

MIDSX: A Monte Carlo Interaction and Dosimetry Simulation of X-rays

John Meneghini^{a)}

Department of Physics, Saint Vincent College, Latrobe, PA 15650

^{a)}Corresponding author: john.meneghini@stvincent.edu

Computed Tomography (CT) imaging, while essential for diagnostics, exposes patients to ionizing radiation. To accurately quantify radiation dosage, this study introduces MIDSX, a specialized open-source Monte Carlo (MC) photon transport code system for x-ray transport in medical imaging. Unlike broader MC systems, MIDSX is tailored for x-ray transport, offering more streamlined implementation. Results from validation simulations show MIDSX's results for specific cases agree to within 0.32% of the mean of established reference data. However, discrepancies in body energy deposition measurements, reaching up to a 6.3% mean percent error, indicate areas for further refinement in the system.

I. INTRODUCTION

Computed Tomography (CT) imaging is a critical diagnostic tool used by medical professionals to diagnose various illnesses and injuries. While the use of CT imaging is essential to provide immediate, life-saving results, it is critical to recognize and assess the dangers associated with ionizing radiation. In particular, ionizing radiation can damage DNA and increase the risk of cancer in patients. While the risk is small, it is cumulative, so it is important for physicians to track a patient's radiation exposure over time¹. To track said exposure, we must first be able to quantify and determine the dosage of radiation a patient receives during an imaging test. Typically, this exposure is measured using absorbed dose in the body and air kerma at the skin layer, which both have units of grays (joule/kg), making it directly related to the energy deposited/incident by photons and their secondary particles. Often, these values are estimated based on the properties/settings of the x-ray source and the specific procedure being performed. However, the most accurate technique is the use of Monte Carlo (MC) methods to simulate the propagation of photons through a computational phantom².

In the MC technique, the 3D space encompassing the phantom and the radiation source is represented as a computational domain. Within this domain, individual photon interactions are stochastically simulated, accounting for each interaction event as photons navigate the phantom. This complexity in simulation offers unparalleled precision, capturing even the most subtle nuances of radiation behavior in biological media. Additionally, the MC method can simulate various medium types, densities, and configurations, making it incredibly versatile and adaptable to varying imaging tests. Furthermore, advancements in computational power and algorithms have expedited the MC simulations, rendering them more accessible and feasible for routine clinical applications³.

In this paper, the newly developed open-source MC photon transport code system MIDSX is presented and validated. While many existing MC transport code systems have shown reliability in dosimetry applications^{3,4}, many of these systems are tailored for general particle transport. While the capabilities of these code systems are far beyond that of MIDSX, the developmental focus on x-ray transport has reduced the complexity from an implementation point of view. In turn, it allows users to easily design and run simulations specifically relating to x-ray transport in the medical imaging energy range. The subsequent sections will delve into the theory of MIDSX, its methodology, and comparative analysis with other established systems.

II. TRANSPORT THEORY

To represent the computational domain discretely, space is broken up into a grid of voxels (a pixel with volume), with each voxel being assigned a particular material ID depending on the geometries and compounds/elements in the domain. Within our discrete space, given a photon position \vec{r} , the corresponding voxel in which the photon resides can be calculated. Therefore, all possible \vec{r} 's can be assigned a particular material M in the domain.

A photon's position in space after taking the n -th step in the domain, \vec{r}_n , is represented by the following parametric ray equation:

$$\vec{r}_n = \vec{r}_{n-1} + \hat{d}t_n, \quad (1)$$

where \vec{r}_{n-1} is the initial position before the n -th step, \hat{d} is a unit vector in the direction of the step, and t_n is the length of the n -th step.

To randomly sample t_n in a homogenous domain, we utilize the following probability density function (PDF) $p(t)$ of the distance traveled t by a photon of energy E through material M before interacting:

$$p(t) = n\sigma \exp[-t(n\sigma)], \quad (2)$$

where n is the number density of M and $\sigma = \sigma(E, M)$ is the microscopic cross-section of M at E .

