the composite material, as a whole, will behave under an applied electric field. However, these methods are not able to describe the local electric fields within a composite material that are needed for this work.

In 1997, Cheng et al. performed a numerical study on two-dimensional samples with various types of inclusions to understand how the electric field behaves in local areas compared to an applied electric field. In the study, Maxwell’s Equations were used along with the dielectric constants of both materials in order to find the value of local electric fields at 200,000 random points within a sample. The study showed that the value of the local electric field inside the composite can be larger or smaller than the
applied electric field depending on the location within the sample. However, for the various types of inclusions, a distribution of electric field values in the composite was found to be between $0E_0$ and $1.5E_0$, where $E_0$ is the applied field, depending on the relative dielectric constants of the matrix and inclusion materials. This study did state however that there would exist so-called “hot spots” for the local electric field within the sample and may tend to be near the edges of the inclusion material if that inclusion material is much more polarizable than the matrix material.

1.2 About This Work

In this work, a model is presented with the capability to study in depth how various material properties affect charge transport within CNT composites, both before and after percolation is reached. Routines have been developed to first generate virtual composite samples and then to study charge transport within those samples using a hopping algorithm similar to the hopping models mentioned previously. In addition, the model has the ability to study samples large enough, in terms of size and number of nanoinserts, to appropriately model system sizes in modern day micromanufacturing labs without periodic boundary conditions in the transport direction. Incorporated into this model are exceptional visualization capabilities through a program that has also been developed in house, with details provided elsewhere.[114] This model can be used to study how CNT waviness, CNT aspect ratio, sample size, dispersion, agglomeration, and the host matrix all combine to determine charge transport within a CNT composite. Besides modeling the transport process with routines written in the Matlab programming language, calculations have been performed in various other modeling software packages
to better understand the underlying physics and to calculate parameters used in the hopping model.

The routines that have been used for charge transport have gone through various improvements throughout the progress of this work. These improvements are significant enough to mention as they allow for larger composite samples, with more NTs, to be simulated more efficiently. To the best of this researcher’s knowledge, the routines developed for these simulations allow the study of samples with the number and aspect ratio of NTs much larger than those capable with other models.[95], [100] This is significant as CNTs are high aspect ratio nanofillers, and therefore in order to better model them, one must be able to include very large aspect ratio CNTs in the simulation.
CHAPTER 2

GENERAL METHODOLOGY

Metropolis Monte Carlo routines have been developed in the Matlab programming language so that they can be run on high performance computing systems. The model in its current state was built on a previous version of the routines, which that have been optimized on several levels. The contributions from this dissertation and how they improve what was implemented beforehand, is indicated in the following sections.

It is well known that Monte Carlo simulation techniques can lead to results that vary due to the randomness of the technique. In order to assure that results presented in this work are statistically accurate, several replicas of the simulation box, all statistically independent from each other, are created and the results averaged. In addition, certain portions of the simulation processes have been parallelized in up to 12 parallel processors in order to increase simulation efficiency. Simulations are composed of three main routines: A generation routine, a neighbor finding routine, and a charge transport routine. Overall, these simulations are generally performed on HPC servers. Two main clusters have been used, the Cerberus cluster, which is a local Louisiana Tech University supercomputer, and the Queen Bee 2 cluster available through the Louisiana Optical Network Initiative (LONI). The Queen Bee 2 cluster is a 1.5 Petaflop system that was commissioned in 2014. At that time, it was ranked as the 46th most powerful supercomputer in the world.
Parameters necessary for these simulations were obtained using a set of other codes and modeling algorithms that include Density Functional Theory, Green’s Functions, and the finite-element-based solutions of the Maxwell’s equations as implemented in a number of commercial codes, particularly Gaussian 09[115], Atomistix Toolkit[116], and COMSOL[117].

2.1 Generation of Virtual Nanocomposite Samples

In order to study electrical properties of CNT composites, virtual samples are generated to a specific size and with specific NT volume concentrations using a Metropolis Monte Carlo algorithm. By growing to specific volume concentrations, accurate comparisons can be made between various simulations.

To start this generation process, a sample of a certain length, width, and height is partitioned into a three-dimensional mesh with the distance between each grid point equal to the specified NT diameter. Therefore, the size of the NT determines the coarseness of the mesh. A seed grid point is chosen randomly as the starting point for growing a NT. A random walker then grows each NT segment by segment along the three-dimensional grid. The random walker starts at the seed grid point and then “walks” to neighboring points based on probabilities defined by user-specified input parameters. Grid points define the start and end point of each of the segments that make up a NT segment. However, each segment can span a user-defined number of grid points. If a point that already hosts a segment is chosen for growth, the choice is rejected and the segment is regrown. All NTs are given a growth orientation in the x, y, or z directions. Waviness is introduced by giving each new segment a probability, based on input parameters, to grow.
along the specified orientation, or to deviate from that orientation. The waviness of NTs is quantified through a waviness factor, $WF$, given in Eq. (2.1).

$$WF = 1 - \left( \frac{\sum_{i=1}^{N_t} \Delta D_i}{L_i} \right) / N_t$$  \hspace{1cm} (2.1)

Here $\Delta D_i$ is the distance each NT spans along the specified orientation, $L_i$ is the length of NT $i$, and $N_t$ is the number of NTs in the sample. Defined in this way, a higher waviness factor indicates a sample with wavier NTs.

The algorithm to generate a nanocomposite sample that was described above is an adaptation of the method reported by Evans et al.\[18\] In that work, aggregates of nanofillers were generated into a volume space using a Metropolis Monte Carlo method. Each nanofiller was considered one grid cell within the volume space and those grid cells were filled randomly such that aggregates arise when neighbor cells were filled. For this work, the same method was followed except that while growing a NT, subsequent cells are randomly chosen, but only among cubes that are a neighbor of the last that was filled. It is only when a NT is completed that a new random location is selected to begin growing the next NT. Figure 2.1 shows a simple schematic of this process.

![Figure 2.1. Schematic representation of the NT's growth process segment-by-segment. In the simulation, the boxes are considered completely occupied by the NT segment.](image)
The generation of the NTs along a three-dimensional grid system allows the user to generate samples in a very time-efficient manner. Other approaches have taken a slightly different strategy to generating nanocomposite samples, in which NTs are still grown segment by segment, but not on a grid system.[25], [95] In this way, NTs are free to grow in any direction, with respect to the previous segment, within some limiting values. The main issue in this method of generation is that after each segment is placed, there is a need to check for overlap with neighboring NTs. This is an essential check if accurate volume concentrations of nanofillers are to be reported. However, this type of generation method can lead to time-consuming simulations, as NTs begin to repeatedly overlap with pre-existing NTs at higher volume concentrations. By growing along a grid, it is sufficient to check if the box where the next segment is to be place is already occupied or empty.

Throughout the generation process, periodic boundary conditions are placed in the two directions perpendicular to the applied electric field in order to minimize edge effects at the sample boundaries. In order to do this, NTs are allowed to grow out one side of the sample, and into the opposite side of the sample. In this way, there is a seamless growth process, which essentially removes the boundaries of the sample perpendicular to charge transport.

2.2 Finding nearest Neighbor Nanotubes

In order to be able to study charge transport, nearest neighbors must be identified for each NT. If the minimum distance between 2 of the segments is smaller than a maximum neighbor distance cutoff, then the NTs are considered neighbors. The minimum distance and the points on each of the NT surfaces, where the distance is a
minimum are then stored. The neighbor distance cutoff in these simulations was typically set at 2 nm, as that is generally where the probability for carrier hopping becomes practically zero (see Section 2.4). The neighbor-finding routine has generally been the most time-consuming portion of the entire simulation process due to the number of calculations needed. As such, the neighbor-finding routine has been modified for this work.

This routine first partitions the sample so that each NT is associated with some portion of the sample. In the original version, only NTs in the same partition or a neighbor partition are checked, in that way unnecessary comparisons between NTs that are too far away are not unnecessarily compared. However, since the size of the partitions must be such that whole NTs comfortably fit, but also needs to be an integer divisor of the sample size, the minimum allowable partition size has generally been set equal to around two times the maximum length of a NT. So even when only neighbor partitions are considered, NTs in those partitions could be much further than the cutoff distance and some unnecessary comparisons could be conducted. A work around was implemented where each partition is allowed to overlap into surrounding partitions by the neighbor cutoff distance. In this way, only NTs in the same partition (including the fuzzy boundaries) need to be counted. Still, it is possible that a NT starts in one partition but extends to a neighbor partition. In such a case, NTs in both partitions (including the fuzzy boundaries) are included in the comparison. Figure 2.2 shows a schematic representation of the partitioning process.
Figure 2.2. Schematic representation of overlapping partitions for neighbor finding routine.

If the simulation is to be run on one processor, each NT can be assigned an identity with its partition or partitions. Therefore, each new NT in the list is only checked with other NTs that have the same identity. However, since these routines are setup in the Matlab programming language, the built in parallelization techniques in Matlab have been used in order to parallelize the neighbor finding routine so that it can be split into the same number of processes as the number of partitions, thus decreasing the overall simulation time. The routine has been coded such that a SPMD loop in Matlab can set up a parallel processing pool with a specified number of workers, equal to the number of partitions in the sample. In the Matlab programming language, an SPMD loop is a parallel processing for-loop where each separate parallel processor performs the same calculation. For example, if there is a calculation to be performed on a variable that has 1000 elements, then an SPMD loop with, for example, four processors can be used so that each processor can perform 250 of these calculations. Now one of the key components of an SPMD loop is that each new calculation is not dependent on any calculations being
performed in each of the other processors. Once the 250 calculations have been performed on each of the 4 processors, the resulting data can be recombined at the end of the SPMD loop.

When these simulations are performed on an HPC system, the user must request a certain number of processors for the simulation. This means that the user must know exactly how many processors are to be used for a simulation for optimal usage of resources. Within this portion of the simulation, the number of processors is generally set up to be equal to the number of partitions needed for the sample, which is again based on the length of each NT, as if one processor is being used. The one caveat here is that this calculation, while simple, must be performed before beginning the simulation to know the appropriate resources to request from the HPC system.

The maximum number of parallel workers available with the current Matlab licenses is 12, hence the samples can be either partitioned into either 1, 2, 4, 6, 8, or 12 partitions. Each worker in the Matlab parallel processing pool then takes a partition, and finds the neighbors of each NT in that partition. For the case of a NT that crosses partition boundaries, it will be checked on separate processors for neighbors in each partition. There is then some necessary overhead in organizing the resulting neighbor information gathered from each of the parallel workers so that it can be output to a file.

In order to characterize the improvement in the neighbor finding process, calculations were run on an HPC cluster to find the neighbors of NTs in a sample with dimensions of 1x0.2x0.2 μm³ and over 5000 NTs within the sample. Four separate calculations were performed using 1, 4, 8, and 12 parallel processes. Figure 2.3 shows the
wall-time that each simulation took to manage overhead, split up processes, and then combine the calculations done over the various processors.

![Graph showing CPU wall times for calculating neighbor NTs for a specific sample with over 5000 NTs.](image)

Figure 2.3. CPU wall times for calculating neighbor NTs for a specific sample with over 5000 NTs.

It can be seen that the amount of time to complete the calculation drops significantly between 4 and 8 processors, illustrating how the use of more parallel processors can speed up the calculation of neighboring NTs. When 4 processors were used, the simulation time went up by almost 10 hours. This illustrates the fact that if 1 processor is used, the number of partitions is set based on the length of the longest NTs, so that partitions may be of appropriate size to efficiently split up NTs into enough partitions. Also, overhead was needed to set up the parallel processes or to recombine the data post-parallelization. Additionally, 4 processors, and therefore 4 partitions, may not have been the most optimal number of partitions. However, when 8 cores were used, NTs were more efficiently sorted into partitions, and as a result, even with the overhead of parallelization, the simulation time decreased to 7.79 hours. The wall time for twelve processors was 7.73 hours, indicating that there would not be much improvement from using 8 processors/partitions. In order to determine the appropriate number of processors
to use, the user would need to determine a number of partitions such that each one is not too small, or too large, and simulations can be parallelized efficiently. This improvement to the model has been a recent development, and further improvements can be made, such as allowing each parallel processor to handle multiple partitions instead of a single partition.

2.3 Nanotube Length Distributions

In these samples, NTs can be grown with uniform lengths or with a given length distribution. However, if a NT finds an obstacle to continue growing before reaching the target length, after a few attempts, it is allowed to stop growing short of the target length. This is a problem if a sample with equal sized NTs is being created, as some NTs will not have the desired length. However, it can be seen in Figure 2.4a that this leads to a distribution heavily peaked at the target NT length with a very low concentration of NTs of other lengths, even at high concentrations. This problem is avoided or minimized when samples with a given length distribution are considered by placing the larger NTs first, so that if a particular NT cannot be placed and it is cut short, it will be counted against the number of NTs to be generated at that shorter length. As seen in Figure 2.4b where a normal distribution of tube lengths was attempted, the distribution of NT lengths at different volume concentrations cannot be distinguished from each other. Distributions shown in Figure 2.4 are averaged over 100 samples.
Figure 2.4. NT length distribution (number of segments) presented as a percentage of NTs in the composite with a particular number of segments (a) for a sample with NTs of uniform length (b) For a sample with NT's length with a Gaussian distribution. In b) the different percentage cases are indistinguishable from each other.

