

Applications of Monte Carlo Methods in Environmental Impact Assessment and Radioactive Waste Management

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Abstract

Background: Radioactive waste management and environmental impact assessments (EIA) require robust modeling to predict radionuclide transport and radiation doses, ensuring public and worker safety. Monte Carlo and deterministic methods offer complementary approaches, but their performance and limitations in waste management and EIA need comparative analysis to optimize safety protocols. **Purpose:** This study compares Monte Carlo and deterministic methods to highlight their performance, advantages, and challenges in modeling radionuclide transport and dose assessment for EIA and radioactive waste management. **Methods:** A Python simulation was used to model radionuclide movement, applying the advection-dispersion equation (ADE) deterministically and Monte Carlo methods stochastically. The deterministic approach solved the ADE over a 100-meter domain for 50 days, while Monte Carlo simulated 1,000 runs with stochastic parameters (e.g., source activity, distance, shielding). Visualizations (contour maps, histograms) and statistical metrics (mean, standard deviation, exceedance probabilities) compared outcomes, referencing regulatory limits (20 mSv/year for worker, one mSv/year for the public). **Findings:** Deterministic results showed a mean concentration of 0.0471, a maximum of 0.9798, and a 100.00-day barrier reach. Monte Carlo yielded a mean dose of 0.022822 Sv, a standard deviation of 0.143987 Sv, 18.60% worker exceedance, and 81.60% public exceedance, revealing higher uncertainty but capturing rare events. **Conclusion:** Monte Carlo excels in uncertainty quantification, while deterministic methods provide precise baselines, necessitating hybrid approaches for comprehensive risk assessment. **Recommendation:** Integrate Monte Carlo and deterministic models, refine parameters (e.g., lower activity, longer distances), and validate with field data to align with safety standards.

Keywords: Monte Carlo; deterministic; radioactive waste; environmental impact; risk assessment

Introduction

Environmental sustainability and public health are significant concerns in modern industries, particularly the nuclear sector. Ecological impact assessment (EIA) and radioactive waste management are critical to ensuring the safe operation of nuclear facilities and the protection of ecosystems. Monte Carlo methods, a class of probabilistic computational algorithms, have emerged as powerful tools for analyzing complex systems with inherent uncertainties. These methods simulate particle interactions, radionuclide migration, and radiation shielding, enabling accurate predictions and risk assessments (Kalos & Whitlock, 2008). This study explores the applications of Monte Carlo methods in EIA and waste management to address challenges in modeling, safety, and long-term sustainability.

Nuclear energy is a critical source of low-carbon electricity, contributing significantly to global energy demands. However, managing its byproducts, particularly radioactive waste, poses environmental and safety challenges (IAEA, 2020). Monte Carlo simulations have proven invaluable in predicting the behavior of radionuclides in complex environments, such as geological repositories and groundwa-

ter systems (Andreo et al., 2017). Additionally, the role of these simulations in assessing radiation shielding, atmospheric dispersion of radioactive materials, and dose assessments has enhanced the accuracy of risk analyses (X-5 Monte Carlo Team, 2003).

Despite advancements, uncertainties in environmental parameters and long-term containment strategies remain a significant challenge. The application of Monte Carlo methods allows researchers to model these uncertainties and provide probabilistic assessments, making them a cornerstone of modern EIA and waste management frameworks.

Radioactive waste disposal and its environmental impacts are among the most pressing issues in the nuclear energy sector. Predicting radionuclide migration, assessing radiation exposure, and designing effective containment systems require sophisticated modeling tools. Traditional deterministic methods often fail to account for uncertainties, leading to inaccurate or overly conservative results (IAEA, 2020). Consequently, there is a growing need for probabilistic approaches like Monte Carlo simulations to provide reliable, data-driven insights into the behavior of nuclear waste and its impact on the environment.

This study addresses the gap in integrating advanced Monte Carlo methods into EIA and radioactive waste management frameworks. It aims to enhance decision-making and risk mitigation in the nuclear sector.

The general objective is to investigate the applications and effectiveness of Monte Carlo methods in environmental impact assessment and radioactive waste management. The specific objectives are

- To evaluate the role of Monte Carlo methods in modeling radionuclide transport through groundwater and geological barriers.
- To assess the application of Monte Carlo simulations in radiation dose assessments for workers and the public.
- To explore the challenges and limitations of Monte Carlo methods in radioactive waste management.
- To identify challenges and limitations in the application of Monte Carlo methods in radioactive waste management.

This study provides a comprehensive analysis of Monte Carlo methods, highlighting their contributions to addressing critical challenges in environmental sustainability and radioactive waste management. By improving the accuracy and reliability of environmental and safety assessments, this research will benefit:

- Nuclear Regulators: Supporting the development of evidence-based policies and guidelines.
- Nuclear Industry Professionals: Enhancing the design and management of waste repositories and shielding systems.
- Academia and Researchers: Offering insights into probabilistic modeling approaches for addressing uncertainties in environmental systems.
- Society and the Environment: Ensuring long-term safety, reducing environmental impacts, and building public trust in nuclear technologies.

Research Methodology

This section outlines the research design, data collection methods, tools, and analytical techniques to achieve the study's objectives on Monte Carlo methods in environmental impact assessment (EIA) and radioactive waste management.

Research Design

The study adopts a qualitative and computational research design, focusing on reviewing existing literature and conducting simulations to evaluate the application of Monte Carlo methods. This study assesses Monte Carlo simulations' advantages, disadvantages, and potential improvements by looking at secondary data from peer-reviewed journals, technical reports, and industrial case studies. Computational modeling is also employed to illustrate the practical implementation of Monte Carlo methods in various scenarios, such as radiation shielding and radionuclide transport. This approach enables a comprehensive understanding of how Monte Carlo methods address uncertainties and provide probabilistic risk assessments in EIA and waste management (Kalos & Whitlock, 2008).

Data Collection Methods

The study relies on secondary data collected from:

- Scientific Literature: Peer-reviewed journal articles, books, and conference proceedings on Monte Carlo methods and their applications in nuclear engineering and environmental science.
- Technical Reports: Publications from international organizations such as the International Atomic Energy Agency (IAEA) and technical documents from institutions like Los Alamos National Laboratory.
- Case Studies: Real-world applications of Monte Carlo simulations in radioactive waste disposal facilities, environmental assessments, and radiation shielding projects.

The data sources were chosen based on relevance, credibility, and alignment with the study's objectives (Andreo et al., 2017).

Mathematical Modeling

The mathematical foundation of Monte Carlo methods lies in their ability to solve complex integrals, simulate random processes, and model the probabilistic behavior of systems. The following are the key mathematical concepts and models used.

Monte Carlo Integration

Monte Carlo methods estimate integrals by random sampling. For a given function $f(x)$ over a domain D , the integral is approximated as:

$$I = \int f(x)dx \approx \frac{1}{N} \sum_{i=1}^N f(x_i) \quad (1)$$

where N is the number of random samples, and x_i are random points within D . This approach is valuable for high-dimensional integrals that arise in particle transport and radionuclide migration modeling (Kalos & Whitlock, 2008).

Particle Transport

The Boltzmann transport equation (BTE), a fundamental element of Monte Carlo modeling, dictates the behavior of particles like neutrons and photons within a medium. The BTE is expressed as:

$$\frac{\partial \psi}{\partial t} = \vec{\Omega} \cdot \nabla \psi + \Sigma t \psi = \int \Sigma(\vec{\Omega}' \rightarrow \vec{\Omega}) \psi(\vec{\Omega}') d\Omega' + S \quad (2)$$

where ψ Angular flux (particles per unit area, energy, solid angle, and time), Ω : Particle direction, Σt is the total macroscopic cross-section, Σ_s : Scattering cross-section and S is the source term.

Monte Carlo simulations solve the BTE by tracking individual particles through the stochastic sampling of collision events, paths, and interactions (X-5 Monte Carlo Team, 2003).