Using the inversion method for sampling a PDF on Equation 2, it follows that random values of the free path t can be generated with the following equation:

$$t = -\frac{1}{n\sigma} \ln \gamma, \quad (3)$$

where γ is a uniformly distributed random number in the interval $[0, 1]$. This value of t is sampled for each step and is used as t_n in Equation 1 to determine the length of the n -th step.

For further reading on the theory presented in this section, see⁵.

A. SURFACE AND δ -TRACKING

If, after taking a step, the photon lands in a voxel with a different material (an inhomogeneous domain), then the corresponding free path for the new material must be accounted for. One method, called surface-tracking, requires photons to be stopped at voxel boundaries and intersections with surrounding voxels to be calculated, which can be computationally intensive for materials that have a large average free path.

Alternatively, the δ -tracking algorithm offers a solution by sampling the maximum cross-section σ_{\max} in the computational domain. This, in turn, brings down the average free path to the minimum in the domain. To account for this decrease in free path, the algorithm introduces δ interactions as an alternative to real interactions, resulting in no change to the energy or direction. The probability of a δ interaction P_δ is given by the following equation:

$$P_\delta = \frac{\sigma_{\max}(E) - \sigma(E, M)}{\sigma_{\max}(E)}, \quad (4)$$

where E is the energy of the photon undergoing the step. Note that when the photon lands in the material corresponding to the maximum cross-section, $\sigma(E, M) = \sigma_{\max}$ and $P_\delta = 0$. On the contrary, if the photon landed in air and the domain's maximum cross-section corresponded to lead, then $\sigma(E, M) \ll \sigma_{\max}$, making $P_\delta \approx 1$.

Overall, δ -tracking is significantly more computationally efficient for domains with similar cross-sections and can be shown to yield equivalent results to surface-tracking.

For more details on these algorithms, see⁵.

B. PHOTON INTERACTIONS

If a δ interaction does not occur, then a real interaction is sampled. Therefore, the probability of a real interaction P_r is directly related to P_δ by

$$P_r = 1 - P_\delta, \quad (5)$$

where P_δ is given by Equation 4.

If a real interaction occurs in material M , then the probability of interaction i occurring is $P_i = \frac{\sigma_i(E, M)}{\sigma(E, M)}$ where σ_i is the cross-section of interaction i . If there are N possible interactions for a particular E and M , then $\sigma(E, M)$ is calculated as so

$$\sigma(E, M) = \sum_{i=1}^N \sigma_i(E, M). \quad (6)$$

For x-rays, there are three possible photon interactions: photoelectric effect, coherent scattering, and incoherent scattering. For the photoelectric effect in MIDSX, the photon is terminated and all energy is deposited at the location of interaction. In general-purpose particle transport

code systems, when a photoelectric interaction occurs, a photon of energy E is absorbed by an electron in subshell i , causing the electron to leave the atom with energy $E_e = E - U_i$, where U_i is the binding energy of the i th subshell. In addition, photons are emitted due to atomic relaxations. For photon energies in the medical imaging range (<120 keV), the energy of the released electrons does not allow for significant traversal through typical biological media. This limited traversal results in a localized dose distribution, in turn, validating the model used by MIDSX.

For coherent and incoherent scattering, the methodology of⁶ was adapted for use in MIDSX, neglecting Doppler energy broadening and the production of secondary particles.

III. IMPLEMENTATION

To simulate x-ray transport in MIDSX, the domain's geometry is represented using a 3D voxel array with each voxel assigned a material ID. Geometries are defined by NIFTI files⁷ and incorporated into a custom .domain file. These geometries are represented by `VoxelGrid` objects which are contained within `ComputationalDomain` objects.

Materials are defined in an SQLite database⁸, which also contains elemental data from⁹ and¹⁰, and interaction data from the EPDL database¹¹. The `InteractionData` object initializes this material data with various interpolation methods, including linear and spline interpolation.

The `PhotonSource` object, initialized with `SourceGeometry`, `Directionality`, and `EnergySpectrum`, determines the photon's initial attributes. MIDSX provides measurement triggers through surface and volume tallies. The geometries of these tallies correspond to regions of interest (ROI) and volumes of interest (VOI), respectively. Users can specify desired tally quantities via the `Quantity` and `QuantityContainer` objects, with post-simulation derived quantities like air kerma computed using the `DerivedQuantity` object.

