

Shielding calculations with Markov chains and genetic algorithms

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Outline

- 1 Motivation
- 2 Inspiration
- 3 Transmission matrices
- 4 Markov chains
- 5 Genetic algorithms
- 6 Results
- 7 Applications

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How to stop a beam?

- \sim few GeV electrons,
 - but does not really matter
- Imagine how you would do it ...

Motivation

How to stop a beam?

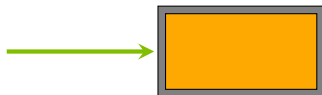


Probably:

- Use a heavy material block to stop γ
 - but keep in mind its Z to reduce (γ, n)

Motivation

How to stop a beam?



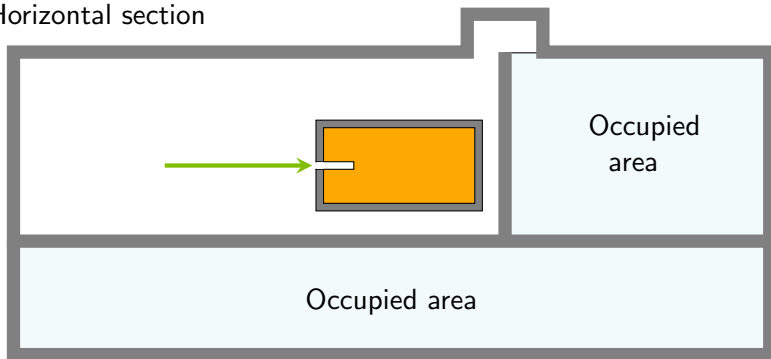
Probably:

- Use a heavy material block to stop γ
- Surround it with something light to slow down fast neutrons
 - to increase $\sigma(n,\gamma)$

Motivation

How to stop a beam?

Horizontal section



Probably:

- Use a heavy material block to stop γ
- Surround it with something light to slow down fast neutrons
- Optimise dimensions to reduce dose rates where needed

Motivation

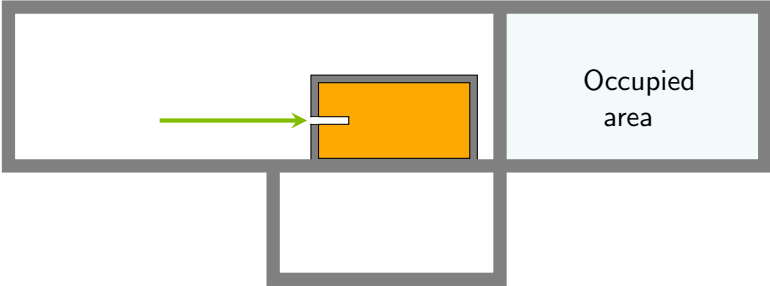
How to stop a beam? — Challenge

- Sounds easy?
- Challenge comes from the **boundary conditions**:
 - beam dumps are normally **large and heavy**
 - but this one should be **small and light**

Motivation

How to stop a beam? — Boundary conditions

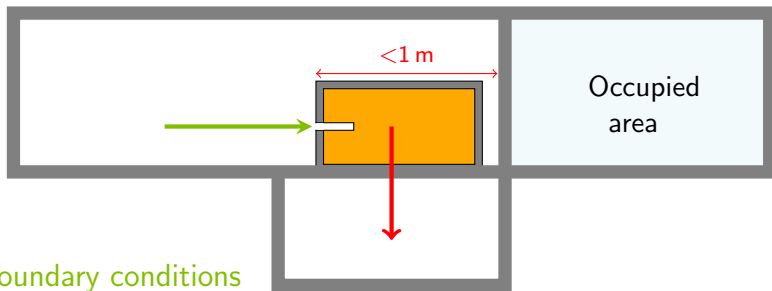
Vertical section



Motivation

How to stop a beam? — Boundary conditions

Vertical section



Boundary conditions

- Limit **dose rates** in the occupied areas
 - both forward and lateral
- Fit within the horizontal **space** (1 m)
- Minimise **weight** (load on the floor)
- Minimise **cost** and **complexity**