2.4 Hopping Transport Algorithm

Once samples have been generated, virtual electrodes are placed on opposing ends to apply an electric field in the x direction and begin the charge transport process. In the hopping formalism, the electric field alters the energy landscape of the system so that charge flow is biased in its direction.[103] Periodic boundary conditions (PBC) are implemented in the directions perpendicular to the field. These particular set of boundary conditions lead to a process where a carrier is injected into the composite from one electrode and hops around in this 3D network until it is collected at the other electrode located opposite to the injection electrode along the x direction. Then a new carrier, statistically independent from the previous one, is injected. When, the carrier reaches one of the sidewalls, PBC allow the charge to continue moving laterally. Using PBC, the edges of the sample are effectively removed so that charges are not confined to the lateral dimensions of the sample, therefore minimizing edge effects.[119]
Over the course of this project, the hopping algorithm has undergone modifications. In the original version, developed by a former student, some of the parameters were used as fitting parameters. A contribution to the field of this dissertation is the development of protocols to determine all fitting parameters from first principles or known properties. However, the original approach (with fitting parameters) was used extensively in the studies reported in Chapter 3 of this work, and led to simulation data that agrees well to experimental results. However, improvements were made to the fundamentals of the hopping algorithm so that charge transport in the system is completely dependent on the underlying physics of the system without fitting parameters. The remainder of this chapter will focus on the original routine. Chapter 4 goes into more detail about the characterization of certain parameters, while also introducing the improvements to the transport routine that have been made over the course of these studies.

Within this version, the conduction is considered ballistic within the NTs, while carriers hop between neighbor NTs across the smallest gap between them. The number of neighbors for each NT, the (closest) distance to each of them, and the x-y-z coordinates of the point where the distance is a minimum are stored in a neighbor table.

The transport process starts by inserting a carrier from one of the electrodes into one of the NTs in contact with that electrode. Then the hopping rate, \( \omega_{ij} \), to each of the neighbor NTs is calculated as described in Eq. (2.2). A normalized cumulative distribution function (CDF) is then constructed from these values to determine the hopping probability to each of the neighbor NTs. Once the CDF is created, a random number is sorted and compared to the normalized CDF to decide to which neighbor NT
the charge will transfer. In this way, charges are more likely to transfer to neighboring NTs with higher associated hop rates.

\[ \omega_{ij} = |V_{ij}|^2 e^{-\frac{\Delta G^0 + |\Delta G^0|}{2k_B T}} e^{\frac{ed \cdot \vec{E}}{k_B T}} \]  

(2.2)

In this equation, \( V_{ij} \) is the transfer integral, accounting for the overlap of the wave function of neighbor NTs and \( \Delta G^0 \) accounts for the difference in the energy of the carrier in the NT where it comes from and the energy of the carrier in the NT to which it hops. \( ed \cdot \vec{E} \) is the work done by the electric field on the carrier (and thus the change in energy of the carrier) as it moves a distance, \( d \), inside a NT. \( T \) is the temperature of the system, and \( k_B \) is Boltzmann's constant.

\( V_{ij} \) in Eq. (2.2) the transfer integral is given by Eq. (2.3).

\[ V_{ij} = J_0 e^{-\alpha R_{ij}} \text{ or } |V_{ij}|^2 = \nu_0 e^{-2\alpha R_{ij}} \]  

(2.3)

Here, \( \alpha \) is the tunneling coefficient and \( R_{ij} \) the distance between the two NTs, \( i \) and \( j \), connected by the hop. The \( \nu_0 \) term is the attempt-to-hop frequency and is dependent on the electron-phonon coupling strength as well as the phonon density of states among other material properties. Values for the attempt-to-hop frequency can be found in the literature for conductive polymers (see for instance [107]), but these authors have been unable to find values for CNTs. This attempt-to-hop frequency, however, is a pre-factor that was used to set the time scale. For this version of the hopping routine, It was found that an attempt-to-hop frequency of \( \sim 10^7 s^{-1} \) leads to a time scale in reasonable agreement with experimental values in reference [120], so this value was used for all simulations where this version was used. The exponential in Eq. (2.3) is the tunneling probability derived from quantum mechanical calculations using the well-known Wentzel–Kramers–
Brillouin (WKB) approximation in the limit of a high and wide barrier. Indeed, the host matrix material will have an effect on charge transport via the quantum phenomenon of charge tunneling. Previous work has been done to quantify charge tunneling between graphene sheets through a polymer layer where a 4-fold decrease in the tunneling constant, compared to vacuum, was observed for Poly Acrylonitrile (PAN). The same process can be employed to calculate tunneling parameters for any material, giving the model the capability to study how these various materials affect charge transport without experimental input.

As said above, $\Delta G^o$ in Eq. (2.2) accounts for the change in the carrier energy when going from one NT to the next and it is given by Eq. (2.4).

$$\Delta G^o = \varepsilon_i - \varepsilon_j$$  \hspace{1cm} (2.4)

$\varepsilon_i$ and $\varepsilon_j$ are the energy of the carrier in each NT. In all results reported in this dissertation, all NTs are modeled as metallic NTs, thus carriers have energies that can differ from the Fermi level with Boltzmann probability, as given in Eq. (2.5). This allows the charge carrier on a specific NT to have an energy slightly above or below the Fermi energy level due to thermal affects.

$$F(\varepsilon) = e^{-\frac{(\varepsilon-E_f)}{k_B T}}$$  \hspace{1cm} (2.5)

$E_f$ is the Fermi level and for this particular work it is considered the same in all NTs.

This is not a limitation of this model, but a choice for the calculations reported here as each nanoinsert can be given its own Fermi Level.

The effect of an electric field on the carrier's energy is accounted for by further modifying the carrier energy $\varepsilon$ in each NT by the electrical potential energy provided by the field, as can be seen in Eq. (2.6).
Here $\epsilon_i$ and $\epsilon_j$ are the energy of the charge in each NT taken from Eq. (2.5) before the applied electric field is taken into account, $e$ is the carrier, $E_x$ is the x-component of the electric field at each particular junction, and $\Delta x$ is the distance the charge travels in the direction of the field. Unlike the random energy fluctuations caused by temperature, the electric field will result in a bias to charge flow in one direction. Notice that Eq. (2.6) accounts for the effect of the electric field when the carrier hops across a junction between NTs, while the second exponential in Eq. (2.2) accounts for the difference in electrical potential energy between the point the carrier enters the NT and the point where it leaves it to hop to the next NT. It should be noted that this exponential will either exponentially increase or decrease based on the difference in the electrical potential energy. It therefore becomes highly unlikely that a charge will attempt to move down a CNT against the force of the electric field.

The hop rates for the neighboring NTs calculated from Eq. (2.2) are then used to determine a transition time for the carrier according to Eq. (2.7).

$$t_{ij} = \frac{-\ln(\zeta)}{\sum \omega_{ij}}$$

Here, $\zeta$ is a random number uniformly distributed between 0 and 1. The mobility of a charge reaching the opposite electrode is calculated as in Eq. (2.8):

$$\mu = \frac{d}{t \cdot E} = \frac{\nu_{drift}}{E}$$

Where $d$ is the distance across the sample in the direction of the electric field, $t$ is the total time it takes the charge to reach the collection electrode, $E$ is the electric field, and $\nu_{drift}$ is the drift velocity. Each carrier moves through the sample until it reaches the opposite electrode unless a specified time cutoff or maximum number of hopping
attempts is met. If one of those limits is reached, the charge is no longer tracked and a new charge is generated. The mobility is averaged among those carriers that reach the opposite electrode as shown in Eq. (2.9).

$$\mu_{avg.} = \frac{1}{n(c)} \left( \sum_{i=1}^{n(c)} \frac{d}{t_i E} \right)$$  \hspace{1cm} (2.9)

Here, \(n(c)\) is the total number of charges that crossed the sample.

The conductivity is calculated using Eq. (2.10) with the assumption that each NT contributes carriers with the same concentration of \(9 \times 10^{10}\) cm\(^{-2}\), which has been calculated elsewhere for graphene sheets.\([122]\) However, only those NTs that have been visited by a carrier are considered to contribute to the carrier density of the sample.

$$\sigma = ne\mu_{avg}$$  \hspace{1cm} (2.10)

Here \(e\) is the elementary charge, \(\mu_{avg}\) is the average mobility from Eq. (2.9), and \(n\) is the carrier density, given by:

$$n = \frac{\sum_{i=1}^{N} 2\pi r_i l_i n_g}{V}$$  \hspace{1cm} (2.11)

In equation (2.11), \(N\) is the number of NTs participating in conduction, \(r\) and \(L\) are respectively the radius and length of NT \(i\), \(n_g\) is the carrier density of each individual NT (calculated from the experimental value for graphene), and \(V\) is the total volume of the sample.

The above formulation for charge transport has been used for many of the simulations presented in this work. A few of the studies on charge transport in later chapters incorporate some new characteristics into the charge transport routines. A maximum hopping rate of \(1 \times 10^4\) s\(^{-1}\) has been found to give a good agreement between the simulation and experimental results for NT composites within this study. This value is
much lower than the $10^{14}$ s$^{-1}$ hopping rate that has been calculated for conductive polymers.[105]

As the goal of this work was to form a predictive model for charge transport, the ideal model would not need to fit simulation data to experimental data by scaling up or down the pre-factor, $v_0$. It was also felt that within this version the electric field was not appropriately affecting charge transport. In addition, Eq. (2.2) is considered the hop rate between the junctions, and the combination of hop rates for neighbors is placed into a cumulative distribution function to determine the most likely hop to occur. Therefore, the last exponential in Eq. (2.2) can cause the hop rate to exponentially increase towards infinity if a charge were to travel a far enough distance through a NT in the transport simulations. This then would cause the time for a hop, as calculated in Eq. (2.7) to be zero, and not representative of the actual system. Therefore, it was in the best interest of this work to formulate a better method, presented in Chapter 4, for calculating the probabilities for various hops and then the time it would take a charge make those hops.
CHAPTER 3

EFFECTS OF MORPHOLOGY ON CONDUCTIVITY

One general limitation of most atomic and coarse grain simulation models is the size of the sample that can be treated. Thus, it is important to determine how size-related effects impact results and to find strategies to minimize those effects. In addition, there are a number of geometrical properties which are tailorable within the fabrication of these types of composite. These properties include the NT waviness, aspect ratio, and agglomeration. In this chapter these issues are studied with the results presented below.

3.1 Nanotube Waviness and Aspect Ratio

Mobility and conductivity values were obtained for four samples with different degrees of waviness. The sample dimensions were 100x50x50 nm$^3$ with 0.5 nm diameter NTs at a target length of 25 nm, leading to an aspect ratio of 1:50 (Figure 3.1).
Figure 3.1. Waviness Factor equal to a) 0.31, b) 0.5, c) 0.67, and d) 0.71

The average mobilities (Figure 3.2) and conductivities (Figure 3.3) for the various degrees of waviness in each simulation are shown below. Table 3.1 lists the fitting parameters for the power law equation.

Figure 3.2. Mobility vs NT concentration as a function of waviness. The two greyed data points are below the percolation threshold. However, they represent the cases in which very few charges reached the opposite electrode (0.49% for WF=0.31 0.46% for WF=0.5) so the statistics on these points is very poor.
Figure 3.3. Conductivity vs. NT concentration as a function of waviness for (a) WF=0.31, (b) WF=0.5, (c) WF=0.67, and (d) WF=0.76 (e) Power law fits to conductivity values for each WF.
Table 3.1. Fitting parameters for the power law for varying waviness in NT composites.

<table>
<thead>
<tr>
<th>Waviness Factor</th>
<th>$\phi_c$ (Volume Fraction)</th>
<th>$t$</th>
<th>$\sigma_c$ (S/cm)</th>
<th>$R$-Square</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.31</td>
<td>0.0068</td>
<td>3.419</td>
<td>48.60</td>
<td>0.9916</td>
</tr>
<tr>
<td>0.50</td>
<td>0.0093</td>
<td>2.817</td>
<td>11.90</td>
<td>0.9975</td>
</tr>
<tr>
<td>0.67</td>
<td>0.0097</td>
<td>2.963</td>
<td>19.70</td>
<td>0.9982</td>
</tr>
<tr>
<td>0.76</td>
<td>0.0148</td>
<td>2.243</td>
<td>1.70</td>
<td>0.9995</td>
</tr>
</tbody>
</table>

The power law, Eq. (1.1), is dominated by the large volume fraction points where tunneling is not a factor. Thus, it is natural that the curves in Figure 3.3a-d do not fit well to the simulation data near the onset of percolation where tunnelling is dominant. Overall, mobility and conductivity values increase with decreasing WF. This was expected as a higher WF means the NT spans a smaller distance along the direction of the electric field, forcing carriers to go through more NT-NT junctions to reach the collection electrode. In addition, the percolation threshold shifts to lower values for decreasing waviness. This trend has been seen elsewhere in the literature.[19], [123] More aligned NTs generally have been seen to increase the charge transport properties of these composite materials. These results indicate that the project was indeed capturing the physics of the system and that results were comparing well with what has been seen in experimental works.

In order to study the effects of NT aspect ratio, two simulations were performed on samples with dimensions of 100x50x50 $\text{nm}^3$. All NTs had a 0.5 nm radius with a maximum possible length of 25 nm and 37.5 nm in the first and second simulations, respectively. These correspond to aspect ratios of 1:50 and 1:75. Charges were injected into the sample under an applied electric field of 1 V/cm along the x-direction. The conductivity of each sample is given in Figure 3.4 with the associated percolation power law, given by Eq. (1.1), fit to the simulation result. The waviness factor is shown in
Figure 3.4 for each sample; notice that given the different aspect ratio, the waviness factor differs slightly (0.31 for a 1:50 aspect ratio and 0.37 for a 1:75 aspect ratio). The parameters for the associated power law fits are shown in Table 3.2.

Table 3.2. Parameters for the power law fits as a function of aspect ratio.

<table>
<thead>
<tr>
<th>Aspect Ratio</th>
<th>Waviness Factor</th>
<th>$\phi_c$ (% Volume)</th>
<th>$t$</th>
<th>$\sigma_c$ (S/cm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1:50</td>
<td>0.31</td>
<td>0.0063</td>
<td>3.415</td>
<td>48.60</td>
</tr>
<tr>
<td>1:75</td>
<td>0.37</td>
<td>0.0044</td>
<td>3.149</td>
<td>80.60</td>
</tr>
</tbody>
</table>

The conductivity is higher at all volume concentrations for higher aspect ratio while the percolation threshold is shifted to lower concentrations, which is in agreement with other authors [24], [25], [80], [99]. Values for the critical exponent fit into the experimental ranges, which vary between 1 and 6.[19] It should be noted that there is a slight variation in waviness in the two different aspect ratio cases in Figure 3.4. However, the average difference between the curves for samples with WF 0.31 and WF 0.5 in
Figure 3.3 is only 0.0003, while the average difference between the two AR curves in Figure 3.4 is 0.001. This illustrates that the effect of the aspect ratio is much more significant than the effect of waviness in charge transport in nanocomposites. It also shows that the difference in WF that was unavoidable in the calculations reported in Figure 3.4 is minimal and the observed behavior is indeed attributed to the aspect ratio.