For more details on the implementation of MIDSX, see the source code¹².

IV. RESULTS

To validate the accuracy of MIDSX, validation simulations were performed and compared to reference data obtained by the American Association of Physicists in Medicine Task Group Report 195 (TG-195)¹³. The simulations performed from TG-195 were Case 1: "Half Value Layer," Case 2: "Radiography and Body Tomosynthesis," and Case 5: "CT with a Voxelized Solid." For Case

1, the primary air kerma was measured on a far away, circular ROI with a cone beam point source collimated such that all primary particles would be incident upon the ROI. The primary air kerma was measured with the domain filled only with air and then compared to the measured air kerma with an Al filter of thickness t placed between the source and ROI. The ratios of the half value layer (HVL) and quarter value layer (QVL) primary air kerma to the primary background air kerma is represented by R_1 and R_2 , respectively. By setting t to correspond to the HVL and QVL for a particular spectrum, one can validate the material attenuation properties of an MC code system by comparing the simulated R_1 and R_2 to their theoretical values of 0.5 and 0.25, respectively. The simulation was performed for the monoenergetic energies of 30 keV and 100 keV, along with the polyenergetic spectrums of 30 kVp and 100 kVp, which were provided by TG-195. MIDSX's results for Case 1 agree to within 0.32% of the mean results published by TG-195. The results of Case 1 are presented in Figure 1.

For Case 2, a full-field and pencil beam x-ray source were directed towards a cuboid tissue phantom at 0° and 15° with respect to the z-axis. Directly behind and inside the phantom, a grid of square ROIs and cube VOIs were placed, respectively. The simulation was performed for the TG-195 provided polyenergetic spectrum of 120 kVp and its mean energy of 56.4 keV. For the 0° , full-field ROI measurements not presented in this paper, a $< 3\%$ mean percent error (MPE) is seen for MIDSX's results to each ROI simulation. Furthermore, for the 0° pencil-beam ROI measurements shown in Figure 3, a $< 2.1\%$ MPE is observed for each ROI simulation except for the case of a single incoherent scatter. In this particular case, MIDSX's results for ROI 4 and 5 are significantly lower, with the MPE reaching 10% for ROI 5. The full-field VOI energy deposition measurements depicted in Fig 2 show a minimal MPE of less than 0.1% for the 0° source. Conversely, for the MIDSX results at 15° , the MPE reaches an unexpectedly larger value of approximately 0.5%.

For Case 5, a fan beam was collimated to the center of a voxelized human torso phantom provided by TG-195. To replicate a CT image, the simulation was repeated for several angles along a circle surrounding the phantom. The simulation was performed with Case 2's 120 kVp energy spectrum, and energy deposition was measured in the different materials/organs composing the phantom. Almost all of MIDSX's results for the 0° source presented in Fig 4 are systematically lower than the mean of the reference code systems, with MPE's ranging from 1.1% to 6.3%. This pattern is disrupted by the thyroid, which is larger than the mean by 2.9%. To quantify the cumulative error, the root mean square percent error (RMSPE) was calculated using each organ

result, which resulted in the RMSPE for MIDSX being 5%.