Multi-dimensional optimisation

- Parameter space
 - Dimensions
 - Materials
 - either heterogeneous or homogeneous

Multi-dimensional optimisation

- Parameter space
 - Dimensions
 - Materials
 - either heterogeneous or homogeneous
- Monte Carlo runtime
 - Single configuration: ~ 4 h on 100 cores ← slow
 - with aggressive variance reduction

⇒ Can't explore large parameter space

Simplification needed

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RESEARCH ARTICLE

MEDICAL PHYSICS

An iterative prediction method for designing the moderator used for the boron neutron capture therapy

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Funding information

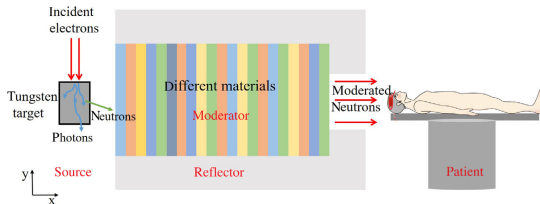
National Natural Science Foundation of China, Grant/Award Numbers: 11735008, 11975137

Abstract

Purpose: To conduct research related to slow neutrons, fast neutrons must be mode-rated and shifted to the desired energy region.

Methods: In this research, an iterated prediction method, in which the neutron transportation properties of all materials were characterized by a reflection matrix, R , and a transmission matrix, T , was proposed to bypass a time-consuming Monte Carlo simulation and predict the performance of the moderator, including the epithermal neutron flux and the dose of fast neutrons and gamma rays, used for boron neutron capture therapy (BNCT). To find the optimal solution in the huge parameter space, a genetic algorithm combined with transmission and reflection matrices was utilized.

Results: The results showed that a 70-loop iteration was able to find a design for the moderator of BNCT with almost 80% higher epithermal neutron flux per kilowatt than that of the empirically optimized moderator that was previously reported in the literature. Compared with the Monte Carlo method, this method had the advantage of reducing the calculation time and statistical errors.



[R. Zhang et al.]

- The paper describes a methodology of 1D neutron transport with **transmission matrices**
 - and optimisation of the moderator layout with **genetic algorithms**
- We extended it to transport **arbitrary number of particle types** including conversion between particle types
 - and improved accuracy utilising the **Markov chain process**

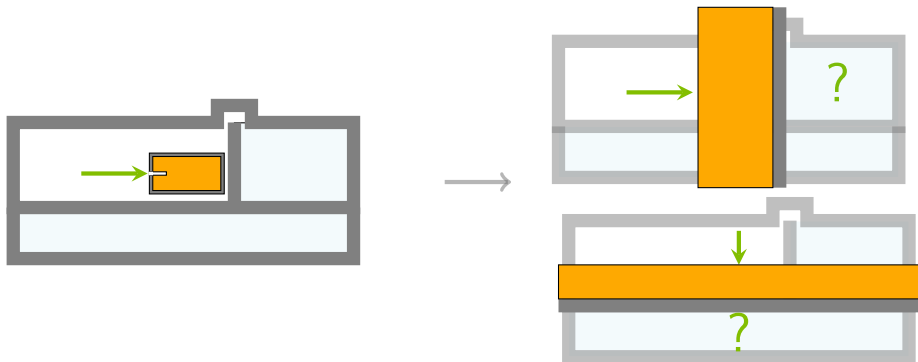
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Transmission matrices

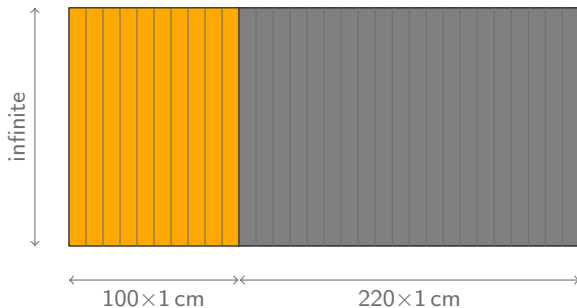
Introduction

The radiation transport is simplified by performing it along a single direction at a time