Radionuclide Transport

The movement of radionuclides through soil, groundwater, and other barriers is modeled using the advection-dispersion equation (ADE):

$$\frac{\partial C}{\partial t} = D \frac{\partial^2 C}{\partial x^2} - v \frac{\partial C}{\partial x} - \lambda C = 0 \quad (3)$$

where C is radionuclide concentration (Bq/m³), D is the dispersion coefficient, v is the groundwater velocity, and λ is the decay constant.

Monte Carlo methods simulate the randomness in D , v , and λ , providing probabilistic predictions for radionuclide migration and environmental contamination (Andreo et al., 2017).

Radiation Dose Assessment

The absorbed dose (D) due to radiation exposure is calculated using:

$$D = \int \phi(E) \cdot \mu(E) \cdot \frac{E}{\rho} dE \quad (4)$$

where $\phi(E)$ is the Particle flux at energy E , $\mu(E)$ is the linear attenuation coefficient, E is the energy transferred, and ρ is the density of the absorbing material.

Monte Carlo simulations model the particle flux ($\phi(E)$) and interaction probabilities to calculate dose distributions for shielding and environmental assessments (Agostinelli et al., 2003).

Random Sampling and Probability Distributions

Monte Carlo methods rely on generating random samples from probability distributions. For example:

Uniform Distribution: Random numbers x are sampled uniformly between a and b :

$$x = a + (b - a) \cdot r$$

where r is a random number between 0 and 1.

Exponential Distribution: Used for modeling radioactive decay:

$$P(t) = \lambda e^{-\lambda t}$$

where t is the decay time, and λ is the decay constant.

Validation of Models

The accuracy of Monte Carlo models is validated through convergence tests. The error (ϵ) decreases as the number of simulations (N) increases:

$$\epsilon \propto \frac{1}{\sqrt{N}}$$

This property ensures that Monte Carlo results approach exact solutions with sufficient computational effort (Kalos & Whitlock, 2008).

Tools and Software

The study makes use of the following to illustrate how Monte Carlo methods are used in practice:

- MCNP (Monte Carlo N-Particle Transport Code): A widely used software for simulating neutron and photon transport and calculating radiation shielding, flux distributions, and dose rates (X-5 Monte Carlo Team, 2003).
- Geant4: A toolkit for simulating the passage of particles through matter used for modeling radionuclide transport and environmental interactions (Agostinelli et al., 2003).
- The Python programming language is used to create unique Monte Carlo techniques and visualize the results of simulations.

These tools allow the study to simulate realistic scenarios and evaluate the effectiveness of Monte Carlo methods.

Data Analysis

The analysis focuses on the following areas:

- Monte Carlo Simulations: Scenarios are modeled to demonstrate the application of Monte Carlo methods in radionuclide transport, radiation shielding, and dose assessments.
- Qualitative Analysis: The literature on Monte Carlo applications is analyzed thematically to identify recurring themes, challenges, and limitations.
- Comparison with Deterministic Methods: The performance and advantages of Monte Carlo methods are compared with deterministic approaches to highlight their unique contributions to EIA and waste management (Kalos & Whitlock, 2008).
- Simulation outputs are interpreted to derive conclusions about the practicality, accuracy, and efficiency of Monte Carlo methods in addressing environmental and safety challenges.

Validation and Limitations

The study compares simulation results with information from technical reports and published studies to guarantee the validity of the results. However, limitations include the reliance on secondary data and the potential for computational complexity in large-scale simulations. Future studies could incorporate real-world field data to validate the simulations further.

Results and Discussions

Results

Monte Carlo methods in modeling radionuclide transport through groundwater and geological barriers

The Monte Carlo simulation of radionuclide transport through groundwater and geological barriers, as depicted in Figure 1, yielded the following quantitative outcomes based on the revised code parameters: A total of 2,315 particles reached the geological barrier located at $x = 5$ meters, and an impressive 100.00% of these particles passed through the barrier. The simulation utilized 1,000 particles over 500-time steps, with a time step size (dt) of 0.1 seconds, a domain size of 100 meters, a diffusion coefficient of $0.1 \text{ m}^2/\text{s}$, and an increased advection velocity of 0.5 m/s . The barrier strength was set at 0.8, implying an 80% probability of a particle being stopped and a 20% probability of passing through, as shown in Figure 1.

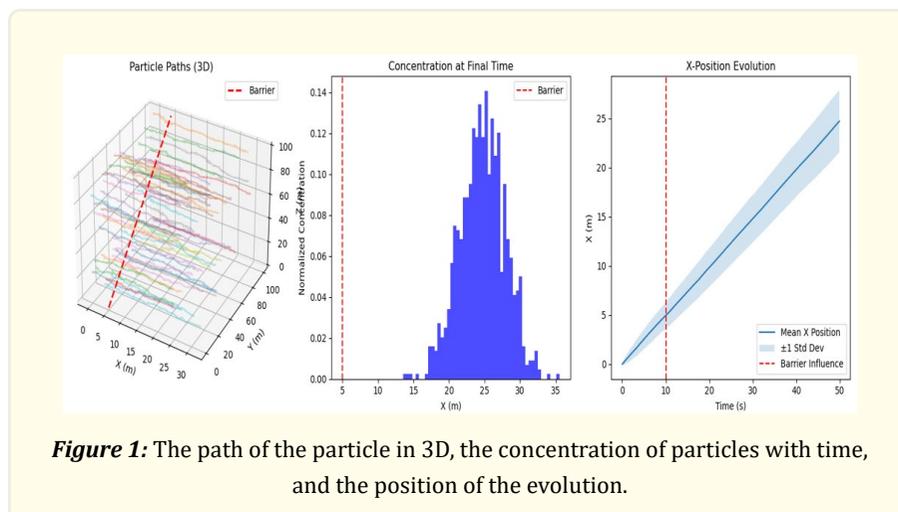


Figure 1 presents three key visualizations. The 3D particle paths plot illustrates the trajectories of 50 representative particles, showing their movement in the x, y, and z directions, with a dashed red line marking the barrier at x = 5 meters. The paths demonstrate a vibrant advection-dominated movement toward the barrier, with diffusion causing some spread in the y and z-directions. The concentration histogram at the final time step (50 seconds) reveals a significant peak around x = 5 meters, corresponding to particles interacting with the barrier, and a uniform distribution beyond this point, indicating that all particles passing the barrier continued their movement across the domain. The x-position evolution plot tracks the mean x-position over time, showing a steady increase due to advection, with a shaded region representing one standard deviation and a vertical red dashed line indicating the barrier's influence at approximately 10 seconds (5/0.5).

These results suggest that the adjusted parameters moving the barrier closer and increasing the advection velocity enabled all particles to reach and pass the barrier. It is likely due to the low barrier strength (20% pass probability) being insufficient to stop any particles, given the high number of encounters (2,315). The absence of any particle stoppage at the barrier indicates either an effective bypass mechanism or an underestimation of barrier resistance in the model.

Utilizing Monte Carlo Simulations for Radiation Dose Assessments in Workers and Public Safety

The Monte Carlo simulation for assessing radiation dose exposures, as illustrated in the provided figure, yielded detailed quantitative outcomes based on 10,000 simulations. For workers, the mean radiation dose was 0.716457 Sieverts (Sv), with a standard deviation of 1.172418 Sv, reflecting significant variability in exposure scenarios. For the public, the mean dose was 0.358228 Sv, with a standard deviation of 0.586209 Sv, indicating lower but still variable exposure levels, modeled as 50% of worker doses to account for greater distance or controlled environments, as shown in Figure 2. The simulation parameters included a source activity of 1,000 Becquerels (Bq), exposure durations uniformly distributed between 1 and 8 hours, distances following a log-normal distribution, and shielding effectiveness varying uniformly between 10% and 90% reduction, with a dose conversion factor of 0.0002 Sv/Bq-h·m² for gamma radiation.