### 3.2 Agglomeration

As previously discussed, many experimental studies have shown that CNT dispersion within a composite will play an important role in determining the electrical properties of the composite.[25], [29], [31]–[34], [56] A major benefit to growing NTs along a three-dimensional grid system, as discussed in Section 2.1, is the ability to easily grow agglomerates or bundles of NTs. The grid system allows NTs to be grown directly adjacent to previously grown NTs without greatly increasing simulation times because the grid points predefine the positions available for growth. Figure 3.5 illustrates a large agglomerate in a sample demonstrating the ability of the routine to generate NTs into agglomerates. The virtual electrodes are on opposite sides of the sample in the x-direction.
The creation of samples with different levels of agglomeration was achieved by three different methods. In each method, a *uniformity index* (UI-x, with x the method number) is defined such that a larger value means a more uniform sample. In the first method, a given number of seed NTs are uniformly placed in the sample and the remaining NTs are forced to grow around those seeds. For this method the uniformity index (UI-1) is the number of seeds. This method will tend to produce samples with the same number of agglomerates at all concentrations, while the size of the agglomerates grows with concentration. The second method also starts by uniformly dispersing seed NTs, but unlike method 1 the number of seeds is not fixed, but determined as a percentage of the total number of NTs to be placed. The remaining NTs are then forced to agglomerate around those seeds. UI-2 is the percentage of NTs that are seeded. With this method, the number of agglomeration sites increases proportionally with the concentration, but the size of the agglomerates will tend to be constant.
In the last method, there are no seed NTs but each new NT is given a probability (UI-3) to be placed randomly within the sample (as opposed to being placed near an existing NT). Samples generated with this method are expected to have a less uniform distribution, both in number and size of agglomerates, than the previous two methods. As the probability increases, the sample becomes more uniformly distributed, so 0% means fully agglomerated and 100% is fully uniform.

In this discussion on agglomeration, this work will refer to the number of seeds in the first method, the percentage of volume as seed NTs in the second, and the probability of growing in a random position in the third, as the uniformity parameter for the method. With this choice, in all methods, a smaller value (number of seeds, or percentage) means less (and larger) agglomerates, while a large number means more uniform samples.

Figure 3.6 shows examples of how agglomerates grow with concentration with each of the methods. Each row in the figure shows one of the methods, while the columns correspond to concentrations of 0.01, 0.02, and 0.03 respectively. As can be seen, when using the first method (Figure 3.6a-c), as concentration increases, agglomerate size grows and percolation is achieved when agglomerates are large enough to connect to each other. Figure 3.6d-f show that with method 2 the number of agglomerates increases with concentration and percolation is achieved when there are enough to connect the two electrodes. Finally, Figure 3.6g-i show a more disorganized sample. This is more obvious at a lower concentration (Figure 3.6g).
Figure 3.6. Method 1 samples with 12 initial seed NTs are shown with periodic representation at concentrations of d) 0.01, e) 0.02, and f) 0.03. Method 2 samples with 5\% of the required volume of NTs being dispersed randomly in volume concentrations of g) 0.01, h) 0.02, i) 0.03. Method 3 samples with each new NT having a 5\% probability for uniform dispersion are shown with periodic representation at concentrations of a) 0.01, b) 0.02, and c) 0.03. In these plots a periodic image above and below the central volume is shown.
Samples with different degrees of uniformity were created with the three methods. The box dimensions in all cases were 50x50x50 nm$^3$ with 0.5 nm radius NTs that had a maximum length of 15nm, or a 1:30 AR. For each configuration that was simulated, 64 independent copies of the simulation volume were generated. All results reported below correspond to an average of all these configurations.

Using method 1, samples with 9 different UI-1, namely 0, 1, 2, 4, 8, 12, 16, 20, and 100, in addition to a uniformly distributed sample (MS), were created (group 1). For simplicity in the notation, samples with uniform distribution, in which the number of seeds is equal to the total number of NTs, and thus different for all concentrations, have been nicknamed MS for maximum number of seeds. Samples created in this way are equivalent to the 100% case in both of the other two methods. For method 2, simulations were performed in which 1, 2.5, 5, 10, 25, 50, 75, and 100% of the required NT volume fractions were randomly dispersed (group 2). Nine different degrees of agglomeration were studied with method 3 (group 3). In these cases, each new NT had a 0%, 1%, 2.5%, 5%, 10%, 20%, 50%, 75%, or 100% chance of being placed randomly, as opposed to near another existing NT. In the 0% case, only the first NT was placed randomly to initiate the generation process.

The normalized fraction of charges that reached the collection electrode in groups 1, 2, and 3, and the mobility of those charges, are shown in Figures 3.7, 3.8, and 3.9, respectively. The histograms were normalized such that the area under each curve is 1 in all cases. This was necessary because the fraction of charges reaching the electrodes is significantly different at different concentrations and comparison becomes difficult.
Figure 3.7. Normalized values for the charges reaching the collection electrode and mobility of those charges for concentrations of a) 0.01, b) 0.02, c) 0.03, and d) 0.04 when using method 1.
Figure 3.8. Normalized values for the charges reaching the collection electrode and mobility of those charges for concentrations of a) 0.01, b) 0.02, c) 0.03, and d) 0.04 when using method 2.

Figure 3.9. Normalized values for the charges reaching the collection electrode and mobility of those charges for concentrations of a) 0.01, b) 0.02, c) 0.03, and d) 0.04 when using method 3.
The trend is consistent in all three methods. At the two lowest concentration values, right below or on the onset of percolation (0.01 and 0.02), a larger number of carriers reaching the electrodes, and higher mobilities, are found for samples with intermediate Ul values. Neither the disperse samples nor the heavily agglomerated samples allow for many carriers to reach the collection electrode. For the lowest Ul in methods 1 and 3, there is just a single clump in the box that is unable to bridge the gap between the two electrodes. For method 2, Ul-2=1% corresponds to one or two agglomerates, as there are between 100 and 200 NTs in that sample at these concentrations, and it is possible that some transport occurs in some of the replicas. However, the number of carriers reaching the electrode is still smaller than at larger Ul values. When fully dispersed, the NT-NT distance is close to the average distance between neighbors for the sample, which at these low concentrations, is quite large even for tunneling. Thus, carriers time out and are discarded, meaning the mobility is below an acceptable threshold. Instead, when some agglomeration is allowed, the volume is more efficiently utilized. By agglomerating, the NT-NT distance within the agglomerate is low enough for conduction to occur and conductivity arises when these agglomerates are able to connect to each other. The sample consists of voids and areas of higher density where conductivity occurs.

For the larger concentrations, top of the percolation threshold (0.03 and 0.04), the number of carriers reaching the collection electrode increases with the Ul value, but the mobility shows the opposite behavior. The concentration is large enough that there are conduction paths formed regardless of the Ul. In addition, a larger Ul value leads to a
larger number of parallel paths available for the carriers leading to the observed increase in carrier collection. On the other hand, the larger the UI, the further apart NTs are from each other in each path, leading to a lower mobility. So at larger concentrations (top of the percolation threshold), with larger UIs, more carriers are collected but with lower mobility, leading to a tradeoff to be considered in the optimization of the composite’s electrical performance.

Conductivity depends on the mobility, but also on the carrier density, which is proportional to the number of NTs that participate in conduction. The carrier density is not the same as the number of carriers reaching the collection electrode. The number of collected carriers does depend on the number of conduction paths available (which is related to the number of participating NTs), but also on the characteristics of those paths. Conductivity values from samples in each group are shown in Figure 3.10a), b), and c), respectively.

![Figure 3.10](image.png)

Figure 3.10. Conductivity vs a) UI-1, b) UI-2, and c) UI-3 for four volume concentrations.

In all the cases and at all concentrations, the lowest conductivity is found for the fully dispersed sample. The only exceptions are for UI-1=1 and UI-3=0 at the largest concentration of 0.04. However, in those two cases, there is roughly just one large
agglomerate that in many cases cannot connect the electrodes. Conductivity becomes more independent from the degree of agglomeration as concentration increases, which is expected as percolation pathways become fully established and conductivity is less sensitive to concentration. For Ul-3, where agglomerates are random in size and number, there is a more random behavior, particularly at low concentrations.

It can be inferred from Figure 3.10 that percolation begins to occur at lower concentrations when there is agglomeration than for uniformly distributed samples. None of the samples, regardless of the method used to create it, shows any conductivity in the fully dispersed case at concentrations of 0.01. Thus, 0.01 is below percolation for uniform samples. However, all three methods show some conductivity for intermediate values of the corresponding Ul (even if for just one case as in method 3), showing that the percolation threshold has indeed shifted down. The lower percolation threshold is caused by the inserts forming localized, interconnected bundles that connect to both electrodes so that charge tunneling is minimized in the agglomerates. This allows for improved charge transport at lower volume fractions. A numerical study on percolation of spheres and rods by Jadrich et al. has also shown this tendency towards a lower percolation threshold with increased attraction between filler particles [124].

To further analyze conductivity, notice that for methods 1 and 3, smaller Ul values mean less seed NTs and thus, for a given concentration, agglomerates are larger for lower Ul leading to less interconnects among agglomerates. In addition, as the relative size of the agglomerate compared to the “hairs” increases, the conductivity between interconnects becomes significantly lower than in the agglomerate and these connections act as bottlenecks decreasing conductivity. This is evident in Figure 3.10a
Figure 3.10b where conductivity is smaller for samples with large agglomeration
than for those with intermediate levels of agglomeration. This contrasts with the results
on method 2 where the agglomerate size is expected to be more uniform while the
number of agglomerates grows towards percolation. This leads to a more uniform
behavior of the conductivity vs UI which monotonously decreases with UI at all the
concentration values studied here (Figure 3.10b). At the larger volume fractions, all of
the samples have reached percolation and it can be seen from Figure 3.10c and d, that
conductivity tends to be uniform across UIs (except maybe for extreme values).

For a perfectly dispersed sample, the concentration has to increase to a point
where the average distance between NTs allows for conductivity. However, if small
(nano-sized) agglomerates form, the NT-NT distance in the cluster is what controls
conductivity as long as agglomerates connect to each other to bridge the electrodes.
Agglomeration leaves voids in the sample in favor of localized bundles that connect to
each to each other through NTs at the edges that stick out like hairs.

As a visual aid to further understand the effect that these nano-agglomerates have
on charge transport, a sample with 8 initial seeds at a volume fraction of 0.01 is shown in
Figure 3.11. The NTs clearly form bundles that are interconnected. This figure was
produced with NTViz,[114] which is a GUI that also allows the user to highlight the
number of times that each NT was visited by charges in the sample (Figure 3.11b). The
frequency of visits increases from blue to red. It is clear from Figure 3.11b and c that the
more heavily trafficked NTs are inside and in between the bundles. Notice also that the
group of NTs in the top-left of Figure 3.11a, are not connected to the rest of the NTs and
therefore, are not visited by any charges.
Figure 3.11. a) A sample at a NT concentration of 0.02 and initialized with 12 seed NTs. b) Each NT highlighted with a color that represents the number of visits each NT received. c) Zoomed in view of the highly visited NTs within an agglomerate.

3.3 Sample Size

The previous two sections have looked at how characteristics of the NTs in the composite affect conductivity within samples. However, any computational model of real materials is limited in scope to the size of the virtual sample that can be modeled. This section then focuses on changing the sample sizes in order to better understand how it will impact simulation results.
Simulations were run on samples with lengths ranging from 10 nm to 400 nm, with the dimensions perpendicular to charge transport kept constant at 100 nm. All samples had relatively the same waviness factor and the maximum NT length was always one half of the box's length. This last constraint was to ensure that the NT-length-to-box-side ratio was constant. Figure 3.12 shows a visual representation of the samples while Figure 3.13 shows the power law fittings to simulation results for conductivity. Table 3.3 lists the fitting parameters for those fits.

Figure 3.12. Visuals for samples of different lengths.
Figure 3.13. Power law fit to conductivity for different box lengths

Table 3.3. Power Law fitting parameters for varying sample lengths.

<table>
<thead>
<tr>
<th>Sample Length (nm)</th>
<th>Aspect Ratio</th>
<th>WF</th>
<th>$\phi_c$ (Volume %)</th>
<th>$\sigma_c$ (S/cm)</th>
<th>t</th>
<th>R-Square</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>1:10</td>
<td>0.72</td>
<td>0.0059</td>
<td>0.314</td>
<td>2.762</td>
<td>0.9966</td>
</tr>
<tr>
<td>40</td>
<td>1:40</td>
<td>0.75</td>
<td>0.0053</td>
<td>26.45</td>
<td>3.551</td>
<td>0.9441</td>
</tr>
<tr>
<td>80</td>
<td>1:80</td>
<td>0.75</td>
<td>0.0049</td>
<td>14.98</td>
<td>2.587</td>
<td>0.9940</td>
</tr>
<tr>
<td>100</td>
<td>1:100</td>
<td>0.76</td>
<td>0.0049</td>
<td>13.79</td>
<td>2.415</td>
<td>0.9978</td>
</tr>
<tr>
<td>200</td>
<td>1:200</td>
<td>0.73</td>
<td>0.0027</td>
<td>30.60</td>
<td>2.145</td>
<td>0.9201</td>
</tr>
<tr>
<td>400</td>
<td>1:400</td>
<td>0.73</td>
<td>0.0020</td>
<td>49.47</td>
<td>2.041</td>
<td>0.9926</td>
</tr>
</tbody>
</table>

What is seen through this study is that the overall sample size and NT aspect ratio does have an important effect on the transport properties within samples. For larger samples and longer NTs (larger aspect ratios), the critical concentration is lower and the conductivity is predicted to be higher. This is in agreement with the well-documented fact that the extreme aspect ratios of CNTs allow them to be quality conductive nanofillers. It then follows that to better mimic this capability through modeling, NTs with high aspect
ratios and samples large enough to contain them are needed. As far as can be known, relatively few, if any, modeling techniques can model such high aspect ratio nanofillers. In the simulations described here, in the larger sample, NTs are 200 nm long and 0.5 nm wide, leading to an aspect ratio of 1:400. Certainly there is a cost due to the large number of NTs in the simulation box, thus requiring a very efficient model. However, it is evident from this set of simulations that the results are size dependent and thus these samples are not equivalent.