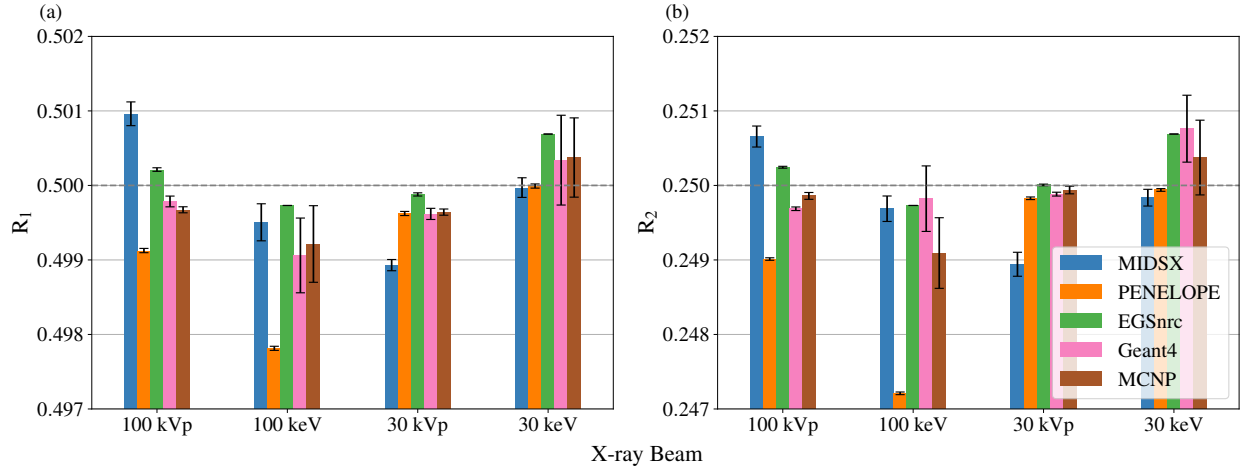


FIG. 1: Results for the (a) HVL and (b) QVL simulations as described by Case 1. The ratios of the primary HVL and QVL air kermas to the primary background air kermas is represented by R_1 and R_2 , respectively. The simulation was performed for the monoenergetic energies 30 keV and 100 keV, along with the polyenergetic spectrums of 30 kVp and 100 kVp, which were provided by TG-195. A dashed-line is placed at $R_1 = 0.5$ and $R_2 = 0.25$ to aid the reader in comparing the simulated results to their theoretical counterpart.

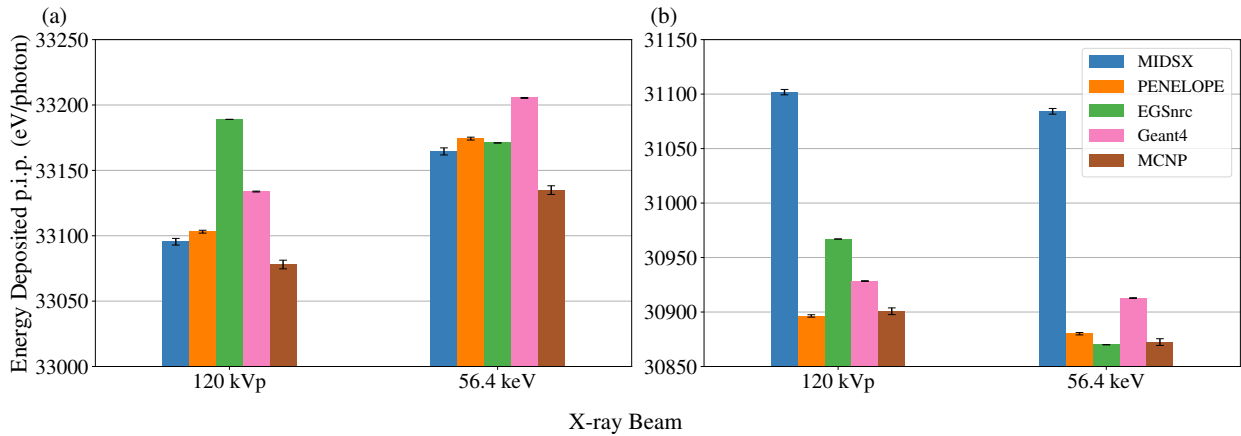


FIG. 2: The energy deposited per initial photon (eV/photon) in the simulated tissue for the full-field simulation as described by Case 2. The simulation was performed at 56.4 keV and 120 kVp at both (a) 0° and (b) 15° , with the 120 kVp spectrum provided by TG-195.