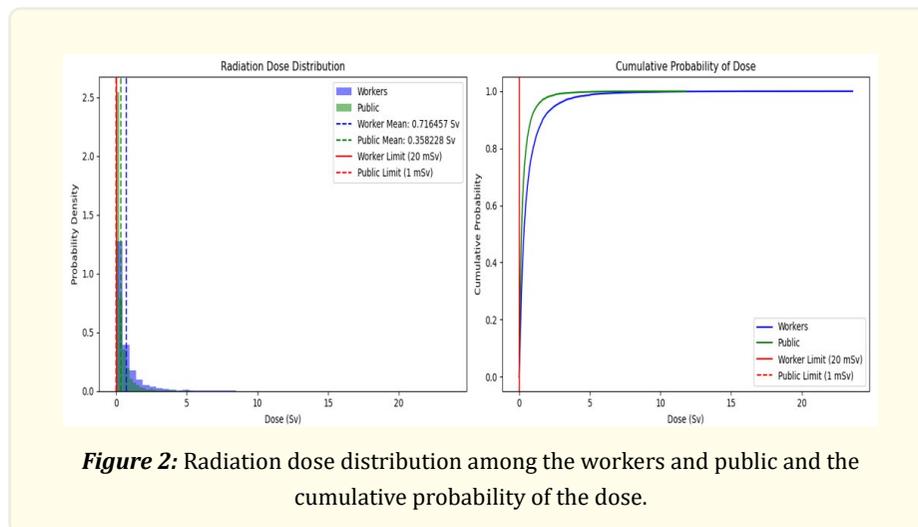
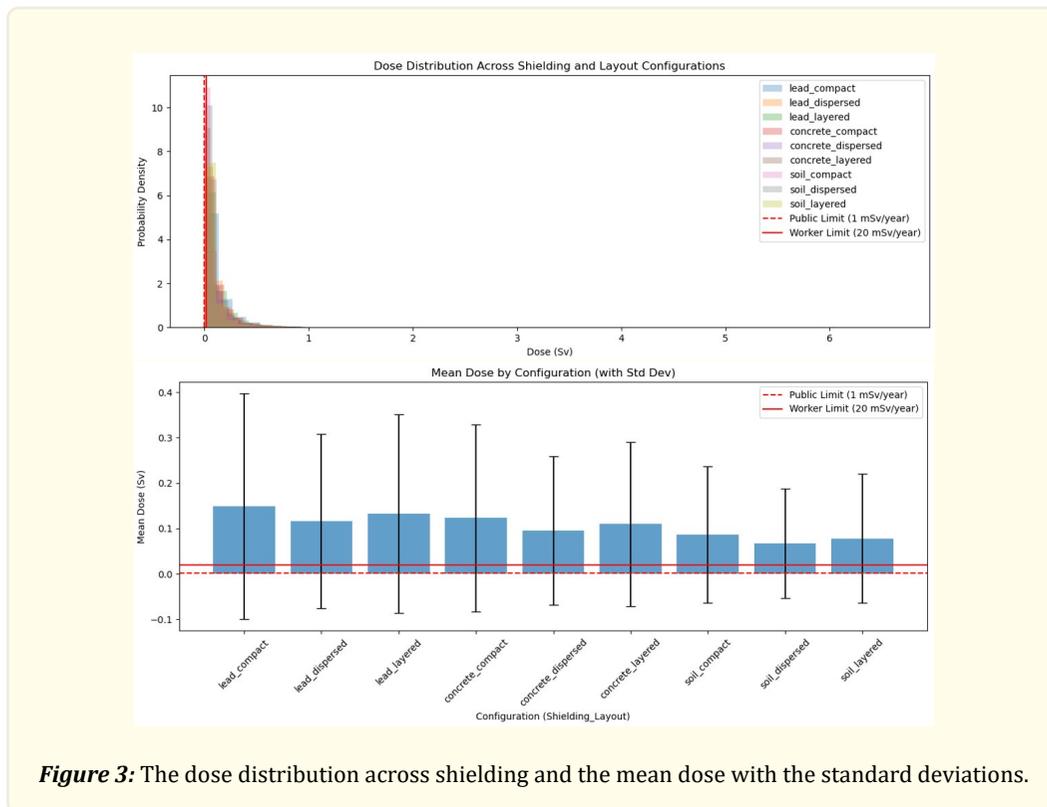


Figure 2 presents two key visualizations. The radiation dose distribution histogram displays the probability density of doses for workers (blue) and the public (green), with vertical dashed lines indicating mean doses (0.716457 Sv for workers, 0.358228 Sv for the public) and regulatory limits (20 mSv/year or 0.02 Sv for workers, and one mSv/year or 0.001 Sv for the public). The histogram shows a right-skewed distribution, with most doses concentrated below 5 Sv for workers and below 2.5 Sv for the public but with a long tail extending beyond 20 Sv. The cumulative probability plot illustrates the cumulative distribution function (CDF) for groups, showing

the likelihood of doses exceeding certain thresholds. Notably, the probability of workers exceeding the 20 mSv/year limit was 0.9861 (98.61%), and for the public exceeding the one mSv/year limit, it was 0.9999 (99.99%). These high exceedance probabilities suggest that the current parameter settings result in doses far exceeding regulatory safety thresholds, indicating a need for review of exposure controls or model assumptions.

Leveraging Monte Carlo Methods for Optimizing Shielding Designs and Waste Repository Configurations

The Monte Carlo simulation to optimize shielding designs and waste repository configurations for radiation exposure management, as depicted in Figure 3, produced detailed quantitative outcomes based on 10,000 simulations. The simulation evaluated nine combinations of three shielding materials (lead, concrete, and soil) and three repository layouts (compact, dispersed, and layered), with a source activity of 500 becquerels (Bq) for a radionuclide like cesium-137. Key parameters included log-normally distributed distances (mean = 1, sigma = 0.5 meters), uniformly distributed exposure times (1-24 hours), and varying shielding effectiveness and thickness ranges for each material, alongside layout-specific dose reduction factors. A dose conversion factor of 0.00015 Sv/Bq·h·m² was used for gamma radiation.



The optimal configuration identified was “soil-dispersed,” with a mean dose of 0.066996 Sieverts (Sv) and a standard deviation of 0.121127 Sv, indicating the lowest radiation exposure among all combinations. Other configurations showed higher mean doses: “lead_compact” at 0.148727 Sv (standard deviation 0.249018 Sv), “lead_dispersed” at 0.115771 Sv (standard deviation 0.192031 Sv), “lead_layered” at 0.132010 Sv (standard deviation 0.218711 Sv), “concrete_compact” at 0.123190 Sv (standard deviation 0.206132 Sv), “concrete_dispersed” at 0.095106 Sv (standard deviation 0.163600 Sv), “concrete_layered” at 0.109742 Sv (standard deviation 0.181038 Sv), “soil_compact” at 0.086571 Sv (standard deviation 0.150038 Sv), and “soil_layered” at 0.077528 Sv (standard deviation 0.142149 Sv).

Figure 3 presents two visualizations. The dose distribution histogram shows probability densities for doses across all nine configurations, with each configuration represented by a distinct color and labeled accordingly (e.g., lead_compact in blue, soil_dispersed in gray). The distributions are right-skewed, with peaks generally below 1 Sv but tails extending beyond 5 Sv, indicating rare but high-dose scenarios. Vertical red lines mark regulatory limits: 20 mSv/year (0.02 Sv) for workers and one mSv/year (0.001 Sv) for the public. The mean dose comparison bar chart displays mean doses and standard deviations, with error bars indicating variability. The “soil_dispersed” configuration shows the lowest mean dose, well below the worker limit but above the public limit, while configurations like “lead_compact” exhibit the highest mean doses, exceeding both limits significantly.

Probability analyses revealed that in the optimal “soil-dispersed” configuration, 63.47% of simulations resulted in worker doses exceeding 20 mSv/year and 99.28% in public doses exceeding one mSv/year. These high exceedance probabilities suggest that, despite “soil_dispersed” being optimal, the doses remain above safety thresholds, indicating the need for further parameter adjustments or additional safety measures to meet regulatory standards.