Further simulations were run lengthening the sample size while keeping the NT aspect ratio constant. Three simulations were performed with the NT aspect ratio set to 1:10 in the first, 1:20 in the second, and 1:400 in the third. In the first simulation, the NT length was set to 5 nm with a radius of 0.5 nm and the box length was set at 75 nm in the y and z directions with PBC. The length in the x direction was varied from 10 nm to 50 nm in 10 nm increments. The various samples are shown below in Figure 3.14 at a concentration of 0.005 volume fraction. In the 10 nm box, approximately 95 NTs were needed at the 0.005 volume concentrations, while around 1300 were needed for the 50 nm box.
Figure 3.14. Samples with 5 nm length tubes at 0.005 volume fraction with lengths of a) 10 nm, b) 20 nm, c) 30 nm, d) 40 nm, and e) 50 nm.

Figure 3.15 shows the conductivity results, with error bars. As the box length increases, the conductivity does indeed decrease. However, the effect of size on conductivity seems to decrease with the box’s length, indicating that these results are beginning to converge to a specific value.
Figure 3.15. Simulation results for changing the maximum box length while keeping the NT size constant.

In the second simulation with NTs of 1:20 aspect ratio, the NT length was set to 10 nm and the sample length was varied from 20 to 60 nm again in 10 nm increments (Figure 3.16). In the 20 nm length box, 277 NTs were needed to reach the 0.015 volume fraction, while 831 NTs were needed in the 60 nm length box.
Figure 3.16. Various samples at a volume concentration of 0.015 for samples with maximum NT lengths of 10nm and sample lengths of a) 20 nm, b) 30 nm, c) 40 nm, d) 50 nm, e) 60 nm.

Figure 3.17 shows the conductivity results from the simulations. The convergence is more obvious here, actually for the two largest sizes (50 nm and 60 nm), except for the very low volume fraction right at the percolation threshold, the two curves are statistically indistinguishable. The conductivity at the lowest concentration is the average of just a very few charges as most of the charges do not make it to the electrode. It can be concluded that for 1:20 aspect ratio NTs, above 50 nm (~5-6 times the NT length), size effects are probably not as important.
Figure 3.17. Conductivity results for changing box length with NTs having constant length of 10 nm.

The 1:400 aspect ratio simulation was performed with much larger samples. The NT length was set at 200 nm and the y and z dimensions were fixed at 100 nm with PBC in effect (Figure 3.18). Figure 3.19 shows the simulation results. The longest sample length was 1 μm in length, and was not run at all volume fractions due to memory requirements for such large samples. In the 250 nm length sample, around 320 NTs were needed to reach the volume concentration of 0.005, while around 1300 were needed for the 1 μm length sample.
Figure 3.18. Samples with lengths of a) 250 nm, b) 300 nm, c) 400 nm, d) 600 nm, e) 800 nm, f) 1000 nm.
These results indicate that a statistically diverse sample is essential to obtain results that are relatively unaffected by simply increasing sample size. As mentioned previously, other models have resorted to PBC along the transport direction to achieve the same goal. However, topological diversity cannot be ensured in that way. The main issue with PBC is that if a charge reaches one end of the sample, it must continue into the periodic box from the point it left the previous box, in many cases confining the charge to specific pathways. This does not guarantee a quality sampling of various topologies that carriers would explore in a real sample. Rather, it only guarantees that each inserted charge would see many identical replicas of the same topology.

3.4 Comparison to Experimental Data with Preexisting Routines

The previous sections within this chapter have focused on simulations using the preexisting transport routines. It was therefore necessary to validate simulation results...
with experimental data. In the simulations in this chapter, the value for the tunneling constant, $a$, in Eq. (2.3) was chosen to be 0.6 Å$^{-1}$, which has been calculated for Poly Acrylonitrile (PAN).[91] As such it was of interest to compare simulation results with experimental values previously presented in the literature for PAN/SWCNT composites. Results from simulations on box sizes of 800x100x100 nm$^3$ and 1000x100x100 nm$^3$ with an applied field of 1 V/cm, were fit to the power law for conductivity, Eq. (1.1). Those power law fits were then compared to experimental data by scaling $v_0$ in Eq. (2.3) by a factor of 0.8 (Figure 3.20) essentially lowering the conductivity values from simulations by 20%. These longer samples, with NT aspect ratios of 1:400, were chosen for comparison because they were statistically indistinguishable from each other while also being some of the largest samples with maximum NT lengths at the time of the comparison.

![Power Law Curves](image)

Figure 3.20. Power law curves from simulation data fit to experimental data from works by Guo et al.[125] and Pradhan et al.[120]
The experimental values used for comparison were at much larger volume concentrations than what are feasible with this model. As such, the power law fits had to be extrapolated out to the experimental results, yet still agree well with the experimental results. Most experimental works with PAN and CNTs focus on the formation of fibrous bundles of PAN/CNTs as in the work by Almuhamed et al. [126], and as such are not necessarily comparable to the simulation results presented within this work. Further work to characterize the tunneling constant, $\alpha$, for different materials has been performed and can be applied to this model for studies on other materials.
CHAPTER 4

CHARACTERIZING THE HOPPING ALGORITHM

4.1 Determining Tunneling Parameters

The host matrix material will have an effect on charge transport via the quantum phenomenon of charge tunneling. Previous work has been done to quantify charge tunneling between graphene sheets through a polymer layer, using Density Functional Theory (DFT) combined with Green’s functions (GF), where a 4-fold decrease in the tunneling constant, compared to vacuum, was observed for Poly Acrylonitrile (PAN).[91] This approach can be employed to calculate tunneling parameters for any material and it is used to calculate the tunneling constant for CP2 polyimide. Unlike reference [91], the non-equilibrium version of Green’s functions (NEGF) are used within the Atomistix Toolkit software package. In addition, semi-infinite (5,5) CNTs were included in the simulation box, as opposed to bent graphene layers, as in the aforementioned study.

The DFT functional employed for the calculations was the widely used generalized gradient approximation of Perdew, Burke and Ernzerhof (PBE) exchange-correlation functional.[127] The DFT-D2 functional by Grimme was employed in order to account for dispersion forces.[128] With this setup, calculations were performed to obtain transmission coefficients for carriers within the device for various voltage differentials. From the transmission spectra, the current through the device can be found using Eq. (3.1) to obtain a current-voltage (IV) relationship.
\[ I = \frac{e}{\hbar} \sum \int T(E) \left[ f \left( \frac{E - \mu_R}{k_B T_R} \right) - f \left( \frac{E - \mu_L}{k_B T_L} \right) \right] d(E) \] (4.1)

Here the summation is over each spin channel, \( e \) is the electron charge, \( T(E) \) is the transmission coefficient per spin channel, \( f \) is the Fermi function of either electrode, \( T_L \) and \( T_R \) are the temperatures of the left and right electrodes, and \( \mu_L \) and \( \mu_R \) are the chemical potentials of the left and right electrode, which are related to the applied voltage bias as in Eq. (4.2).

\[ \mu_R - \mu_L = eV_{bias} \] (4.2)

Previously, it has been seen that, for a constant \( V_{bias} \), the carrier transmission rate exponentially decays with distance through a non-conducting region.\[91\] As such, the current-distance relationship can be approximated by Eq. (4.3).

\[ I = A e^{-2\alpha d} \] (4.3)

Here \( A \) is a constant determining the maximum amplitude of the current, \( \alpha \) is the tunneling parameter, and \( d \) is the distance through the insulating region. The exponential term in Eq. (4.3) can be found directly by using the Wentzel-Kramers-Brillouin (WKB) method for approximating the transmission probability for a particle encountering an energy barrier. In a simple system where the particle and barrier energy do not change as a function of \( x \), \( \alpha \) is a measure of the tunneling barrier height, \( V \) relative to the energy, \( E \), of the particle. In the limit of a high and wide energy barrier, \( V \) is much larger than \( E \) and the transmission probability exponentially depends on the barrier width.\[121\] Thus, for wide and tall barrier, the current versus tunneling distance can been approximated by Eq. (4.3).

To characterize \( \alpha \) for other materials, the current voltage relationships for various devices were fit to Eq. (4.3) using \( \alpha \) and \( A \) as the fitting parameters. If the process is
purely tunneling, then the natural logarithm of the current has a linear relationship with the distance between the NTs as in Eq. (4.4). Therefore, a simple least squares fit algorithm can be used to fit the current-distance relationship. In this work, the Matlab Curve Fitting Toolbox was used as the fitting tool.

\[
\log(I) = \log(A) + 2ad
\]  

(4.4)

Within this work, most of the simulations reported assumed the matrix material to be Polyacrylonitril (PAN) since the tunneling constant for this material and experimental conductivity values were readily available for PAN/CNT composites. However, in the interest of further validating this model and to illustrate its capability, the tunneling constant of other polymeric materials was needed. A study performed in 2003 by Ounaies et al. was selected as a possible experimental work to compare with simulations. In that work, the matrix material was CP2 Polyimide with single walled CNT inclusions.[88]

To begin the process to characterize \( \alpha \) for this system, a single CP2 Polyimide monomer (Figure 4.1) was built and its geometry optimized using DFT calculations in Gaussian 09'. The B3LYP functional with the 6-31G basis set was used. The optimized structure was then ported into the Atomistix Toolkit software package.

Figure 4.1. Optimized CP2 molecule within Gaussian 09'.
This monomer was then placed in between two metallic CNTs in various configurations in order to sample various arrangements of this polymer/CNT interface. The first configuration contained a single CP2 monomer placed in between the two CNTs as in Figure 4.2.

![Figure 4.2. Single monomer of CP2 Polyimide between two metallic CNTs.](image)

The calculation was repeated with two and then with three monomers within the junction. In order to do this, the CP2 molecules were rotated before being placed within the device. These first configurations are thus termed the "rotated" devices. Within each new device, the central region was optimized, with the CNTs being held fixed to speed up simulation times and provide structure to the device. For these devices, shown in Figure 4.3, the initial configurations were such that the distance between CNTs was 1 nm for the single monomer, 2 nm for the 2 monomers, and 3 nm for the 3 monomers.
Two other devices were also created, one with a single monomer placed vertically so that the length of the monomer spanned a larger distance, and the second with two monomers placed end to end within the junction between NTs. These two configurations are deemed the "bridge" configurations, and are shown in Figure 4.4. The distance between the two CNTs was 3 nm for the single monomer and 5.8 nm for the two monomers.
Figure 4.4. a) 1 and b) 2 monomers stretched end to end bridging the gap between the two metallic CNTs.

The current vs. tunneling distance for both the rotated and bridge configurations are shown in Figure 4.5 with the current shown in log-scale. There is a clear exponential decay with both the rotated and bridged configurations as the junction distance between the two CNTs increases, as expected. It can also be seen that the two configurations also lead to similar decay rates, which was also expected since the same molecule is bridging the junction gap, just in different orientations.

Figure 4.5. The current-distance relationship for the rotated configurations and the bridged configurations at the various distances.
The tunneling constants for the rotated configuration and the bridged configurations are shown for varying voltages in Table 4.1. The current-distance relationships for voltages of 0.3 and 0.5 were not plotted in Figure 4.5 to avoid convolution within the figure.

Table 4.1. Tunneling constants, $\alpha$, for various voltages in the rotated and bridged configurations.

<table>
<thead>
<tr>
<th>Bias (V)</th>
<th>$\alpha$ (1/Å) - Rotated</th>
<th>$\alpha$ (1/Å) - Bridge</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.1</td>
<td>0.4934</td>
<td>0.5412</td>
</tr>
<tr>
<td>0.3</td>
<td>0.4879</td>
<td>0.5230</td>
</tr>
<tr>
<td>0.5</td>
<td>0.4831</td>
<td>0.5044</td>
</tr>
<tr>
<td>0.7</td>
<td>0.5004</td>
<td>0.5308</td>
</tr>
</tbody>
</table>

It can be seen that, overall, the change in $\alpha$ with voltage overall is minimal. Therefore, the average $\alpha$ for all biases in all configurations was obtained so that an average tunneling constant could be used for future studies within transport routines. For CP2 Polyimide, the tunneling constant to be used in transport routines would be 0.5068. This same process can be used to characterize the tunneling constants for other materials.

4.2 Charge Transport through Junctions.

Intuitively, the maximum hopping rate, or transition rate, in a junction between NTs is essentially the number of times per second that a charge would attempt to move to the neighboring conductive site. The hopping rate in this model will be at the maximum, or equal to $v_0$, when two conductive sites are such that the tunneling distance for a charge is considered to be zero, and $\Delta G^*$ in Eq. (2.2) is equal to zero. Since the formulation of time in hopping models involves taking the inverse of the hopping rate, it intuitively makes sense that one could determine the time it would take for a charge to move through
the contact resistance between two CNTs and then take the inverse of that time to find the transit rate. Ohm's Law, which is given in Eq. (4.5), can be used as a classical representation of the time $dt$ it would take and amount of charge $dq$ to move through a contact resistance, $R$, between two CNTs.

$$I = \frac{dq}{dt} = \frac{V}{R}$$  \hspace{1cm} (4.5)

Equation (4.5) can be rearranged so that $dt$ can be represented by Eq. (4.6).

$$dt = \frac{eR}{Ed} = \nu_0^{-1}$$ \hspace{1cm} (4.6)

Where the voltage, $V$, can be written in terms of the local electric field across the resistor, $dq$ in this case can be represented as the value of one charge, and $d$ is the distance across the resistor. Therefore, the maximum hoprate, $\nu_0$, will be equal to $dt^{-1}$ when two NTs are in contact with each other, and proportional to the electric field between the two CNTs. This is significantly different from the original version of the transport routine as the electric field has an increased effect on charge movement by altering the maximum transfer rate.

