# January Report: Adapting Particle Transport Monte Carlo Code to Solve the 1D Time Independent Schrödinger Equation

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#### Recap

• The following differential equations have the same form.

1D Time Independent Schrödinger Equation

$$\frac{-\hbar^2}{2m}\frac{d^2}{dx^2}\varphi(x) + V(x)\varphi(x) = E\varphi(x)$$

Steady State Neutron Diffusion Equation

$$-D\frac{d^2}{dx^2}\phi(x) + \Sigma_a\phi(x) = \frac{\nu}{k}\Sigma_f\phi(x)$$

 $\phi$ : neutron flux,

*D*: diffusion coefficient,

 $\Sigma_a$ : macroscopic absorption cross section,

*k*: multiplication factor,

 $\Sigma_f$ : macroscopic fission cross section ,

 $\nu$ : average number of neutrons released per fission

#### Recap

 The particle transport Monte Carlo code already exists for neutron diffusion.

ex) MCNP (Monte Carlo N-Particle)

• We can interpret the result of the Monte Carlo simulation as the solution of the Schrödinger equation.

#### Recap

- The Monte Carlo method is highly effective for solving problems with complicated internal geometries.
- Existing particle transport codes (MCNP, OpenMC, etc.) have been developed and validated in nuclear engineering. Adapting these codes allows for efficient reuse of mature algorithms and demonstrates their versatility in addressing new problems, such as quantum mechanics.
- While Quantum Monte Carlo (QMC) methods are well-established for solving the Schrödinger equation, the direct adaptation of particle transport Monte Carlo codes for quantum problems remains a relatively unexplored approach.

#### **Current Progress**

- 1) Study Monte Carlo particle transport method ✓ Done
- 2) Implement code for fundamental mode ✓ In progress
- 3) Implement code for higher modes
- 4) Interpret as a Schrödinger equation solution
- 5) Add quantum effects to code
- 6) Apply to more complex situations
  - > Various potentials
  - Higher Dimensions
- 7) Study numerical and QMC methods
- 8) Compare performance between methods

#### Algorithm

- 1. Input boundary conditions, total number of neutrons, and cycles.
- 2. Initialize two stacks (fission banks) to store neutrons.
- 3. Fill one stack (**parent fission bank**) with the total number of neutrons. Each neutron has an initial position and direction, which is determined by the random number generator.

## Algorithm

#### 4. Simulate neutron diffusion.

- For the total number of cycles:
  - While the parent fission bank is not empty:
    - Pop a neutron from the parent fission bank.
    - While the neutron is not terminated:
      - Generate a random number and calculate the distance for the neutron to move.
      - Calculate the new position, considering boundary conditions.
      - Sample the collision type at the new position and add any produced neutrons to the child fission bank.
      - Count neutron flux at each position.
    - Tally neutron flux and k value.
    - Update the child fission bank to become the new parent fission bank.

#### Algorithm

- 5. Calculate mean neutron flux and standard deviation.
- 6. Calculate mean k value and standard deviation.
- 7. Plot mean neutron flux with one sigma confidence intervals.

#### Key Variables

```
boundary_condition // 0 for vacuum, 1 for reflective
N // total number of neutrons
M // total number of cycles
active_cycle // number of active cycles
inactive_cycle // number of inactive cycles
width // width of the reactor (potential well)
sigma_s // scattering cross section
sigma_c // capture cross section
sigma_f // fission cross section
```

## Key Variables

struct neutron // stores the neutron's position and direction struct neutron \*parent\_fission\_bank // array of struct neutron struct neutron \*child\_fission\_bank // array of struct neutron gsl\_vector \*neutron\_flux // stores neutron flux for each cycle double k\_cycle // stores k value for each cycle double k\_average // average of k value over all cycles double k\_sample\_standard\_deviation // standard deviation of k values over all cycles

double k1, k2 // theoretical k values for the first and second mode

# **Key Functions**

```
initialize_rng() // initializes the random number generator instance
random_number_generator() // generates a random number
initialize_fission_bank() // initializes neutron positions and directions in
the parent fission bank
determine_direction() // determines neutron directions
simulate_neutron_diffusion() // simulates neutron diffusion for one cycle
sample_collision_type() // samples collision type
add_fission_bank() // adds neutrons to child fission bank when fission
occurs
flux_counter() // updates the neutron_flux variable
fission_bank_size_adjustment() // adjusts the child fission bank size to N
```