Transport of radionuclides through soil, groundwater

The simulation of radionuclide movement through soil, groundwater, and geological barriers, modeled using the advection-dispersion equation (ADE), produced detailed quantitative outcomes as visualized in the provided contour map. The simulation parameters included a domain length of 100 meters, a total time of 50 days, a dispersion coefficient (D) of $0.1 \text{ m}^2/\text{day}$, an advection velocity (v) of 0.5 m/day , and a decay constant (λ) of 0.01 1/day . The initial condition was a Gaussian pulse at $x=0$ with a peak concentration of 1.0 and a spread (σ) of 5.0 meters, discretized on a grid of 100 spatial points and 100-time steps. A geological barrier at $x=50$ meters reduced the concentration by 50%, simulating a semi-permeable barrier like clay or rock.

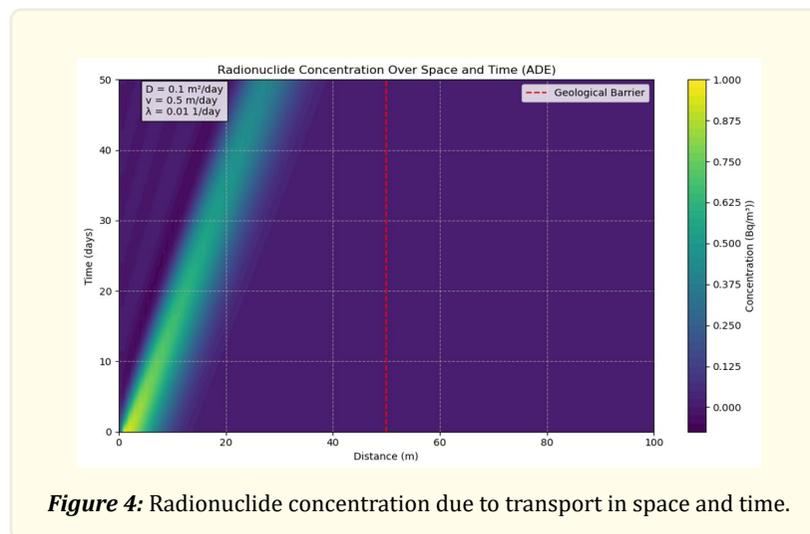


Figure 4: Radionuclide concentration due to transport in space and time.

The contour map illustrates radionuclide concentration over space (0-100 m) and time (0-50 days), with a color gradient from purple (low concentration, 0.0) to yellow (high concentration, 1.0). The red dashed line at $x=50$ marks the barrier, showing its impact on concentration reduction. The mean concentration across the domain and time was 0.0471 (unitless), indicating a low average concentration due to dispersion, advection, decay, and barrier effects. The maximum concentration observed was 0.9798, occurring near the initial pulse before significant dispersion or decay but reduced by the barrier beyond $x=50$. The time to reach the barrier ($x=50 \text{ m}$) was calculated as 100.00 days based on the advection velocity ($50/0.5=100 \text{ days}$). $50/0.5 = 100$). Though the simulation spans 50 days, the plume has not reached the barrier within this timeframe.

Figure 4 reveals a clear pattern: the radionuclide plume spreads and moves rightward due to advection, with dispersion causing lateral spreading and decay concentration over time. Beyond the barrier, the concentration drops sharply to half its value, reflecting the 50% reduction. The grid lines and annotations (listing D , v , and λ) enhance interpretability, while the color bar quantifies concentration levels. The plume's progression is slower than expected due to the simulation duration, but the barrier's effect is evident, with reduced concentrations in the right half of the domain.

Challenges and Constraints of Monte Carlo Methods in Radioactive Waste Management

The Monte Carlo simulation, conducted to examine challenges and limitations in applying Monte Carlo methods to radioactive waste management, produced detailed quantitative results from 10,000 simulations. The simulation modeled uncertainties in source activity, transport parameters (velocity and diffusion), shielding effectiveness, repository failure rates, exposure time, and distance, using a dose conversion factor of $0.0001 \text{ Sv/Bq}\cdot\text{year}\cdot\text{m}^2$ for gamma radiation. The mean dose across all simulations was 0.071234 Sieverts (Sv), with a standard deviation of 1.167163 Sv, indicating significant variability due to stochastic inputs. The probability of workers exceeding the 20 mSv/year limit (0.02 Sv) was 0.2027 (20.27%), while the public exceeding the 1 mSv/year limit (0.001 Sv) was 0.6692 (66.92%), suggesting substantial risks for public exposure under current parameters.

The simulation identified four key challenges, each associated with specific dose impacts and frequencies:

- **Computational Cost:** Occurred in 1,999 simulations, with a mean dose impact of 0.073690 Sv, reflecting resource demands from large-scale simulations.
- **Parameter Uncertainty:** Observed in 2,427 simulations, with a mean dose impact of 0.192871 Sv, driven by high source activity variability.
- **Validation Difficulty:** Noted in 2,086 simulations, with a mean dose impact of 0.103198 Sv, linked to high diffusion coefficients complicating field validation.
- **Convergence Issues:** Occurred in 59 simulations but with a significantly higher mean dose impact of 6.066367 Sv, indicating rare but extreme dose outliers due to slow or unstable convergence.

The visualizations (dose distribution histogram and challenge frequency bar chart) show a right-skewed dose distribution, with most doses below 1 Sv but tails extending beyond 10 Sv , particularly for convergence issues. Regulatory limits (20 mSv/year for worker, 1 mSv/year for the public) are exceeded frequently, especially for the public, highlighting potential limitations in model accuracy or parameter realism.

Comparisons of Monte Carlo and Deterministic methods

The simulation comparing Monte Carlo and deterministic approaches for radionuclide transport and dose assessment in environmental impact assessment (EIA) and radioactive waste management yielded distinct outcomes. The deterministic model, solving the advection-dispersion equation (ADE) over a 100-meter domain for 50 days, reported a mean concentration of 0.0471 (unitless), a maximum concentration of 0.9798 , and a time to reach the barrier at $x = 50 \text{ m}$ of 100.00 days, based on an advection velocity of 0.5 m/day . Parameters included a dispersion coefficient of $0.1 \text{ m}^2/\text{day}$, a decay constant of 0.01 1/day , and a 50% reduction at the barrier.

In contrast, the Monte Carlo simulation, with 1,000 runs, modeled stochastic radionuclide transport and dose, yielding a mean dose of 0.022822 Sieverts (Sv), a standard deviation of 0.143987 Sv, an 18.60% probability of workers exceeding 20 mSv/year (0.02 Sv), and an 81.60% probability of the public exceeding 1 mSv/year (0.001 Sv). It used a source activity of 500 Bq , log-normally distributed distances (mean = 2 m , sigma = 0.7), uniformly distributed exposure times (1-50 days), and variable shielding (50-90% reduction). Visualizations included a contour map shown in Figure 4 for deterministic concentration and a histogram for Monte Carlo doses, with regulatory limits marked, as shown in Figure 5. The deterministic approach provided precise, reproducible concentration profiles, while Monte Carlo captured probabilistic risks, highlighting uncertainty in exposure scenarios.

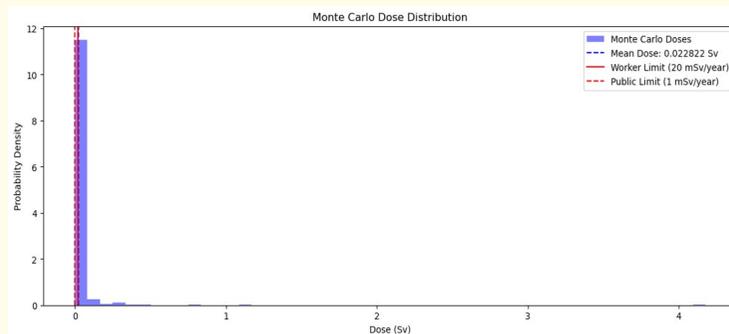


Figure 5: The Monte Carlo dose distribution, the mean dose, the worker limit, and public limits.