The estimation of the value of the contact resistance between two CNTs varies within the literature between $10^3$ and $10^8 \Omega$.\cite{129} In order to obtain a specific value of the contact resistance between two CNTs, ab initio calculations were performed on the device shown in Figure 4.6 in Atomistix Toolkit using DFT. The GGA PBE functional with the Grimme DFT-2 dispersion correction was used to obtain the current-voltage relationship, as previously discussed in Section 4.1. Both CNTs in the above figure are armchair CNTs, meaning that the chiral vector $(m,n)$ was such that $m$ and $n$ are equal, in this case $(5,5)$. Since $m$ and $n$ are equal, the CNTs were expected to behave as metallic objects in the device.\cite{2} The distance between two graphitic carbon sheets has been
found to be around 3.414 Å.[130] However, within polymeric materials, the CNTs may be forced to be closer or may actually be farther apart than this distance. A full optimization of the CNTs could be performed in order to find this distance. However, as an approximation, the two CNTs were placed in the sample so that the minimum distance between them was around 3 Å.

Figure 4.6. Two armchair CNTs distanced approximately 3 Å apart to calculate the contact resistance between two metallic CNTs.

The current-voltage relationship for the device, shown in Figure 4.7, clearly indicates Ohmic behavior, which was to be expected for the two metallic CNTs separated by a junction resistance. A linear fit of the I-V relationship led to a value of 1.754x10^6 Ω within the device, which is well within reported experimental values of the contact resistance for CNTs.[129] With this value for the resistivity, the transit time for one charge moving through the above resistor, using Eq. (4.6), would be 9.366x10^-6 s for an electric field of 1 V/cm within the junction. Using this transit time, the number of times per second that one charge would move through this resistor would be 1.067x10^5 s^{-1}. 
Figure 4.7. Current versus voltage for the device with two armchair CNTs.

As a side note, this new method for parameterizing $v_0$ leads to a maximum transfer rate that is proportional to the voltage in the junction between two NTs. In the original version of the transport routine, $v_0$ was a constant. Within these simulations, if the voltage between two CNTs is constant, then $v_0$ will remain a constant. However, if the local voltages across the junction resistance vary, then $v_0$ will also vary. More work to validate this method for calculating $v_0$, and characterizing the voltage in these types of contact junctions will be discussed in the future works section of this work. However, it will be shown later in Chapter 5 that considering the same electric field in all junctions does lead to acceptable results as long as the value of that field is equal the average of field among all the junctions, which happens to be a few times larger than the applied field.
4.3 Charge Transport through CNTs.

Historically, hopping algorithms being used for charge transport in semiconducting materials and conductive polymers dealt with charges that are localized to sites. Previous hopping studies generally have not needed to take into account delocalized carriers, or carriers that are allowed to move around in localized confined region.[103], [104] In the hopping algorithm presented so far in this work, the carrier is given the ability to move along the conductive NT. In the version of this hopping algorithm presented in Section 2.4, each CNT was considered to promote ballistic carrier transport, and therefore it was approximated that the time it would take a carrier to move through a CNT would be negligible when compared to the time of the hop between NTs. In this way, the hopping algorithm would only be concerned with charge hopping between the NTs, much like other hopping algorithms previously discussed. In reality, the CNT will have an intrinsic resistance and it will take the carrier some amount of time to move through the CNT over a specified distance. This time can be approximated using the classical equation for Ohms law, as described below.

It is a known phenomenon that any ballistic transport process will be limited by the quantum of conductance, $G_0$, given in Eq. (4.7).[7], [131], [132]

$$G_0 = \frac{2e^2}{h} \quad (4.7)$$

Here, $e$ is the value of the electron charge and $h$ is Planck's constant. The factor of 2 accounts for the degeneracy of the electron spin. This quantum of conductance is the maximum conductance, or minimum resistance, that can be obtained in a 1-D channel separating two ideal contacts at very low temperatures. It arises in systems where the distance to be traveled by the charge is much smaller than the mean free path of the
charge, and therefore when there is no scattering within the transport channel.

Essentially, there is a maximum current carrying capacity for the 1-D channel for a given voltage bias, which leads to a maximum conductance. For a detailed derivation of Eq. (4.7), the reader is referred Charles Kittel’s *Introduction to Solid State Physics.*[133]

To determine the maximum conductance of a NT, a current voltage calculation was performed in Atomistix Toolkit on a single armchair CNT with a diameter of 0.617 nm spanning the two electrode regions of the device, shown in Figure 4.8.

Figure 4.8. A single (5, 5) armchair CNT bridging two electrodes to measure the resistance of a CNT.

The current versus voltage is plotted in Figure 4.9. A linear fit to the current voltage relationship from ATK gives a resistance value of 6.454kΩ through the device which is equivalent to a conductance $G = R^{-1}$ that would be equal to $1.55 \times 10^{-4} \text{ S}$. Since $G_0$ in Eq. (4.7) equals to $7.75 \times 10^{-5} \text{ S}$, this (5, 5) NT has a conductance of $2G_0$, which is within the experimental conductance of CNTs.[132] This value should henceforth be the
limiting factor in how quickly a charge could move through the NTs within these simulations.

![Current-Voltage relationship](image)

Figure 4.9. Current-Voltage relationship for a single CNT bridging two electrodes.

Ohm's law is again used to find the transit time for charge movement, but this time through a CNT, according to Eq. (4.8).

\[ J = \frac{E}{\rho} = \frac{E}{RA} \]  \hspace{1cm} (4.8)

\( J \) is the current density, \( \rho \) is the electrical resistivity, \( E \) is the electric field that is influencing the charge, \( R \) is the resistance of the conductor, and \( A \) is the cross sectional area of the conductor. For a metal, the current density can be represented by Eq. (4.9).

\[ J = \frac{I}{A} = \frac{n Av q}{A} = n v q = \rho v \]  \hspace{1cm} (4.9)

Here \( A \) is the cross sectional area the current is moving through, \( n \) is the number density of free charged particles within the metal, \( q \) is the value of the charge on each of those particles, and \( v \) is the drift velocity of the charged particles. \( nq \) is therefore equal to the
charge density, $q$ within the conductor. This generalized version of Ohm’s law can be rewritten as in Eq. (4.10) in order to find the drift velocity of charges moving through the conductor.

$$v = \frac{E l}{R A q} \quad (4.10)$$

In the transport routines with NTs, it is assumed that $E$ is the applied electric field along the x direction, $l$ is the length of the NT, $A$ is the cross sectional area of the NT, and $q$ is the charge density within NTs. $q$ is the carrier density for the individual CNT, calculated using the carrier density for graphene, $n_g$, taken from Eq. (2.11). From the drift velocity the time $t$ it takes a charge to move through the conductor can be found through Eq. (4.11).

$$t = \frac{\Delta x}{v} \quad (4.11)$$

Here, $\Delta x$ is the distance the charge moves along the x-direction under the influence of the applied electric field, which in this case is the distance between the point where the charge enters the NT and the point where the charge exists the NT.

4.4 Improvements in the Transport Algorithm

In order to incorporate the calculations presented in Sections 4.2 and 4.3, the hopping algorithm was altered as described below. The probability for hops to occur is still governed by Eq. (2.2). However, Eq. (2.2) is deemed a pseudo probability as the exponential calculating the effect of the electric field can allow this value to exponentially increase. The individual values between each neighboring NT are compiled into a normalized cumulative distribution function. The next NT to hop to is then chosen
according to this cumulative distribution function and a sorted random number. Thus far, this is no different than the original version of the transport algorithms.

In that version of the transport algorithm, the hopping rate, Eq. (2.2) was the term used for the calculation of the transit time. However, because the hopping rate exponentially increases with NT length or with the electric field, due to the last exponential, the transit time can decrease exponentially. This means that the maximum drift velocity for charges will have an infinite limit, and could be larger than the maximum conductance of the conductive filler, as calculated in Section 4.3, would predict. Therefore, it was of interest to separate the calculation of time into two parts. The first is the time for hops to occur, and second is the time it takes for a charge to move through a NT. Therefore, the actual hopping rate used for calculating the time for a hop is similar to Eq. (2.2), except that the last exponential term is removed, leading to Eq. (4.12). In this way the time calculated from the hopping rate accounts for the time involved in the hop from NT to NT and not the transit time in the NT.

\[
\omega_{ij_{time}} = \nu_0 e^{-2\alpha R_{ij}} e^{\frac{\Delta G + \alpha G}{2k_B T}}
\]  

The time, \( t \), for a charge moving through the CNT can then be calculated using Eq. (4.11). Therefore, the total time, \( \tau \), for the charge to move down the CNT and then to transit from one NT to a neighboring NT can be written as in Eq. (4.13).

\[
\tau = \frac{\ln(\zeta)}{\sum \omega_{ij_{time}}} + t
\]  

The first portion of Eq. (4.13) is taken from Eq. (2.7), but using Eq. (4.12) instead of Eq. (2.2). The additional time \( t \) then incorporates the maximum conductance allowed through an ideal CNT.
Still one major problem remains that needs to be solved, which is how to deal with the statistically possible case of a carrier moving in the CNT against the electric field. This is possible due to the nature of the Metropolis algorithm that allows an energetically unfavorable move, with Boltzmann probability. Such a move in the direction of the field has been predicted in conductive polymers with geometrical disorder, as traps in the sample may make going around (and thus against the electric field in some regions) more favorable. The problem is that if the current algorithm were used, the time $t$ calculated using Eqs. (4.9) to (4.11) would be negative. So the task then becomes finding a suitable means of calculating this time for carriers moving against the field.

If there is no applied electric field, the carrier would be equally likely to move in either direction along the NT. Upon applying an electric field, carriers will on average move in the direction of the field. The term governing the change in energy due to a carrier moving down the NT under an electric field is given by Eq. (4.14).

$$e^{\frac{qEad}{k_BT}}$$

This factor will lead to an increased hopping rate for carriers moving down the field and a decreased hopping rate for carriers moving against the field. Since the transit time is obtained as the inverse of the hopping rate a ratio between the factor in Eq. (4.14) for down the field transport to the same factor for up the field transport through the same distance should be proportional to the ratio between the up-the-field transit time to the down-the-field transit time. In other words,
since in this case $d_B$ is equal to $-d_F$, which is equal to $2d$, where $d$ is the distance travelled by the carrier. Now, the transit time for carriers moving against the electric field can be calculated as in Eq. (4.16).

$$t_B = P_{FB} \ast t_F$$  \hspace{1cm} (4.16)

Here, $t_F$ is the time $t$ obtained from Eq. (4.16) and $P_{FB}$ is the ratio given by Eq. (4.15). In this way, the time penalty for a charge moving through the NT against the electric field is proportional, by a factor that is proportional to the ratio of the hopping rate, to the time it would take the charge to move along the direction of the electric field if the charge had chosen a forward path of the same distance. As the travelled distance for backwards movement increases, or the electric field increases, the time penalty for making a backwards hop increases.

In fact, this formalism can be generalized if $d_B$ in Eq. (4.15) is replaced by $d_T$ where $d_T$ is the distance travelled by the charge inside the NT which is numerically equal to $d_F$, but positive for forward transport and negative for backwards transport. In that way, Eq. (4.15) can be rewritten as in Eq. (4.17).

$$P_{FB} = \begin{cases} \frac{qEd_F}{k_BT} & \text{for backward transport} \\ \frac{qEd_F}{k_BT} \ast e^{\frac{qE(d_F-d_B)}{k_BT}} & \text{for forward transport} \end{cases}$$  \hspace{1cm} (4.17)

Then the time $t$ for this movement can be rewritten as in Eq. (3.18).

$$t = P_{FB} \ast t_F$$  \hspace{1cm} (4.18)

Now, the total time for a charge to move from one NT to another can remain in the form of Eq. (4.13), with $t$ being given by Eq. (4.18). This, along with $v_0$ being calculated using
Eq. (4.6), allows the entire algorithm to be based on first principles, as intended, in hopes of capturing the full physics of the system without experimental parameterization.
CHAPTER 5

EFFECTS OF THE LOCAL AND APPLIED ELECTRIC FIELDS

The majority of the simulations in Chapter 3 were performed using the original version of the model. Within that version, it was assumed that, on average, the local electric fields within the NT-NT junctions would be equal to the applied electric field. Even though this version did see good agreement with experimental works, it was decided that this assumption needed to be tested in order to see if it indeed is a valid assumption. The study by Cheng et al., which is discussed in more detail in Section 1.1, found that the distribution of the x-component of the local electric field, sampled from random locations within the composite, was peaked either above or below the value of the applied electric field, depending on the relative dielectric constants of the matrix and filler material. The study indicated that the majority of the values of the local field would not be much different than the applied electric field. However, it was also mentioned that there may exist locations within the composite where the electric field may be much larger than the applied field, but those locations were not classified within the study.[108]

The study by Cheng et al., while very helpful and informative, was only done for two dimensional materials and relatively low aspect ratio inclusions. As such, simulations were performed in COMSOL, a finite element based simulation software, in order to determine the local electric fields for various composite materials. The study described below in Section 5.1 differs from the one by Cheng et al. in that it is done in a three
dimensional system made to replicate the composites simulated in this work. It also focuses on the electric fields within the junctions between neighboring inclusions in addition to the distribution of local electric fields throughout the sample.