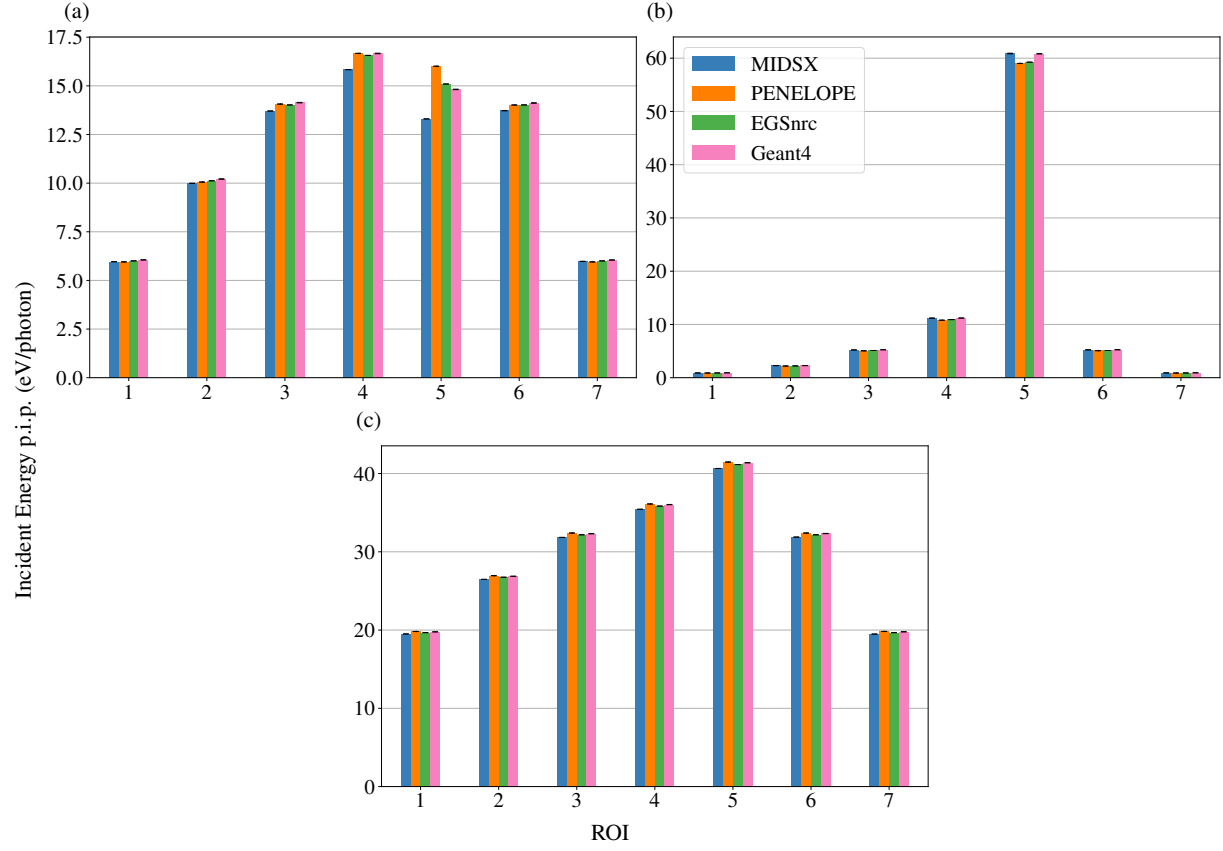


FIG. 3: The energy per initial photon (eV/photon) of photons incident upon each region of interest (ROI) for the 0° , pencil beam, 56.4 keV simulation as described by Case 2. The incident energy was determined separately for photons that underwent (a) a single incoherent scatter, (b) a single coherent scatter, (c) and multiple scatters.