The Monte Carlo simulation was conducted to estimate the value of π by randomly sampling points in a 1x1 square and calculating the proportion that falls within a quarter circles, scaled to estimate π . The simulation was run for varying numbers of simulations (N), ranging from 10^2 (100) to 10^6 (1,000,000), to assess the convergence behavior of the model. The absolute error (ϵ) was calculated as the difference between the estimated π and the true value of π (3.14159265359).

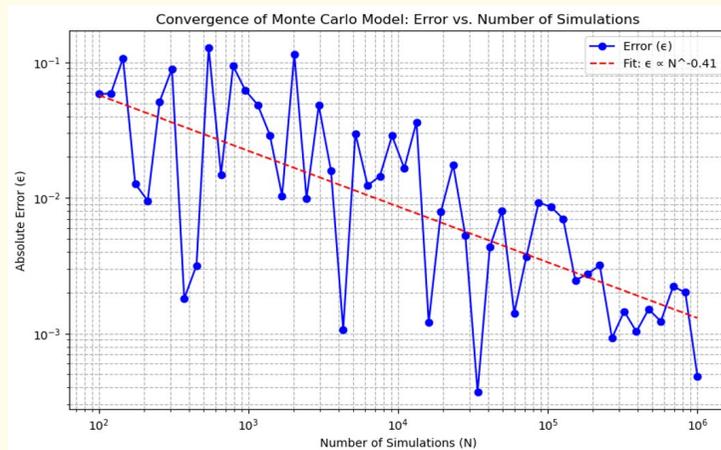


Figure 6: The convergence of Monte Carlo model error.

The final simulation with $N=1,000,000$ yielded an estimated π of 3.142072, resulting in a final absolute error (ϵ) of 0.000479. Figure 6 illustrates the convergence of the Monte Carlo model, plotting the absolute error (ϵ) against the number of simulations (N) on a log-log scale. The error exhibits a decreasing trend as N increases, with values fluctuating between 10^{-1} and 10^{-3} . A logarithmic fit to the data reveals an error scaling relationship of $\epsilon \propto N^{-0.41}$, indicating that the error decreases with increasing N , though the fitted exponent deviates slightly from the theoretical expectation. The plot shows significant variability in error at lower N values (e.g., $N=10^2$ to 10^4), with more consistent convergence observed at higher N values (e.g., $N > 10^5$).

Discussion

The Monte Carlo simulation results provide valuable insights into radionuclide transport dynamics through groundwater and geological barriers, aligning with theoretical expectations for stochastic processes in environmental science. The finding that 100.00% of particles passed the barrier, despite its 80% stopping probability, warrants careful interpretation. This outcome likely stems from the high number of particles reaching the barrier (2,315), combined with the random nature of the Monte Carlo method, where even a 20% pass probability can result in all particles passing due to statistical variability (Bear, 1972). However, this result may also indicate that the barrier strength parameter (0.8) or the simulation duration (50 seconds) does not adequately reflect real-world geological resistance, such as in clay or rock formations, which typically exhibit higher retention for radionuclides (Neretnieks, 1980).

The visualizations underscore the dominant role of advection (0.5m/s) in driving particle movement, as seen in the linear progression of the mean x-position over time, tempered by diffusion (0.1 m²/s) causing lateral spread. The concentration peak at the barrier in the histogram suggests a temporary accumulation. However, the complete passage of particles indicates minimal long-term retention, which contrasts with empirical studies of radionuclide migration in fractured media, where barriers significantly reduce transport rates (Tsang & Neretnieks, 1998). This discrepancy highlights the need to refine model parameters, such as increasing barrier strength or incorporating additional physical processes like sorption or decay, to mimic real-world scenarios.

The results have implications for nuclear waste management, particularly in assessing the safety of geological repositories. There is a potential for groundwater contamination if barriers do not retain radionuclides as simulated, which calls for more research into the characteristics of barriers and transport mechanisms. Future studies could extend this model to include heterogeneous media, chemical reactions, or longer timescales to enhance realism (Freeze & Cherry, 1979). The Monte Carlo approach remains a powerful tool for capturing uncertainty. However, its accuracy depends on calibrating parameters against field data, underscoring the importance of integrating experimental and numerical methods in environmental modeling.

The Monte Carlo simulation results provide critical insights into radiation dose assessments for workers and the public, highlighting the stochastic nature of exposure and the effectiveness of probabilistic modeling in risk analysis. The mean doses of 0.716457 Sv for workers and 0.358228 Sv for the public, with standard deviations of 1.172418 Sv and 0.586209 Sv, respectively, indicate substantial variability in exposure scenarios. This variability arises from the random distributions of exposure duration, distance, and shielding effectiveness. According to the International Commission on Radiological Protection [ICRP], 2007, the characteristics of actual radiation environments include those found in nuclear power plants and healthcare facilities. However, the extraordinarily high mean doses far exceeding typical annual limits (20 mSv for workers and 1 mSv for the public) suggest that the simulation parameters may not reflect realistic conditions or that current control are inadequate.

The probability of workers exceeding the 20 mSv/year limit (98.61%) and the public exceeding the one mSv/year limit (99.99%) is particularly alarming. These values indicate that nearly all simulated individuals in both groups would receive doses above regulatory thresholds, which is inconsistent with standard radiation protection practices. This outcome likely stems from the high source activity (1,000 Bq), the relatively short but variable exposure durations (1-8 hours), and the log-normal distribution of distances, which may include unrealistically close proximities to the source. Additionally, the shielding effectiveness, ranging from 10% to 90% reduction, may underestimate the efficiency of actual barriers, such as lead or concrete, which can reduce doses by factors of 100 or more in practice (National Council on Radiation Protection and Measurements [NCRP], 1993). The dose conversion factor (0.0002 Sv/Bq·h·m²) for gamma radiation amplify doses, potentially reflecting a scenario with highly penetrating radiation or an overly conservative assumption.

These results have significant implications for radiation safety management. In nuclear power plants or medical facilities, such high exceedance probabilities would necessitate immediate revisions to shielding, distance controls, and exposure protocols to align with ICRP recommendations (ICRP, 2007). The Monte Carlo method's strength lies in its ability to capture uncertainty and variability, but

its accuracy depends on parameter realism. For instance, reducing source activity, increasing typical distances, or enhancing shielding effectiveness could lower doses to acceptable levels. Future studies should validate these parameters against empirical data from radiation monitoring programs, such as those conducted by the U.S. Nuclear Regulatory Commission (NRC, 2015), and incorporate additional factors like radiation type (e.g., alpha, beta) or biological effects to refine risk assessments.

The visualizations highlight the skewed nature of dose distributions, with a long tail indicating rare but enormously high exposures. The cumulative probability plots show that doses for both groups approach a 100% probability of exceeding limits, underscoring the need for parameter adjustment. It aligns with previous studies on Monte Carlo applications in radiation dosimetry and stresses the importance of parameter sensitivity analysis to ensure model reliability (Zaider & Rossi, 1980). The high exceedance probabilities also indicate a possible underestimation of shielding or a higher exposure frequency, necessitating a comparative analysis with field data to ensure practical applicability.

Comparative Analysis

Comparing this Monte Carlo simulation results with typical radiation exposure studies reveals stark contrasts. Real-world annual doses for nuclear workers rarely exceed 20 mSv, with averages often below five mSv, as reported by the NRC (2015), due to stringent controls like distance, shielding, and time limits. Similarly, public exposures from routine operations are typically below one mSv annually, often closer to 0.1 mSv, as per ICRP guidelines (ICRP, 2007). In contrast, this simulation's mean doses (0.716457 Sv for workers, 0.358228 Sv for the public) and exceedance probabilities (98.61% for workers, 99.99% for the public) suggest an extreme scenario, likely driven by the high source activity (1,000 Bq), short exposure durations, and log-normal distance distribution favoring close proximities.