5.1 COMSOL Studies on Local Fields

Within this work, the hopping model presented previously only considers direct current, where the composite is subjected to a constant electric field. As such, the underlying physical equations to be solved by COMSOL are Maxwell’s equations under the assumption that the magnetic field, $B$, and displacement field, $D$, do not vary in time.

\begin{equation}
\nabla \times H = J + \frac{\partial D}{\partial t}
\end{equation}

\begin{equation}
\nabla \times E = -\frac{\partial B}{\partial t}
\end{equation}

\begin{equation}
\nabla \cdot D = \rho
\end{equation}

\begin{equation}
\nabla \cdot B = 0
\end{equation}

With

\begin{equation}
\frac{dD}{dt} = 0 = \frac{dB}{dt}
\end{equation}

The fact that the curl of $E$ is equal to zero in the electrostatics case leads to Eq. (5.6).

\begin{equation}
E = -\nabla V
\end{equation}

Where $V$ is a scalar potential, in this case the electric potential. Inside a dielectric material, the electric displacement, $D$, can be found as in Eq. (5.7).

\begin{equation}
D = \varepsilon_0 E + P
\end{equation}

Here, $\varepsilon_0$ is the permittivity of vacuum and $P$ is the polarizability of the material. If the dielectric material is assumed to behave as a linear dielectric, then $P$ becomes proportional to the electric field, as in Eq. (5.8).
\[ P = \varepsilon_0 \chi_e E \]  
(5.8)

Where \( \varepsilon_0 \) is the permittivity of vacuum and \( \chi_e \) is the dielectric constant of the material. This then results in the displacement field, \( D \), being proportional to the electric field, \( E \), as in Eq. (5.9).

\[ D = \varepsilon E \]  
(5.9)

Which leads to Eq. (5.10).

\[ \nabla \cdot \varepsilon E = \rho \]  
(5.10)

Where \( \varepsilon = \varepsilon_0 + \varepsilon_0 \chi_e \) and is the permittivity of the material. This combined with the fact that the electric field, \( E \), is equal to the negative gradient of a scalar potential allows the above equation to be manipulated into Eq. (5.11).

\[ \nabla \cdot \varepsilon (-\nabla V) = \rho \quad \text{or} \quad \nabla^2 V = -\frac{\rho}{\varepsilon} \]  
(5.11)

Which is the traditional form of Poisson's Equation in a linear dielectric material. While this is a rather trivial and well-known derivation to Poisson’s Equation, this is the underlying partial differential equation being solved by COMSOL, so a thorough understanding of the background physics is necessary. It should be noted that these studies were performed in the electrostatics module in COMSOL and were performed under the assumption that the dielectric material is behaving as a linear dielectric.

Therefore COMSOL takes the relative permittivity, \( \varepsilon_r = 1 + \chi_e = \frac{\varepsilon}{\varepsilon_0} \), as an input parameter for determining the behavior of the composite material in the presence of an applied electric field.

It is recognized by this researcher that not all materials behave as linear dielectric materials, especially at a nanoscale level. However, it is expected that this study will give
a good overall idea of the behavior of the local electric fields within a composite under an applied, time-independent voltage.

In the beginning of this study, an attempt was made to create NT composite samples with the COMSOL environment by actually working around in the COMSOL GUI. This however was very time inefficient. Thus, in order to streamline the process in setting up a COMSOL simulation, Matlab scripts were used to automatically replicate the virtual composite samples used in charge transport simulations into a COMSOL environment and then solve for the local electric fields at the finite element grid locations. The LiveLink for Matlab module within COMSOL allows for Matlab and COMSOL to take advantage of the Java API language in order to integrate the two software packages. So what might take days to generate in the COMSOL GUI can be performed in a matter of minutes. With this tool box, COMSOL allows the user to save the COMSOL’s .mph file as a model Matlab’s .m file. This .m file then converts every action performed within the COMSOL GUI into the corresponding commands that would be used in a Matlab script while using the LiveLink module. Therefore, the user can do a small scale version of what is desired in the COMSOL GUI, and then save the .mph file as a .m file to see what commands to replicate for scripts to do larger scale tasks. This greatly improved the efficiency of simulations as the task of creating thousands of objects in the COMSOL environment could be automated from the output files of the generation routines and be completed in minutes rather than hours.

In the process of this study, it was discovered that the algorithms that COMSOL uses to set up a 3D grid of finite elements will occasionally produce an error in the COMSOL environment stating that it is unable to appropriately place the grid sites. In
fact, COMSOL would occasionally have trouble even replicating the geometry due to the high aspect ratio nature of these systems and produce the same type of error. The error is well documented within the COMSOL community and is most frequent for curved surfaces in small dimensions, with no available solution other than to alter the geometry slightly. As such, the CNTs were approximated as rectangular prisms in the COMSOL environment, illustrated in Figure 5.1a in the hopes that removing the curved surfaces of the cylindrical NTs would avoid these type of errors. Figure 5.1b shows the correspondence between the cylindrical NTs used in the carrier transport model and the rectangular prisms generated in COMSOL. The length of the diagonal of the face of the prism was set equal to the diameter of the NT. In this way, the rectangular prism does not overlap the boundaries of the NTs generated with the generation routine.

Figure 5.1. a) Simulated COMSOL environment where NTs were approximated as rectangular prisms. b) Representation of cylindrical tube as rectangular prism, where the length of the diagonal on the rectangular face is equal to the diameter of the NT.
Approximating the NTs by these rectangular prisms did indeed avoid many of the errors within the COMSOL. However, to avoid further errors caused by high aspect ratio elements, it is recommended that a mesh always be set up so that the minimum element size is 1 Å. One of the automatically replicated samples with the finite element mesh is shown in Figure 5.2.

![Figure 5.2. The finite element mesh of a replicated sample in Comsol. The denser areas, or where there are more finite element grid locations, are located around the metallic inclusions.](image)

COMSOL comes with a broad materials library and the user can even set up custom materials and give the materials specific properties. However, for the purposes of characterizing the local electric fields in an insulating matrix with metallic inclusions, the matrix in COMSOL was set to poly(methyl methacrylate) (PMMA) while the metallic inclusions were modeled as ideal copper inclusions. The PMMA matrix was given a relative permittivity of 5, which is within the range of experimental values for this material. In terms of static fields, the electric field would be zero inside an ideal conductor, or the copper inclusions in the COMSOL environment. In terms of a linear dielectric material, this means that the relative permittivity of the conductor would be
infinite. Within the COMSOL environment, this is not realizable. Therefore, the system was approximated by giving the metallic inclusions a much larger dielectric constant than the insulating matrix (100 000 for the copper inclusions, and 5 for PMMA). In this way, the copper inclusions are much more polarizable than the insulating matrix. The last step before the simulating the finite element grid was to set up a voltage differential the x direction of the sample to induce an applied electric field.

Four samples with dimensions 100x50x50 nm$^3$ were generated to volume concentrations of 0.01, 0.02, 0.03, and 0.04 with straight NTs oriented along the direction of the electric field. The NTs were 50 nm in length with a 0.5 nm radius. These samples were then replicated within the COMSOL environment as specified above so that finite element simulations could be performed to determine the local fields within these composites.

The value of the x component of the local electric field at each finite element grid point was characterized and binned in each of the samples (Figure 5.3). The maximum and minimum values in the distribution are representative of the maximum and minimum values of the local electric fields found within the sample.
Figure 5.3. Distributions for the x component of the local electric fields at volume concentrations of a) 0.01, b) 0.02, c) 0.03, and d) 0.04.

The distributions within the sample are highly peaked near the applied field. The scale in the horizontal axis was set such that the maximum and minimum values for the local fields can be shown. Therefore, it can be seen also that there do exist hot spots within these samples where the local field is almost 500 times the applied field, $E_0$, which for these simulations was set to 1000 V/m. With the high dielectric constant of the inclusions, the electric field inside the inclusions will be close to zero. Figure 5.4 shows the same distributions as in Figure 5.3, but zoomed in on the peaks of the distributions, in this case only points that were between less than $5E_0$ are shown.
Figure 5.4. Focused distributions for the x component of the local electric fields at volume concentrations of a) 0.01, b) 0.02, c) 0.03, and d) 0.04.

From these distributions, it is difficult to characterize the locations where large local fields may exist. Therefore, to better visualize where hotspots may occur, Figure 5.5 shows the locations where the x component of the electric field was larger than 10E₀ within the 0.04 concentration sample. It can be seen that the hot spots for the x component of the electric field do exist within the sample, but are also denser near the electrode regions at the opposing ends of the x-direction.

Figure 5.5. The locations (red stars) in the 0.04 volume concentration sample where the x component of the electric field was greater than 10E₀. The metallic inclusions are represented by the blue lines.
For the transport algorithms, only the junctions between NTs are important as charges will rarely transition at other points. Therefore, the value of the x component of the local electric field at the midpoint between each of the NT junctions within the four samples was characterized. Since the finite element grid points may not exist exactly at this midpoint, all the grid sites near the midpoint were located, and the one with the largest value for the x component of the electric field was stored. Those field values are plotted for each junction versus the junction distance in Figure 5.6 for the 4 concentrations.

Figure 5.7. The x component of the electric field in the junction between neighbor NTs vs. the junction distance in samples with straight NTs at volume concentrations of a) 1%, b) 2%, c) 3%, and d) 4%. The applied electric field, $E_0$ was 1000 V/m.
From Figure 5.7, the reader might notice that there are set junction distances within the sample. The pixilation method used for many of these studies generates NTs along a three dimensional grid, therefore neighboring NTs will be set distances apart from each other, with those distances corresponding to the size of the grid boxes, which correspond to the diameter of the NTs. In order to show how these set distances can arise, Figure 5.8 shows a two dimensional slice through two parallel segments in NTs with diameters of 5 Å. In a sample such as these where the grid element size was set to this diameter, the minimum distance is shown by the red arrow to be equal to $\left(\frac{5}{2}\sqrt{2} - 2.5\right) \times 2$, or 2.0711.

![Diagram showing set distances](image)

Figure 5.8. Simple representation of how one of the shortest distances can arise in the sample.

A distance of 3.66 comes from two NT end caps occupying two three dimensional cubes diagonally from each other. The distance through the diagonal of a cube with sides of length 5 Å is approximately equal to 8.66 (red dashed line in Figure 5.9). Therefore,
the length of the solid red double arrow, which is the closest distance of approach between the two end caps is approximately $8.66 - 2\times radius$, which is equal to $8.66 - 2\times2.5$, leading to a length of 3.66 Å.

![Diagram](image)

Figure 5.9. Visual representation of how the junction distance of 3.66 can arise within samples. The two solid blue dots represent the two endpoints of two neighboring NTs. Each endpoint is capped with a semi-sphere. The two endpoints would exist in the cubic lattice as diagonal to each other.

It can be seen from Figure 5.7 that many of the junctions have local fields near or even below the value of the applied electric field. However, it can also be seen that the majority of the largest values for the x components of the local electric fields exist where the junction distances are smaller.

Figure 5.10 shows the binned values of the x components of the electric fields in junctions with the four smallest distances within the 0.04 concentration sample studied in COMSOL. A distance of 5 Å was the largest junction distance characterized in this study, as transport through junctions larger than 5 Å, will be very improbable. The exponential
term for tunneling in Eq. (2.2) will be approximately equal to 0.0025 at this distance.

Electric field values at the shortest distances were also examined more closely due to the propensity of the electric field to have larger field values within the smaller junctions, as seen previously in Figure 5.7.

Figure 5.10. The distributions of the x component of the local electric fields between junctions at the four shortest possible distances between NTs. The specified junction distances are in Å.

In order to attempt to capture the overall distribution in local electric fields between junctions, the distributions from all four distances in Figure 5.10 were combined. A Gaussian and Weibull distribution were both fit to the combined data using the Matlab Curve Fitting Toolbox (Figure 5.11). This was done in order to be able to find
a simple representation of the distribution of local fields so that transport simulations
could incorporate these distributions to study the effects of the local field variations.

![Graph of probability density function (PDF) for a) Gaussian and b) Weibull distribution fit to the distribution of local fields in the junctions between neighboring NTs.](image)

**Figure 5.11.** A a) Gaussian and b) Weibull distribution fit to the distribution of local fields in the junctions between neighboring NTs.

The Weibull distribution in Figure 5.11b was chosen as a simple representation of the distribution of fields within junctions due to the tail end of the Weibull fit enveloping
more of the larger electric field values than the Gaussian fit. This would then create a sort of worst-case scenario of the spread of local field values in transport simulations.

Matlab’s Curve Fitting Toolbox fits the simulation data to a general Weibull distribution shown in Eq. (5.12).

\[ f(x) = abx^{b-1}e^{-ax^b} \]  
(5.12)

In terms of general Weibull distributions, the constant \( b \) is the shape parameter of the distribution, and the general scale parameter, \( \lambda \), for a Weibull distribution is given by \( \left( \frac{b}{a} \right)^{-1} \). The Matlab fit led to values of 1.109 for \( b \), and 0.09822 for \( a \), meaning that \( \lambda \) for this distribution would be equal to approximately 8.105. The expectation value for a Weibull distribution is given by Eq. (5.13).

\[ E[x] = \lambda \Gamma \left( 1 + \frac{1}{b} \right) \]
(5.13)

Where here \( \Gamma \) is the gamma function. This leads to an expectation value of approximately 7.7998\(E_\theta\) for this distribution.

Simulations were run with local electric field values chosen according to the Weibull distribution in Figure 5.11 by using the inverse transform method.[134] In Matlab, a uniform random number, \( u \), is sorted between 0 and 1, which is then is transformed into the variable, \( X \) with a target distribution. This is done by equating the cumulative distribution of the target function to the uniform random variable, as in Eq. (5.14).

\[ u = F(X) = 1 - e^{-\left( \frac{x}{\lambda} \right)^b} \]
(5.14)

Inverting the equation then solves for \( X \) in terms of \( u \).

\[ X = \lambda (-\log (1 - u))^\frac{1}{b} \]
(5.15)
Within these simulations, the local field would be determined by multiplying $X$ by the applied electric field. In order to ensure that the local fields within junctions were varied as desired, carriers were then run through a test sample where the local electric fields varied according to the Weibull distribution. Each time a charge attempted to move to neighboring NTs, the value chosen for the electric field in each junction between neighbor NTs was calculated according to the Weibull distribution and then stored. The distribution of local electric fields from the test simulation is shown in Figure 5.12, and indeed indicates that the values followed the target Weibull distribution.