Transmission matrices

Introduction



- The material is split by thin layers
- For each material, its Green's functions are pre-calculated with Monte Carlo (needs to be done once)

Green's functions — definition

If one knows the solution $G(x, x')$ to a δ -function

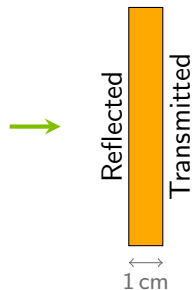
$$\hat{L}(x) G(x, x') = \delta(x - x')$$

then one can fold them to build the solution

$$u(x) = \int f(x') G(x, x') dx'$$

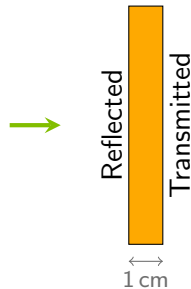
for a general source term

$$\hat{L}(x) u(x) = f(x)$$



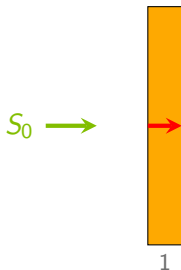
Green's functions — calculation

- 1 Bin the **energy range** of interest
 - 2 Bin the **angular range** $[-\pi, \pi]$
 - 3 For each incident **particle type, energy and angle** calculate double **differential spectra** of reflected and transmitted particles with Monte Carlo
 - 4 Use these spectra to build **reflection** and **transmission** matrices for each particle type, i.e. for neutrons and photons:
 - $R_{n \rightarrow n}, R_{n \rightarrow \gamma}, R_{\gamma \rightarrow n}, R_{\gamma \rightarrow \gamma}$
 - $T_{n \rightarrow n}, T_{n \rightarrow \gamma}, T_{\gamma \rightarrow n}, T_{\gamma \rightarrow \gamma}$
- Matrix shape: $(N_E \cdot N_\Omega) \times (N_E \cdot N_\Omega)$



Transmission matrices

Solution



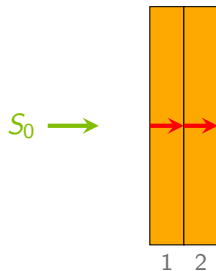
Solution for 1 layer

- $S_1 = S_0 \times T_1$

- S_i — spectra exiting layer i (and entering layer $i + 1$)
 - S_0 — source term (mixed particle spectra allowed)
- T_i — layer i transmission matrix
- R_i — layer i reflection matrix

Transmission matrices

Solution

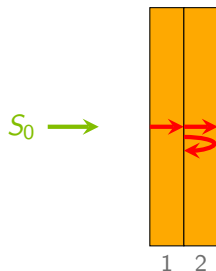


Solution for 2 layers

- $S_1 = S_0 \times T_1$
 - $S_2 \approx S_1 \times T_2$
-
- S_i — spectra exiting layer i (and entering layer $i + 1$)
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Transmission matrices

Solution



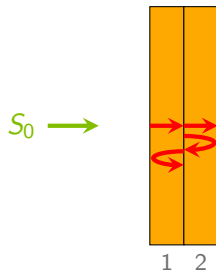
Solution for 2 layers

- $S_1 = S_0 \times T_1$
- $S_2 \approx S_1 \times T_2 + S_1 \times R_2$

- S_i — spectra exiting layer i (and entering layer $i + 1$)
 - S_0 — source term (mixed particle spectra allowed)
- T_i — layer i transmission matrix
- R_i — layer i reflection matrix

Transmission matrices

Solution



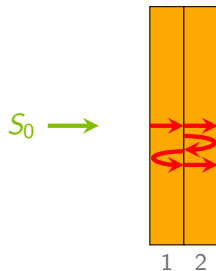
Solution for 2 layers

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Transmission matrices

Solution



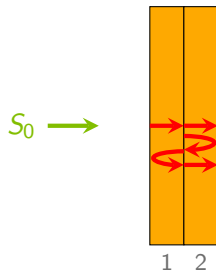
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Transmission matrices