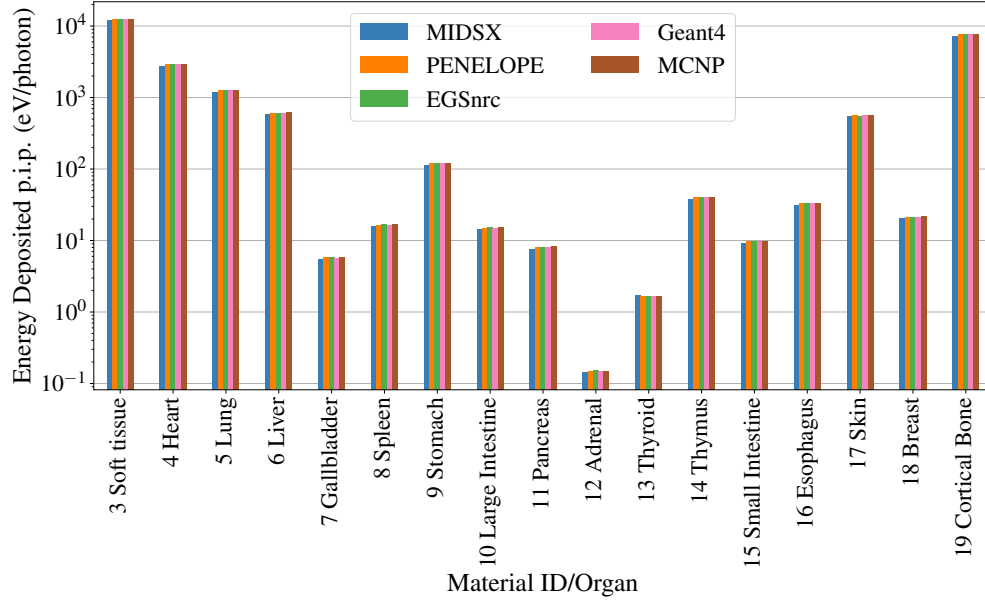


FIG. 4: The energy deposited per initial photon (eV/photon) in the material IDs/organs composing a voxelized human phantom for the 0° , 120 kVp simulation as described by Case 5.

V. DISCUSSION

Overall, MIDSX shows varied but reasonable agreement with the reference codes systems of TG-195. For Case 1, excellent agreement is observed, in effect validating the total and max cross-section data used in δ tracking, and the sampling of a discrete x-ray energy spectrum. This makes MIDSX a reliable and accurate option for primary particle measurements.

For the ROIs of Case 2, agreement is seen almost universally, except for the single incoherent scatter deposition energy in ROI 4 and 5 for the pencil beam source shown in Fig 3. In particular, the results reached a max MPE of 10% for ROI 5, indicating a potential error in the incoherent scattering energy/angular distribution sampling algorithm. This discrepancy is likely a result of an error in the rejection sampling algorithm employed by MIDSX. While this algorithm shows agreement for the full-field ROI 5, the geometry of the ROI, combined with the pencil beam, results in only narrow-angle scatters hitting the ROI. Since the scattering angle distribution of incoherent scattering at the medical imaging energy range becomes extremely steep at the scattering angle $\theta = 0^\circ$, there is likely numerical instability presenting itself in the algorithm that needs to be analyzed. In addition, the 15° full-field tissue energy deposition measurements were larger than the reference code systems', with an MPE of 0.5%. With the MPE increasing significantly from the 0° to 15° measurements, this hints at a possible geometric error with the source and/or body. However, the source's position and angular distribution, along with the domain's and tissue's dimensions very extensively verified, making the geometry of the scene less likely to be the source of the discrepancy.

For Case 5, almost all organ energy deposition results were lower than the reference code results, with the MPE reaching 6.3%. An exception was seen with the thyroid, which had a uniquely larger energy deposition with an MPE of 2.9%. One common error reported by TG-195 that could result in the observed discrepancies is the incorrect orientation of the voxelized phantom in the computational domain. On top of checking the scene's geometry and the .domain file, the orientation was verified by taking the RMSPE of the MIDSX data with respect to the results of each simulated angle reported by TG-195. As expected, the RMSPE with respect to 0° was the minimum, further solidifying the belief that the phantom's orientation during the CT simulation was correct.

However, despite the orientation being verified, the deviation of MIDSX's energy deposition results for both Cases 2 and 5 raises concerns. This suggests that there may be other underlying

issues in the MIDSX system that need further investigation. Potential factors could include the software’s handling of scattering events, cross-section data initialization, and interpolation. It’s imperative for future research to delve deeper into these aspects to pinpoint and rectify the source of the systematic errors observed in the MIDSX results.

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