![Figure 5.12](image)

Figure 5.12. The distribution of values for the x component of the local electric fields within each junction classified for the charge transport routine.

Once the correct implementation was assured, three separate simulations were performed on samples that contained wavy NTs and a high aspect ratio. In the first, the field within the junctions was considered a constant and equal to the applied electric field. In the second, the local fields were varied according to Eq. (5.15), while in the third
the local fields were kept constant, equal to the expectation value of the distribution, which is given by Eq. (5.13). The NTs were 225 nm in length and with a radius of 0.5 nm in samples with dimensions that were 1 x 0.2 x 0.2 μm³. For these simulations the latest version of the hopping algorithm, as described in Section 4.4, was used. The conductivity values of these simulations are shown in Figure 5.13.

![Figure 5.13. Conductivity results for varying the electric fields within samples using version 2 of the transport routine.](image)

The conductivity is slightly larger for the cases where the electric fields in the junctions are set according to the Weibull distribution or the expectation value of that distribution than in the case where the applied electric field is used. This indicates that although using the applied field as the field in the junction leads to an under prediction of the conductivity, the use of a constant electric field is a reasonable approximation, if the mean value of the electric field distribution is used. From these results, it can be seen that results obtained throughout this dissertation may not correspond to what would be expected at the electric field chosen for those simulations. However, they are expected to
correspond to a case with a smaller electric field, which from the case studied here can be expected to be about 8 times smaller. Thus far, this section has focused on the x component of the electric field. This is because in many of the studies within this work, it was assumed that the local fields within the junctions between neighboring NTs would on average be dominated by the electric field along the x direction. However, it was of interest to study the behavior of the y and z components of the local electric fields within junctions to check this assumption. Figure 5.14a and b show the distributions of y and z components of the electric field within the junctions, respectively, within the 0.04 volume concentration sample. This concentration was chosen as it had an acceptable number of NT-NT junctions.

![Figure 5.14. The distribution of the average values of the a) y component and b) z component of the fields within the junctions between neighboring NTs.](image)

It is clear that the polarization of the conductive inserts will generate y and z components of the electric field. However, the distribution is centered and peaked about zero, indicating that these components on average will be zero. In order to see how these induced polarizations may affect the local electric fields within the junctions, the total
value of the electric field at the midpoint of each NT-NT junction is characterized for all
four concentrations. Figure 5.15 shows a scatter plot of the local norm of the electric field
at the midpoint of each junction versus the distances of the junctions.

![Junction Fields — $E_0$](image1)

![Junction Fields — $E_0$](image2)

![Junction Fields — $E_0$](image3)

![Junction Fields — $E_0$](image4)

Figure 5.15. The norm of the electric field at the midpoint in each junction versus the
junction distances for concentrations of a) 0.01, b) 0.02, c) 0.03, and d) 0.04.

As with the x components, there are locations within junctions where the local
field is larger than $E_0$, with the larger values tending to exist in the shorter junctions.

Thus, the distributions of the total value of the electric field for the four shortest junction
distances in the 0.04 volume concentration sample are shown in Figure 5.16.
Figure 5.16. The distributions of the local electric field norm between junctions at the four shortest possible distances between NTs.

These four distributions were then combined and fit to a Weibull distribution (Figure 5.17), as they were in the case of the x component of the electric field. The fit was a least squares fit performed in the Matlab Curve Fitting Toolbox using the least absolute residual method.
For this fit, the parameters, $a$ and $b$, in Eq. (43) were 0.04852 and 1.443, respectively. The R-square for this fit was 0.9961. This compares to values of 1.109 for $b$ and 0.09822 for $a$ in the general Weibull distribution found for the x component of the electric fields within junctions. The Weibull distributions found in Figure 5.11 and Figure 5.17 are shown together in Figure 5.18 to compare the distribution of the local norm of the electric field to the distribution of the x component of the electric field. It can be seen that the Weibull distributions are actually quite similar, indicating that overall, the y and z components of the local electric field will not significantly alter the value of the electric field within the junctions, and the x component of the junction should be sufficient to describe charge transport within each junction. However, in the interests of complete understanding, future works are planned to incorporate these induced polarizations in the y and z directions.
Figure 5.18. The Weibull distributions for the x component and norm of the electric field in the midpoints of the junctions for the four shortest distances in the 0.04 concentration sample.

5.2 Comparison to Experiments with the Improved Transport Routine

Once the model was altered, as described in Section 4.4, and the behavior of the local electric fields within NT-NT junctions was better understood, it was again of interest to see how well the model matched up with experimental values when the parameters within the model were solved for using first principles. Simulations were run with the improved charge transport routine on samples that were 1x0.2x0.2 μm$^3$ in length with an applied electric field of 10 V/cm. The results were then compared to the same experimental results on PAN/CNT composites used in Section 4.4, shown in Figure 5.19. The extrapolation of the results was performed by again fitting simulation results to the power law for conductivity, Eq. (1.1).
Figure 5.19. Comparison to experimental results from Guo et al. [125] and Pradhan et al. [120] with results under an applied electric field of 10 V/cm.

This version of the model also sees excellent agreement with the experimental results. This is significant as this version of the model is based solely on the physics of the system, with no fitting parameters to be used for the simulation. Now it does need to be noted that with the original version of the model, the applied electric field was set equal to 1 V/cm in the simulations in section 4.4, not 10 V/cm as in this case with the improved version of the model.

To begin to explain this, this author must note that in the simulation with the improved model, the electric field within the junctions was set to be the expectation value of the distribution of the electric fields within junctions (see Section 5.1). This expectation value was around 8 times the applied electric field. It should be remembered from Section 4.2 that \( v_0 \) is calculated such that it is proportional to the junction electric field in the improved version of the model. This means \( v_0 \), for an applied field of 10 V/cm, would have been \( 8.54 \times 10^6 \) s\(^{-1} \), leading to a minimum hopping time of \( 1.17 \times 10^{-7} \) s.
In the comparison using the original routines, the value for $v_0$ was found to be around $0.8 \times 10^4 \text{ s}^{-1}$, which would give a minimum time for a hop that was equal to $1.25 \times 10^{-4} \text{ s}$. The difference in the time scale for hops using the two versions can be explained by the fact that in the original version, the charge was assumed to travel down the NTs without taking any time. The improved model adds the time for the charge to move through the NT, but also added a larger time penalty for hops when the charge moves through a NT against the force of the electric field (see Section 4.4). Therefore, the time scale for hops within this improved version of the model should be smaller to account for this.

In the experimental studies on PAN/CNT composites, used for comparisons within this work, it is not reported what the voltage was that was used for measuring the conductivity of these composites. Rather, the experimental studies used here for comparison are more concerned with conductivity at and around the critical concentration and the factors leading to the observed increase in conductivity.[120], [125] In fact, many of the experimental works with CNT composites are similar in that the voltages, and therefore the electric field, used to study DC conductivity are not usually reported. It is therefore unknown what the applied electric field was in these experimental studies. This can be problematic as changes in the voltage can lead to differences in electrical properties of these materials, and make it difficult to compare to experimental results.

### 5.3 Effect of the Applied Electric Field on Conductivity

In Section 5.1, it was determined that the distributions of the local electric fields within junctions do not have significant impacts on charge transport, within the framework of Eq. (3.12) where $v_0$ was made dependent of the electric field. It was therefore of interest to see how changing the applied electric field on samples affects
charge transport within these simulations. Samples that were 1 x 0.2 x 0.2 µm³ in dimensions with NTs having a maximum length of 200 nm with a radius of 0.5 nm were simulated with an applied electric field of 1 V/cm, 10 V/cm, and 100 V/cm. The local fields within junctions were assumed to be the average of the distribution of the local fields, which was around 8 times the applied field (see Section 5.1). Figure 5.20 shows the conductivity for the samples under the 3 different applied electric fields.

![Conductivity results for varying the applied electric field.](image)

Figure 5.20. Conductivity results for varying the applied electric field.

It can be seen that by increasing the electric field applied to the sample the conductivity of the samples increases. A larger electric field will give more energy to a charge moving down a conductor, or a NT in these simulations, which in turn effectively lowers the tunneling barrier between two neighboring NTs. This is captured within this model as the probability for downfield hops between neighboring NTs increases with increasing electric field, therefore decreasing the probability for a backwards hop to occur. This same effect has also been seen within other models that incorporate tunneling into their model. In the theoretical work done by Bao et al., a decrease in the tunneling
barrier height between neighboring NTs also led to a decrease in conductivity within composite samples.\cite{95} In addition, $v_0$ in Eq. (3.12) is calculated such that it is proportional to the electric field within the junction. Therefore, increasing the electric field will increase $v_0$, which then will decrease the transport time for hops. With this in mind, a further look at the calculation of $v_0$ for specific samples is needed and discussed in the future plans section of this work.
CHAPTER 6

CONCLUSIONS

A stochastic approach has been taken to model charge transport properties in CNT composite materials. Simulation boxes as large as 1 \( \mu \text{m} \) long have been modeled with over 5000 NTs that were 0.5 nm in diameter and up to 225 nm long (1:450 aspect ratio). This shows the capability of the model to study samples that can be directly compared to experiments while keeping a nanoscopic resolution on its physical properties. The results presented in this work see quantitative agreement with experimental works, and predictions from the model are qualitatively sound on the relationship between various aspects of geometries and conductivity.

When it comes to practical aspects of the simulations, it is determined that the simulation box needs to be 4 to 5 times longer than the NTs for the diversity in topology to be enough to mask size effects. Above this size, carriers will sample a configuration that is diverse enough to ensure that increasing the size further does not significantly alter results. The progress towards a more diversified sample is observed to be smoother for samples with NTs with a larger aspect ratio. Finally, notice that by simply re-scaling some parameters such as the tunneling constant, these simulations can be scaled up to larger sizes.

Qualitatively, results from the model show an excellent agreement with the percolation behavior expected for these samples. The model also predicts the correct
relationship with NT waviness and aspect ratio that are here explained from the structural features at the nanoscale. Indeed, as the junction resistance is the limiting factor, low aspect ratios or high waviness will lead to more NT-NT junctions between electrodes, decreasing conductivity. Theoretical percolation does not account for the junction resistance and transport in a percolated network will be insensitive to aspect ratio or waviness. Conductivity is found to be more sensitive to aspect ratio than waviness, although waviness does affect conductivity.

Simulations have been performed to understand how agglomeration impacts charge transport within nanocomposite materials. Results indicate that heavy agglomeration is indeed detrimental to charge transport in these materials. However, it has been shown that for optimal charge transport, especially at low concentrations, the formation of agglomerates actually improved transport. Interconnected bundles of NTs connected to each other by NTs that stick out like hairs from each agglomerate allow for better charge transport than both randomly dispersed NTs and large bundles of NTs that are poorly interconnected. This is due to more efficient distribution of conduction paths that leaves voids in the sample in favor to more densely packed groups of NTs that connect the electrodes. The effect will be almost imperceptible above percolation and will have a larger impact at the onset of percolation and at the threshold. In fact, some agglomeration leads to a lower percolation threshold. Looking in more detail, it is observed that as concentration grows into early percolation, more uniform samples do lead to a larger number of carriers reaching the collection electrode, but with a lower mobility. At these relatively larger concentrations, pathways exist at almost all degrees of agglomeration. However, for a more uniform distribution, the number of paths is larger
than for the percolated samples and thus more carriers reach the collection electrode.

Nonetheless, the average distance between NT’s is also larger in those uniform samples, explaining the lower mobility.

It is evident from this work that the local electric field within composites will not be equal to the applied electric field. In fact, there will be hotspots within these composites where the local electric field is significantly larger than the applied electric field. It has also been shown that there are points within junctions between NTs where the electric field is actually lower than the applied electric field. This work has shown that the distributions of local electric fields within composites with a conductive filler and insulating matrix will have minimal effects on charge transport routines within these materials. Thus, assigning the same electric field value to all the junctions would produce very similar results compared to assigning valued based on a distribution, provided that the value assigned is the mean value of the distribution of electric fields. It does remain a fact however that the local electric fields within composite materials will not necessarily equal the applied electric field and this must always be kept in consideration.
CHAPTER 7

RELATED AND FUTURE WORKS

7.1 Further Improving this Model

While this model does indeed capture many qualities of charge transport within CNT composites, there are a few characteristics of the model that need to be studied further. Improvements have been made in attempts to characterize systems from first principles using the fundamental physics of the system. In an attempt to do this, \( v_0 \) has been changed from a fitting parameter, to a parameter that is based on Ohm’s law and DFT calculations. A further study into the validity of this change needs to be performed. First, \( v_0 \) has been made to be proportional to the electric field within the junction between NTs, which may or may not be an appropriate value. As has been presented in Section 5.1, the local fields within a composite sample do vary due to the propensity of the conductive inclusions to polarize within the insulating matrix. If \( v_0 \) is to be continued to be calculated in this way, then a possible improvement may be to calculate the voltage differential between neighboring CNTs. In this way, the voltage drop between neighboring CNTs, and therefore the voltage drop across the junction resistances can be known and applied to the calculation of \( v_0 \). This type of study can be performed in Comsol as were the studies presented in Section 5.1, and could potentially show that the local voltage drops do indeed vary widely within the material.
This would be slightly different than the study presented in Section 5.1, as the
distribution of electric fields within junctions was for junctions where the tunneling
distance was greater than zero. In this coarse-grained model, it is assumed that the NTs
are touching when the tunneling distance is zero. Therefore, when calculating the fields
within junctions in the replicated geometries in COMSOL, there would have been no
junction distance to characterize when these NTs are touching. The junction distance
would be zero, and therefore the field infinite, if COMSOL is even registering a drop in
voltage across the NTs where they touch. However, it could be that an average voltage
within the smallest tunneling distance may be a suitable means to determine the average
field across a junction resistance. This type of study is outside the scope of this work, and
reserved for future works.