#### How to Run the Program

- Ensure all files are in the same directory.
  - > trial1.c // main source code
  - > trial1\_functions.s // custom function source code
  - > trial1.h // header file
  - input.txt // stores input data
  - Makefile // build script
  - draw\_flux\_data.py // plotting script
  - > flux\_data.txt // neutron flux results
  - > text\_data.txt // k value results and information for debugging
  - >position\_data.txt // neutron position

#### How to Run the Program

- Open input.txt and specify the desired values.
  - boundary\_condition, N, M

```
input.txt
    Boundary condition (zero = vacuum, one = reflective): 0
    Number of neutrons: 100000
    Total number of cycles: 100
```

- Run the following command in the terminal.
  - make; ./trial1
- The program will take a few seconds to minutes to run.
- Check the terminal output for the k value and standard deviation.

#### How to Run the Program

 Adjust draw\_flux\_data.py to match boundary conditions and execute it.

60 61	<pre># plt.ylim(0.3, 0.5) # for reflective plt.ylim(0.0, 0.4) # for vacuum</pre>
79 80	<pre>#plt.ylim(0.3, 0.5) # for reflective plt.ylim(0.1, 0.5) # for vacuum</pre>

## Results – (1) Vacuum Boundary Condition

#### Input values

- ➤ Boundary condition: 0
- ➤ Number of neutrons: 100000
- > Total number of cycles: 100

#### Variables

- $\triangleright$  width = 20.0
- $\triangleright$  sigma\_s = 0.1
- $\triangleright$  sigma\_c = 0.07
- $\rightarrow$  sigma\_f = 0.06

## Results – (1) Vacuum Boundary Condition

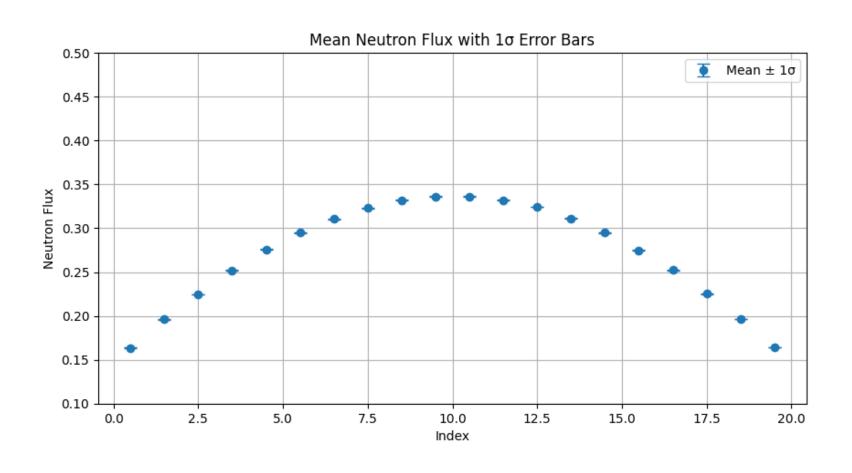
Terminal output

```
Boundary condition (0 = vacuum, 1 = reflective): 0
Number of neutrons: 100000
Total number of cycles: 100
Number of inactive cycles: 9
Number of active cycles: 91
k_a=0.790104, k_sa=0.624279, std=0.000403
k1=0.879588, k2=0.533992, k2/k1=0.607093

k1, k2, k2/k1
```

- $k_{average} = 0.790104 \pm 0.000403 (68.27\% Cl)$
- k1 is not within the range of k<sub>average</sub>