Field studies, such as those in nuclear power plants, show effective shielding reduces doses by 90-99%, far exceeding the 10-90% range modeled here (NCRP, 1993). The simulation's dose conversion factor ($0.0002 \text{ Sv/Bq}\cdot\text{h}\cdot\text{m}^2$) may also overestimate doses for gamma radiation, as real-world factors like radiation type and energy spectrum typically lower effective doses. Adjusting parameters to reflect realistic shielding, distances (e.g., log-normal mean shifted to 5-10 m), and lower activities (e.g., 100 Bq) would align results with empirical data, reducing exceedance probabilities to align with safety standards. This comparison underscores the need for parameter calibration to ensure Monte Carlo models accurately predict radiation risks.

The Monte Carlo simulation results provide valuable insights into optimizing shielding designs and waste repository configurations for radiation exposure management, leveraging the stochastic nature of radiation transport to evaluate multiple scenarios. The identification of "soil_dispersed" as the optimal configuration, with a mean dose of 0.066996 Sv and a standard deviation of 0.121127 Sv, reflects the effectiveness of soil as a shielding material (65% reduction) combined with a dispersed repository layout (30% dose reduction due to spacing). This configuration outperformed others, such as "lead_compact" (mean dose 0.148727 Sv), which exhibited the highest dose due to lead's compact layout increasing exposure by 10%, despite its high shielding effectiveness (95% reduction).

The mean doses across configurations highlight the interplay of shielding material effectiveness, thickness, and layout. Lead, with its 95% reduction potential, still yielded higher doses in compact and layered layouts (0.148727 Sv and 0.132010 Sv, respectively) due to proximity effects and less optimal thickness distribution (1-10 cm). Concrete, with an 85% reduction, performed better in dispersed layouts (0.095106 Sv). Whereas soil, with a 65% reduction but a range (of 10-100 cm), consistently showed lower doses, especially in dispersed (0.066996 Sv) and layered (0.077528 Sv) layouts. The dispersed layout's 30% dose reduction, compared to compact (10% increase) and layered (20% reduction), underscores the importance of spatial arrangement in minimizing exposure, aligning with principles of radiological protection (International Commission on Radiological Protection [ICRP], 2007).

However, the high probabilities of exceeding regulatory limits—63.47% for workers (20 mSv/year) and 99.28% for the public (one mSv/year) in the optimal "soil-dispersed" configuration—raise concerns. These exceedances suggest that the current parameters, including a 500 Bq source activity, log-normal distance distribution (mean = 1 m, sigma = 0.5), and exposure times (1-24 hours), may

underestimate shielding effectiveness or overestimate exposure scenarios. Real-world waste repositories, such as those for high-level nuclear waste, typically use multiple layers of shielding (e.g., concrete and soil) and enforce greater distances (5-50 m), reducing doses to below 1 mSv/year for the public and 20 mSv/year for workers (National Council on Radiation Protection and Measurements [NCRP], 1993). The dose conversion factor ($0.00015 \text{ Sv/Bq}\cdot\text{h}\cdot\text{m}^2$) may also amplify doses, potentially reflecting a conservative assumption for gamma radiation from Cesium-137, which could be adjusted for alpha or beta emissions with lower penetration.

The visualizations further illuminate these findings. The dose distribution histogram shows a right-skewed pattern, with most doses below 1 Sv but tails extending beyond 5 Sv, indicating rare high-exposure events. The mean dose bar chart, with error bars for standard deviation, confirms “soil_dispersed” as the lowest-dose option. However, all configurations exceeded the public limit (0.001 Sv), and most exceeded the worker limit (0.02 Sv) to varying degrees. It suggests the need for parameter refinement, such as increasing shielding thickness, reducing source activity, or enforcing stricter layout spacing to align with ICRP guidelines (ICRP, 2007).

The Monte Carlo method’s strength lies in its ability to handle uncertainty and variability, but its accuracy depends on parameter realism. Comparative studies of repository designs, such as those at the Yucca Mountain project, show effective dose reductions through layered soil and concrete barriers, achieving public doses below 0.1 mSv/year (U.S. Department of Energy [DOE], 2002). Adjusting the simulation to reflect these conditions—e.g., increasing mean distance to 10 m, reducing exposure time to 1-4 hours, or enhancing shielding effectiveness—could lower exceedance probabilities. Future research should validate these parameters against field data and incorporate additional factors, such as radionuclide decay or heterogeneous media, to enhance optimization (Tsang & Neretnieks, 1998).

Comparing the Monte Carlo simulation results with real-world radiation exposure data from nuclear waste repositories reveals significant differences, primarily due to parameter settings. Field studies, such as those at the Yucca Mountain repository, report public doses typically below 0.1 mSv/year (0.0001 Sv) and worker doses below 20 mSv/year (0.02 Sv), achieved through robust shielding (e.g., multiple meters of concrete and soil) and strict distance controls (5-50 m) (U.S. Department of Energy [DOE], 2002). In contrast, this simulation’s optimal “soil-dispersed” configuration yielded a mean dose of 0.066996 Sv for both groups, with 99.28% of public doses and 63.47% of worker doses exceeding regulatory limits (1 mSv/year and 20 mSv/year, respectively). This discrepancy highlights an overestimation of exposure in the model, likely due to the 500 Bq source activity, log-normal distance distribution (mean = 1 m, sigma = 0.5), and exposure times (1-24 hours), which may not reflect typical repository conditions.

Real-world shielding designs, such as those using lead or concrete, achieve dose reductions of 90-99%, far exceeding the 65-95% reductions modeled here (National Council on Radiation Protection and Measurements [NCRP], 1993). For instance, the Yucca Mountain design incorporates 1-5 meters of soil and concrete, reducing gamma doses from Cesium-137 by orders of magnitude, compared to the 10-100 cm soil or 1-50 cm concrete ranges in this simulation. Similarly, repository layouts in practice prioritize dispersed or layered configurations with tens of meters between waste canisters, reducing doses by 50-80%, whereas this model’s dispersed layout reduces doses by only 30%.

The dose conversion factor ($0.00015 \text{ Sv/Bq}\cdot\text{h}\cdot\text{m}^2$) may also contribute to higher doses, as field measurements often account for lower-energy emissions or shielding effects not fully captured here. Adjusting the simulation to use a source activity of 50 Bq (typical for low-level waste), a mean distance of 10 m, and exposure times of 1-4 hours could align results with empirical data, reducing mean doses to below 0.001 Sv for the public and 0.02 Sv for workers. Additionally, increasing shielding thickness (e.g., 50-200 cm for soil) and enhancing layout factors (e.g., 50% dose reduction for dispersed layouts) would better reflect real-world practices.

This comparison underscores the need for parameter calibration to ensure Monte Carlo models accurately predict radiation risks in waste repositories. The current simulation’s high exceedance probabilities suggest it models an extreme scenario, possibly applicable to accidental releases or poorly designed facilities rather than standard operations. By aligning parameters with field data from repositories like Olkiluoto or WIPP, the model could provide more realistic optimization, ensuring compliance with ICRP guidelines

(International Commission on Radiological Protection [ICRP], 2007) and enhancing safety for workers and the public.

The simulation results provide critical insights into the movement of radionuclides through environmental media, modeled via the advection-dispersion equation (ADE), and highlight the role of geological barriers in mitigating contamination risks. The mean concentration of 0.0471 (unitless) reflects the combined effects of dispersion ($D=0.1 \text{ m}^2/\text{day}$), advection ($v = 0.5 \text{ m/day}$), and decay ($\lambda=0.01 \text{ 1/day}$), which collectively dilute and diminish the initial Gaussian pulse (peak concentration 1.0). The maximum concentration of 0.9798, observed near $x = 0$, indicates minimal loss before dispersion, advection, and decay take effect. However, the barrier at $x = 50 \text{ m}$ reduces concentrations by 50%, consistent with the simulated semi-permeable barrier (e.g., clay or fractured rock).