Also, within this work, only the x component of the electric field is used within
the junctions. It was seen in Section 5.1 that the y and z components of the field, on
average, will equal to zero. However, this does not mean that those components will be
zero in individual junctions. Therefore, it is planned to incorporate these components of
the junction electric fields into the transport routines. This will be done by further
classifying those components using similar methods as found in Section 5.1.

7.2 Voids or Non-Conductive Inserts

In many studies involving CNTs, there are often times other nanoparticles
inserted into the material for various reasons. For example, there have been studies with
CNTs embedded into polyurea materials along with halloysite NTs in the hopes that each
one may contribute to enhancing a particular property of the material leading to the
overall improvement of the properties of the polyurea.[135]–[137] Polyurea is a well-
known and versatile material that is used for various coatings, containers, structures, and components due to their chemical and mechanical durability.

In light of this, it was determined that the model needed the capability to consider voids within the sample in order to simulate a non-conducting nano-insert. Not only would it be useful if the desired composite contains multiple types of NTs, but also in the event that the specific polymer has a specific geometry that does not allow the nano-inserts to occupy a certain space. In the above-mentioned example of polyurea, the polyurea, upon the mixture of the two precursor components, will form soft regions and hard crystalline regions. In the event one wanted to simulate a geometry where the nano inserts were only dispersed within the soft region, this type of generation would be useful.

In the current implementation, two types of voids can be created, cylindrical or spherical. Cylindrical voids are created by considering a straight line between two random points within the sample and then giving a cylindrical volume where those two points are connected, with semispherical end caps on the ends. In the case a spherical void is desired, these two points in space could just simply be created in consecutive grid points. As in the other generation methods, it is ensured that these new nanoinserts do not intersect with each other so that accurate volume concentrations are reached. Once these nanoinserts have been generated within a sample, the NTs are placed around them, as previously discussed. Samples with different volume percentages of the second type of nanoinsert are shown in Figure 7.1. The normal NTs are shown in green while the second NTs are shown as voids in the sample. For this simple visualization, the CNTs were 0.5 nm in diameter and 30 nm in length while the voids were 3 nm in radius and 20 nm long.
Samples were 100x30x30 nm³ in dimension. The NTs were kept at a constant volume fraction of 0.04.

![Image](image_url)

Figure 7.1. Samples with a second type of insert, shown here as voids, for volume concentrations of a) 0.00 and b) 0.05. The volume concentration of the normal NTs (in green) was kept constant.

Clearly there is an increase in the amount of white space (voids) as the volume concentration of the second type of nanoinsert is increased. This type of generation adds to the versatility of the types of materials that can be studied.

7.3 Further Characterization of Various Composites

Further work can be done within this model to classify other composite materials. It is planned that continued work will be done in conjunction with the Air Force Research Laboratories and associated collaborators to classify not only CNT composites, but other composites with other various conductive nanoinserts and various polymers. The model is at a point that any polymer material can be classified for tunneling properties and employed within this model. Materials with nonconductive inserts can also be studied within this model through the use of the portion of the routine that creates voids within
the samples. These voids can also be modeled as a second type of insert that has different conductive properties, whether insulated or semiconducting.

In addition, the current NTs within the samples here are all considered metallic. However, the model can be parameterized to more appropriately represent CNT composites by making a portion of the NTs semiconducting in nature. Generally speaking, 2/3 of the possible ways that a sheet of graphene can be rolled, or the chirality of the CNT, will lead to semiconducting CNTs for single-walled CNTs. The remaining 1/3 will be metallic. In addition, large diameter single-walled CNTs and many multi-walled CNTs tend to be metallic in nature.

A metallic CNT will have a relatively small gap between the valence and conduction states, and therefore it can be approximated that a charge at the Fermi level can easily be promoted into the conduction band of the CNT. However, semiconducting NTs will have an energy gap between the valence and conduction states, and therefore any charge attempting to move to or through a semiconducting NT will have to receive some amount of energy to promote charges into the conduction state. The inclusion of semiconducting NTs can lead to three new situations within the transport routine: 1) A charge attempting to move from a metallic CNT to a semiconducting CNT, 2) A charge attempting to move from a semiconducting CNT to another semiconducting CNT, or 3) A charge moving from a semiconducting CNT to a metallic CNT. In case one, the charge moving in the metallic CNT attempting to transition to a semiconducting CNT will have to overcome the energy in the gap between the Fermi level of the metallic CNT and the conduction state in the semiconducting CNT. In case 2, the charge attempting to move between two semiconducting CNTs will move from the conduction state in the first CNT
to the conduction state in the second CNT. In case 3, the charge will have to move from the conduction level in the semiconducting CNT to the Fermi level of the metallic CNT. The exact references for the Fermi level in each CNT and the relative gaps between conduction and valence states in semiconducting CNTs will need to be known to incorporate these into the transport routine.

7.4 Effects of Structural Changes on Charge Transport

7.4.1 Cracks in Samples

Simple simulations were performed in order to study how changes in the morphology of a CNT composite material might affect charge transport through these materials. One of the major structural defects that can occur in many materials under various loading amounts is micro cracking. In order to structurally monitor one such instances of damage, the structure to be monitored could be a nanocomposite itself or a nanocomposite coating can be applied to the structure. If structural changes relate to changes in conductivity, then those changes in conductivity can be used to sense changes in the structure, particularly cracks. In order to accomplish this, a short routine was created that takes a previously generated sample, and incrementally cracks the sample from bottom-to-top in a user-specified number of different sections. In the first cracking step, the bottom section was spread apart a specified amount. In the second cracking step the bottom section was spread apart once again the same amount, while the next section above was also cracked. In this way, a crack begins at the bottom of the sample, and grows upwards towards the top of the sample. Figure 7.2 shows a sample, containing 6% volume NTs, cracking in various stages where each cracking step spreads the bottom
section of the sample by 0.5 Å. Figure 7.3 shows the number of charges, out of 1000, that are collected at the opposite electrode for each of the cracking steps.

Figure 7.2. A simple crack forming through the middle of a virtual nanocomposite sample.

![Graph showing charges collected vs. cracking step.](image)

Figure 7.3. The number of charges, out of 1000, that reach the opposite electrode at each cracking step.

Clearly there is a drop in the number of charges that are able to reach the opposite electrode as the crack develops through the middle of the sample. However, the
maximum width for the crack at step fourteen was 7 Å at the bottom of the sample, tapering towards the top of the sample. This crack width is still small enough that some charges are still able to reach the opposite end of the sample. So charge transport is inhibited somewhat, but not fully stopped. In fact, most charges travel through a path that runs through the top of the sample, as is illustrated in Figure 7.4, where the path of one charge is shown running through the top of the sample. The color, blue to red, represents the time, early or later, in the transport that each NT was visited by the charge.

![Figure 7.4](image)

Figure 7.4. The path taken by one charge, highlighting the visited NTs.

It was of interest to see the type of crack that would be needed to cause a more drastic change in charge transport through these samples. In order to accomplish this, another crack was induced in a sample containing a larger volume percentage, 8%, of NTs to ensure a quality conductive sample before cracking. In this sample, the crack again began at the bottom section, with the sections being spread by 2 Å at each step. The final crack is shown Figure 7.5 and compared to the thinner crack from the 6% volume sample, which was taken from Figure 7.2.
Figure 7.5. a) The thinner crack from the 6% volume NT sample and b) The wider crack in the 8% volume NT sample.

Figure 7.6 shows the number of charges, again out of 1000, that reach the opposite electrode in both these cracked samples. There is a drastic drop in the number of charges that reach the opposite electrode for the sample with the much larger crack. The tunneling parameter for these two samples was $0.6 \, \text{Å}^{-1}$, the value calculated for PAN. At the tenth crack step, the top sections of the samples begin their first separation of $0.5 \, \text{Å}$ and $2 \, \text{Å}$ for the small and large cracks respectively. In the 8% volume NT sample with the large crack, it is at this point, the tenth crack step, that the steep decline in charges reaching the opposite electrode begins to occur. So, for this material, if a crack begins to form that is any greater than $2 \, \text{Å}$ throughout the sample, charge transport will begin to be inhibited by the crack. If the electrical conductivity of this material is constantly monitored in the correct way, the formation of cracks could be detected even before they are visible without the use of advanced microscopes.
Figure 7.6. The number of charges that reach the opposite electrode for the 6% volume sample (with a thinner crack) and 8% volume sample (with the thicker crack).

Now the above work is detailed on a very simple crack forming through the middle of the sample. This very preliminary work shows that the model can show that if there is a crack through the sample, the conductivity of the sample will be lowered, and should be able to be sensed by monitoring the electrical response of the material.

However, a more accurate prediction of a crack forming through the sample is needed for this application of the model to continue. What would be most advantageous would be to take into account the structural and mechanical properties of the composite, either through experimental or modeling approaches, so that an understanding of how the material would crack could be determined. If the structural damage or deformation of the composite could be predicted, then the electrical response of the material during and after the formation of the crack could also be predicted. More detailed study and work is required for this application of the model to reach its full potential.

7.4.2 Effect of Compression on Charge Transport

Another type of structural change that occurs frequently in these types of materials is that of compression or stretching due to external forces. Simple compression
algorithms have been developed to compress virtual samples from all directions towards the center of the sample. No mechanical properties of the underlying polymer or CNTs within the sample have been taken into account. The CNTs were simply pushed closer towards the center of the sample, however, the point to which the CNTs were being compressed can be altered by the user. Figure 7.7 shows how samples could be compressed.

Figure 7.7. Compression of samples towards the center of the sample. Increasing compression from left to right.

Clearly the volume concentration of the CNTs increases as the sample shrinks.

Figure 7.8 shows the percentage of injected charges reaching the collection electrode at various compression steps.

Figure 7.8. Percentage of charges reaching the collection electrode as the sample is compressed.
The model already has the capability to predict conductive properties of samples, and to compress samples. The main task for this portion of the model to have a greater impact in future studies is for the mechanical properties of each material to be taken into account in the compression routines. As of right now, the NTs in the samples are moved arbitrarily, when in reality the compression of the sample and the movement of the filler is going to be dependent on the mechanical properties of both the filler and the polymer.

7.4.3 Vibrational Studies with Charge Transport

Another possible future work that can be performed with this model is to study the effects of vibrations in these materials. The idea behind this work would be to monitor materials through changes in electrical resistance of nanocomposite coatings. If one could monitor the vibrational frequencies of the underlying material through electrical resistance measurements, any change in the vibrational frequencies of the composite itself or the underlying material would indicate either a change has occurred in the material, or some event has occurred to the material.

The method for pursuing this study would be to induce vibrations within the material, and then monitor charge transport characteristics versus various types of vibrations and various frequencies. The way this would be implemented into the current model is to track at what time the charge reaches a particular NT. As the user is the one determining the vibrational pattern, or the equations governing the vibrational motion of the material, the change in position of the material can be known at that time and incorporated into the transport routines.

Through studies such as this, the conductivity of samples for various types of vibrations, such as normal modes, or say a compression wave through the sample, may be
known. If for example, a compressive shock wave propagates through the material, and the conductivity profile for such a wave is known, one can easily monitor such an even occurring.

For this work to progress, one must take into account the micro and nano-scale mechanics in the composite. The exact interactions of the material may need to be taken into account. The underlying normal modes of various composites will also need to be known.

### 7.5 Polymer Hydrogel Simulations

Recently there has been much study on polymer hydrogels and their use in controlled drug delivery. Hydrogels are polymer materials that swell in the presence of water and have been widely used in various materials ranging from contact lenses to absorbent materials in diapers. The main premise behind these hydrogels is that they could behave as smart materials for controlled drug delivery due to their propensity to swell when water is introduced to the polymer. They then will shrink back once the water has been removed. The drug to be delivered can be loaded into a swollen hydrogel, which can be then be dehydrated, leaving only the drug in the shrunken hydrogel material. These materials have been widely studied both experimentally and theoretically and swelling rates have been established. However, modeling of these materials can allow further understanding of the nature of the swelling hydrogel depending on the type of hydrogel.

A portion of the model developed in this dissertation has been used in preliminary studies of polymer hydrogels for drug delivery in order to simulate the release of various drug molecules in Monte Carlo dynamics simulations. The first task to simulate these
processes is to generate samples that resemble hydrogels in geometry. The generation routine discussed in Section 2.1 has been altered slightly so that polymer strands are represented by connected cylindrical segments, as were the NTs. However, these coarse grain polymer strands are allowed to overlap with each other to more accurately represent the physical and chemical crosslinking that occurs in these hydrogels. This crosslinking creates a matrix type structure in these hydrogel materials, which is one of the main reasons why they can swell and shrink.[138] The generation routine can be easily adapted for the physical crosslinking of the polymer strands. However, for chemical crosslinking of the polymer, any point at which two strands crosslink must remain attached to each other, and will be a major factor in determining how the polymer hydrogel swells and/or shrinks.

In addition, the routine has been adapted so that drug molecules can be added within the polymer hydrogel matrix, or outside the matrix. Each molecule is also a coarse grain approximation as a simple sphere, or as a straight cylinder depending on the drug molecule being studied. In the preliminary Monte Carlo dynamics simulations, the same neighbor-checking algorithm that has been used to find the nearest NT neighbors as in Section 2.2 has been employed in order to calculate the distance between polymer strands and drug molecules. A sample is shown in Figure 7.9.
The ability to use these routines and algorithms has led to an efficient means to begin the process of studying and modeling these materials. Further work to continue studying drug delivery with hydrogels is needed. The physical interactions between polymer strands and drug molecules will need to be calculated and included into the dynamics simulations. Previous works have studied drug delivery using halloysites, a naturally occurring clay NT, and many of the underlying physical equations used in that study could be applied to the dynamics simulations for drug delivery in hydrogel systems in a synergistic interaction with this work.[139]
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