# Results – (1) Vacuum Boundary Condition



## Results - (1) Vacuum Boundary Condition

- $k_{average}$  being smaller than k1 means bigger leakage.
- In the Monte Carlo simulation, the flux at the boundary must be **theoretically zero**. Since every neutron that exits the boundary is terminated, and neutrons **cannot re-enter**, unlike in the reflective boundary condition, **this results in larger overall leakage**, **leading to a smaller k value**.
- In the **1D model**, the neutron's traveled distance represents its **3D** movement. However, since the full distance is applied in 1D, the neutron effectively travels farther than it would in **3D**.
- Instead of **sampling the direction in 1D** (left or right), a better approach is to **sample the solid angle** and project the traveled distance onto the x-axis.
- To Do: Implement a function to refine both approaches and verify the results.

# Results - (2) Reflective Boundary Condition

#### Input values

- ➤ Boundary condition: 1
- ➤ Number of neutrons: 100000
- > Total number of cycles: 600

#### Variables

- $\triangleright$  width = 20.0
- $\triangleright$  sigma\_s = 0.1
- $\triangleright$  sigma\_c = 0.07
- $\geqslant$  sigma\_f = 0.06

## Results - (2) Reflective Boundary Condition

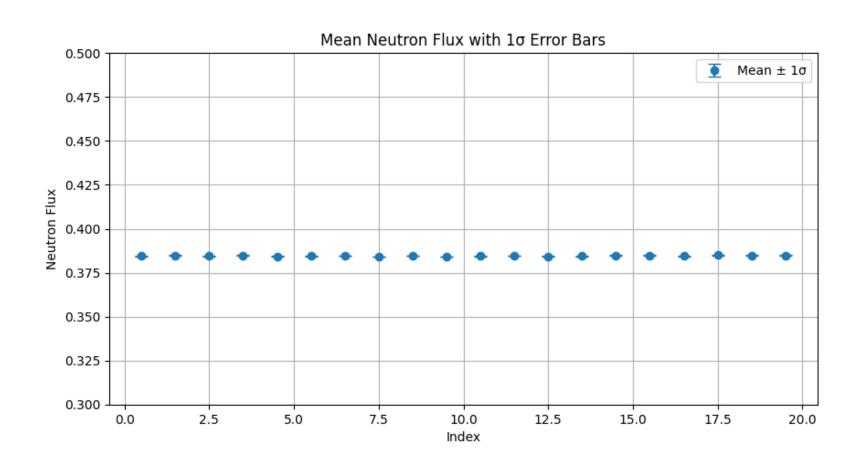
Terminal output

```
Boundary condition (0 = vacuum, 1 = reflective): 1
Number of neutrons: 100000
Total number of cycles: 600
Number of inactive cycles: 18
Number of active cycles: 582
k_a=1.121446, k_sa=1.257658, std=0.000168
k1=1.121538, k2=0.879588, k2/k1=0.784269

k1, k2, k2/k1
```

- $k_{average} = 1.121446 \pm 0.000168 (68.27\% Cl)$
- k1 is within the range of k<sub>average</sub>

# Results - (2) Reflective Boundary Condition



#### To Do List

- 1) Study Monte Carlo particle transport method
- 2) Implement code for fundamental mode
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  - > Various potentials
  - Higher Dimensions
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#### Resources

- Alex F Bielajew . (2020). Fundamentals of the Monte Carlo method for neutral and charged particle transport.
- Brown, F. B. (n.d.). Monte Carlo Techniques for Nuclear Systems. Lecture.
- Boundary conditions diffusion equation. Nuclear Power. (2021, October 28).
   https://www.nuclear-power.com/nuclear-power/reactor-physics/neutron-diffusion-theory/boundary-conditions-diffusion-equation/
- Leppänen, J. (2007). Development of a new Monte Carlo Reactor Physics Code (thesis). Development of a new Monte Carlo reactor physics code. VTT, Espoo.
- Shentu, J., Yun, S.-H., & Cho, N.-Z. (2007). A Monte Carlo method for solving heat conduction problems with complicated geometry. *Nuclear Engineering and Technology*, 39(3), 214. https://doi.org/10.5516/net.2007.39.3.207