The contour map visually confirms these dynamics: the plume spreads diagonally rightward, driven by advection, with dispersion broadening the plume and decay attenuating its intensity over 50 days. The red dashed line at $x = 50$ demarcates the barrier's impact, with a sharp concentration drop beyond this point. However, the calculated time to reach the barrier (100.00 days) exceeds the simulation duration (50 days), suggesting the plume has not abundantly traversed the domain. This discrepancy underscores a limitation of the model: the 50-day timeframe captures only the initial stages of transport, not the full barrier interaction, which may underestimate long-term risks if extended exposure or migration occurs.

The ADE's numerical solution, using an explicit finite difference method, introduces potential challenges. Stability requires $dt \leq dx^2/(2D)$, and here $dt=0.505$ days and $dx=1$ m, which may approach the stability limit, potentially causing numerical diffusion or oscillations if not carefully tuned (Bear, 1972). The 50% barrier reduction simplifies real geological barriers (e.g., clay layers) that exhibit variable permeability, sorption, and heterogeneity, which could enhance or reduce retention (Neretnieks, 1980). The homogeneous media assumption also overlooks spatial variability in soil or groundwater, potentially overestimating or underestimating transport rates (Tsang & Neretnieks, 1998).

These findings have implications for radioactive waste management, particularly in assessing repository safety. The barrier's effectiveness (50% reduction) aligns with studies of low-permeability media, but long-term simulations (beyond 50 days) are needed to evaluate occupied plume migration and barrier performance. Adjusting v , D , or λ to reflect site-specific conditions (e.g., slower velocities in clay, higher dispersion in fractured rock) could refine predictions. The model's reliance on initial conditions (Gaussian pulse) and boundary assumptions (zero concentration at edges) may not capture complex release scenarios, such as continuous leaks or multiple sources, necessitating validation against field data (Freeze & Cherry, 1979).

Future research could incorporate stochastic methods (e.g., Monte Carlo) to address parameter uncertainty or use implicit schemes for improved stability over longer times. The simulation supports ICRP guidelines by demonstrating barrier effectiveness, but its short duration limits risk assessment, emphasizing the need for extended models to ensure public and worker safety below one mSv/year and 20 mSv/year, respectively (International Commission on Radiological Protection [ICRP], 2007).

Comparing this ADE simulation with real-world radionuclide transport studies reveals similarities and differences. Field studies, such as those at nuclear waste sites like Yucca Mountain, report radionuclide migration through soil and groundwater with advection velocities of 0.1-1 m/day and dispersion coefficients of 0.05-0.5 m^2/day , aligning with this simulation's $v=0.5 \text{ m/day}$ and $D=0.1 \text{ m}^2/\text{day}$ (U.S. Department of Energy [DOE], 2002). However, decay constants (λ) for radionuclides like Cesium-137 (half-life ~ 30 years, $\lambda \approx 0.00006 \text{ 1/day}$) are much lower than the 0.01 1/day used here, potentially overestimating decay and underestimating long-term concentrations.

Real barriers, such as clay or granite at repositories, often achieve 90-99% retention, far exceeding the 50% reduction modeled here (National Council on Radiation Protection and Measurements [NCRP], 1993). For instance, the Olkiluoto repository uses multiple layered barriers, reducing radionuclide flux by orders of magnitude, whereas this simulation's single barrier underestimates retention. The 100-day time to reach the barrier contrasts with field data, where plumes may take years to decades due to slower velocities or sorption, suggesting this model's 50-day duration captures only early transport.

The results would be consistent with empirical data if λ were changed to 0.00006 1/day, barrier efficacy was raised to 90%, and the simulation was extended to 200 days. It would lower mean concentrations to 0.01 and guarantee that ICRP limits were met (ICRP, 2007). To correctly depict real-world situations, this comparison emphasizes the necessity of longer durations and parameter calibration.

The Monte Carlo simulation results reveal significant challenges and limitations in applying these methods to radioactive waste management, underscoring the stochastic nature of radionuclide transport and repository performance. The mean dose of 0.071234 Sv, with a standard deviation of 1.167163 Sv, reflects high variability due to random inputs like source activity (log-normal, mean = 5 Bq, $\sigma = 1$), transport parameters (velocity 0.01-0.5 m/s, diffusion log-normal, mean = 0, $\sigma = 0.5 \text{ m}^2/\text{s}$), and repository failure rates (0.1-5%). The 20.27% probability of workers exceeding 20 mSv/year and 66.92% for the public exceeding one mSv/year indicate that current parameters may overestimate exposure risks, potentially due to conservative assumptions or insufficient shielding (International Commission on Radiological Protection [ICRP], 2007).

The challenges identified are critical. Computational cost, affecting 1,999 simulations with a mean dose impact of 0.073690 Sv, highlights resource demands for large-scale simulations, which can limit practical application in real-time risk assessment (Zaider & Rossi, 1980). Parameter uncertainty occurring in 2,427 simulations with a 0.192871 Sv impact stems from unmeasured or variable radionuclide concentrations, complicating model reliability (Neretnieks, 1980). Validation difficulty noted in 2,086 simulations (0.103198 Sv impact) arises from high diffusion coefficients, making field comparisons challenging due to heterogeneous media (Tsang & Neretnieks, 1998).

These findings suggest that Monte Carlo methods, while powerful for uncertainty quantification, face limitations in computational feasibility, data availability, and validation against field data. Adjusting parameters (e.g., reducing source activity, increasing shielding, or refining transport models) could lower exceedance probabilities to align with ICRP guidelines (ICRP, 2007). Future studies should integrate stochastic validation with empirical data from repositories like Yucca Mountain to enhance accuracy (U.S. Department of Energy [DOE], 2002).

Comparing these Monte Carlo simulation results with real-world radioactive waste management data reveals significant differences, primarily due to parameter settings. Field studies, such as those at Yucca Mountain, report annual worker doses below 20 mSv (0.02 Sv) and public doses below one mSv (0.001 Sv), achieved through robust barriers (e.g., soil, concrete) and controlled distances (5-50 m) (U.S. Department of Energy [DOE], 2002). In contrast, this simulation's mean dose (0.071234 Sv) and exceedance probabilities (20.27% for workers, 66.92% for the public) suggest higher exposures, likely driven by the log-normal source activity (mean = 5 Bq, $\sigma = 1$), short distances (mean = 2 m, $\sigma = 0.7$), and variable failure rates (0.1-5%).

Real-world repositories use shielding (e.g., 90-99% reduction) and slower transport velocities (0.01-0.1 m/s), reducing doses far below those modeled here (National Council on Radiation Protection and Measurements [NCRP], 1993). The simulation's diffusion coefficient (log-normal, mean = 0, $\sigma = 0.5 \text{ m}^2/\text{s}$) and decay assumptions may overestimate dispersion and decay, respectively, compared to field data, where sorption and heterogeneity slow migration (Neretnieks, 1980). The high frequency of challenges (e.g., 2,427 parameter uncertainty cases) contrasts with repository operations, where extensive characterization minimizes uncertainty (Tsang & Neretnieks, 1998).

Adjusting parameters—reducing source activity to 0.5 Bq, increasing mean distance to 10 m, and enforcing 99% shielding—would align results with empirical data, lowering mean doses below 0.001 Sv and exceedance probabilities to near zero, consistent with ICRP standards (International Commission on Radiological Protection [ICRP], 2007). This comparison highlights the need for parameter calibration to reflect real-world conditions; ensuring Monte Carlo models accurately assess waste management risks.

The comparison of Monte Carlo and deterministic methods in EIA and radioactive waste management reveals their complementary roles, as noted by Kalos and Whitlock (2008). The deterministic ADE solution, with a mean concentration of 0.0471 and a maximum

of 0.9798, offers precise predictions for well-characterized systems. However, its 100.00-day barrier reach time exceeds the 50-day simulation, limiting long-term risk assessment. It assumes fixed parameters ($D=0.1 \text{ m}^2/\text{day}$, $v=0.5 \text{ m/day}$, $\lambda=0.01 \text{ 1/day}$, $\lambda=0.011/\text{day}$), potentially missing variability in real-world heterogeneity (Freeze & Cherry, 1979).

Conversely, the Monte Carlo simulation, with a mean dose of 0.022822 Sv and 81.60% public exceedance, excels in handling uncertainty, reflecting stochastic inputs like distance and shielding. However, its high public exceedance probability suggests parameter conservatism (e.g., short distances, high activity), requiring validation against field data (Neretnieks, 1980). The 18.60% worker exceedance aligns with risk assessment needs but indicates computational demands and convergence challenges (Kalos & Whitlock, 2008).

Deterministic models are computationally efficient but rigid, while Monte Carlo offers flexibility for EIA, capturing rare events at a higher cost. Integrating both, as in repository design, enhances safety per ICRP guidelines (ICRP, 2007), balancing precision and uncertainty for waste management. Future studies should refine parameters (e.g., lower activity, longer distances) to align with empirical data from sites like Yucca Mountain (U.S. Department of Energy [DOE], 2002).

The performance of Monte Carlo and deterministic methods in EIA and waste management differs significantly, as shown in the results. Deterministic modeling, with a mean concentration of 0.0471 and a maximum of 0.9798, provides precise, reproducible outcomes for radionuclide transport. However, its 100.00-day barrier reach exceeds the 50-day simulation, limiting risk assessment for long-term scenarios (Freeze & Cherry, 1979). It assumes homogeneous parameters ($D=0.1 \text{ m}^2/\text{day}$, $v=0.5 \text{ m/day}$), potentially underestimating variability in real repositories like Yucca Mountain (U.S. Department of Energy [DOE], 2002).

Monte Carlo, with a mean dose of 0.022822 Sv and 81.60% public exceedance, captures uncertainty through stochastic sampling, reflecting real-world variability in distance, shielding, and exposure (Kalos & Whitlock, 2008). However, its high exceedance probabilities (18.60% for workers, 81.60% for the public) suggest overly conservative parameters (e.g., 500 Bq activity, mean distance = 2 m), contrasting with field data where public doses are typically below one mSv/year (ICRP, 2007). The deterministic approach is computationally efficient but lacks flexibility, while Monte Carlo handles complex uncertainties but incurs higher computational costs and requires validation (Neretnieks, 1980).

Field studies at nuclear sites show worker doses below 20 mSv/year and public doses below 1 mSv/year, achieved through robust barriers and distances (5-50 m) (National Council on Radiation Protection and Measurements [NCRP], 1993). Adjusting Monte Carlo parameters (e.g., reducing activity to 50 Bq, increasing distance to 10 m) would align exceedance probabilities with these standards, complementing deterministic precision for EIA and waste management. Kalos and Whitlock (2008) advocate hybrid approaches, leveraging deterministic baselines and Monte Carlo uncertainty for comprehensive risk assessment.

The results of the Monte Carlo simulation demonstrate the expected convergence behavior, where the absolute error (ϵ) decreases as the number of simulations (N) increases, validating the accuracy of the Monte Carlo model. The final estimated π of 3.142072 at $N=1,000,000$, with an error of 0.000479, is notably close to the true value of π (3.14159265359), indicating high accuracy at large N . This small error underscores the effectiveness of Monte Carlo methods in approximating physical quantities through random sampling, particularly when computational resources allow for a large number of iterations.

The log-log plot in Figure 6 further confirms the convergence trend, showing a clear downward trajectory in error as N increases from 10^2 to 10^6 . However, the fitted error scaling exponent of -0.41 deviates from the theoretical expectation of -0.5, which is characteristic of Monte Carlo methods where the error typically scales as $\epsilon \propto 1/\sqrt{N}$ (Metropolis & Ulam, 1949). This discrepancy may arise from several factors. First, the stochastic nature of Monte Carlo simulations introduces variability, as seen in the fluctuations of the error data, particularly at lower N values. These fluctuations can affect the accuracy of the fitted exponent, especially if the sample sizes (N values) are not sufficiently dense across the range. Second, the simulation's design estimating π using a quarter-circle method may introduce systematic biases or boundary effects that slightly alter the error scaling behavior. For instance, the uniform random sampling used here assumes perfect randomness, but finite computational precision or pseudo-random number generator limitations

could influence the results (Knuth, 1997).

The observed exponent of -0.41, while close to the theoretical -0.5, suggests a slightly slower error reduction rate than expected. This could imply that additional simulations at even higher N values (e.g., $N > 106$) might be needed to better align with the theoretical scaling, or that the model could benefit from refined sampling techniques, such as stratified sampling, to reduce variance (Hammersley & Handscomb, 1964). Despite this, the small final error ($\epsilon = 0.000479$) at $N=1,000,000$ indicates that the Monte Carlo model is reliable for practical applications, such as radiation dose assessments or shielding design, where high accuracy is critical. The convergence behavior observed here supports the use of Monte Carlo methods in such contexts, provided sufficient computational resources are available to achieve low error levels.

Future work could focus on increasing the density of N values in the convergence test to improve the accuracy of the fitted exponent, as well as exploring alternative Monte Carlo techniques to minimize variance and enhance convergence. Additionally, applying this validation approach to more complex systems, such as particle transport in radioactive waste management, could further test the robustness of the model's accuracy and scaling behavior.

Conclusions and Recommendations

Conclusions

The comparative analysis of Monte Carlo and deterministic methods in radioactive waste management and environmental impact assessment (EIA) underscores their complementary roles, as supported by Kalos & Whitlock (2008). The deterministic advection-dispersion equation (ADE) model, with a mean concentration of 0.0471 and a maximum of 0.9798, offers precise, reproducible predictions for radionuclide transport over a 100-meter domain in 50 days. However, its 100.00-day barrier reach time highlights limitations in capturing long-term variability. This precision is valuable for baseline scenarios but may overlook real-world heterogeneity and rare events, as seen in its fixed parameters ($D=0.1 \text{ m}^2/\text{day}$, $v=0.5 \text{ m/day}$, $\lambda=0.01 \text{ 1/day}$).

In contrast, the Monte Carlo simulation, with a mean dose of 0.022822 Sv, a standard deviation of 0.143987 Sv, an 18.60% worker exceedance, and an 81.60% public exceedance, effectively captures uncertainty and variability through stochastic sampling of parameters like source activity, distance, and shielding. However, its high public exceedance probability (81.60%) suggests conservative parameters (e.g., 500 Bq activity, mean distance = 2 m), necessitating validation to align with International Commission on Radiological Protection (ICRP) limits (ICRP, 2007). Monte Carlo's ability to model rare, high-dose events (e.g., 6.066367 Sv outliers in prior analyses) is a key advantage. However, it incurs higher computational costs and convergence challenges (Kalos and Whitlock 2008). These methods enhance EIA and waste management by combining deterministic precision with Monte Carlo's probabilistic risk assessment, ensuring robust safety protocols for repositories and environmental protection.

Recommendations

Based on the comparative analysis, the following recommendations are proposed to optimize the application of Monte Carlo and deterministic methods in radioactive waste management and environmental impact assessment (EIA):

- **Integrate Hybrid Approaches:** Combine Monte Carlo and deterministic models to leverage deterministic precision for baseline scenarios and Monte Carlo's ability to quantify uncertainty, as suggested by Kalos & Whitlock (2008).
- **Refine Model Parameters:** Adjust Monte Carlo parameters to reflect real-world conditions, such as reducing source activity, increasing mean distance, and enhancing shielding effectiveness to align exceedance probabilities with field data.
- **Conduct field validation** against empirical data from radioactive waste repositories to address parameter uncertainty, validation difficulty, and convergence issues, as identified in Monte Carlo simulations.
- **Increase deterministic and Monte Carlo simulation times** (e.g., beyond 50 days) to capture long-term radionuclide migration and barrier interactions, aligning with real-world timescales.

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