Solution



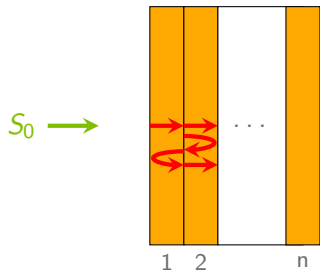
Solution for 2 layers

- $S_1 = S_0 \times T_1$
- $S_2 \approx S_1 \times T_2 + S_1 \times R_2 \times R_1 \times T_2$
+ higher order reflections

- S_i — spectra exiting layer i (and entering layer $i + 1$)
 - S_0 — source term (mixed particle spectra allowed)
- T_i — layer i transmission matrix
- R_i — layer i reflection matrix

Transmission matrices

Solution



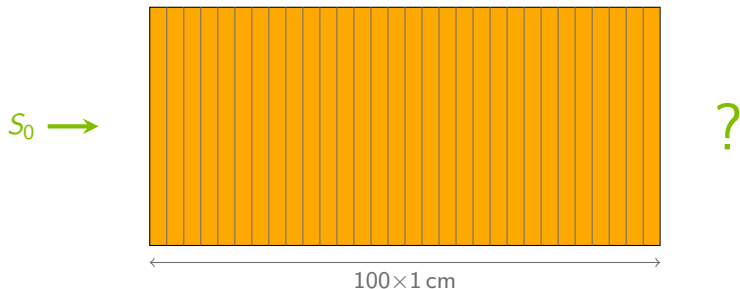
Solution for n layers

- $S_1 = S_0 \times T_1$
- $S_2 \approx S_1 \times T_2 + S_1 \times R_2 \times R_1 \times T_2$
+ higher order reflections
- $S_n \approx \dots$

- S_i — spectra exiting layer i (and entering layer $i + 1$)
 - S_0 — source term (mixed particle spectra allowed)
- T_i — layer i transmission matrix
- R_i — layer i reflection matrix

Transmission matrices

Results



- Incident beam: 3 GeV electrons
- Material: Concrete
- Thickness: 1 m (100 layers)
- Transported particles: n , γ , e^\pm , μ^\pm
- Energy: 130 bins between 1 meV and 5 GeV
- Emission angles: 18 bins between $-\pi$ and π

Matrix shape:

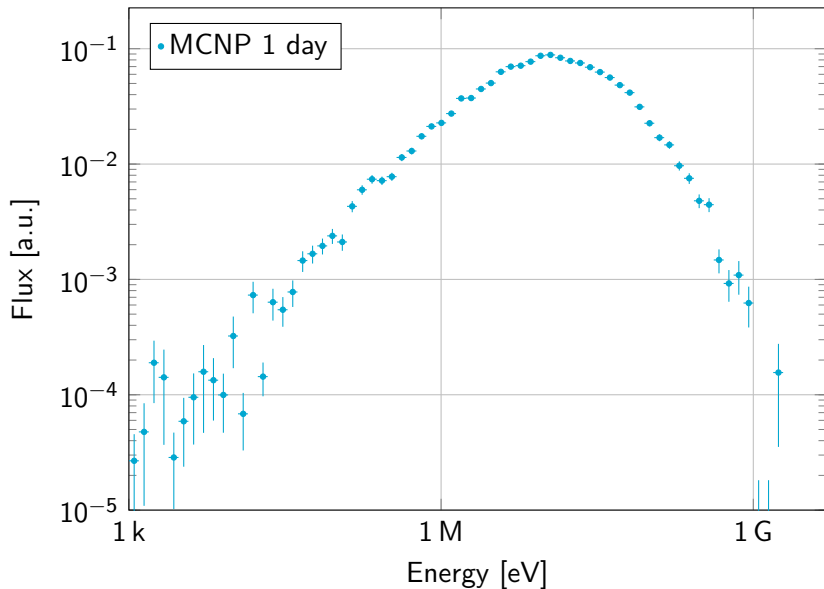
$130 \cdot 18 \Rightarrow$

2340×2340

Photons

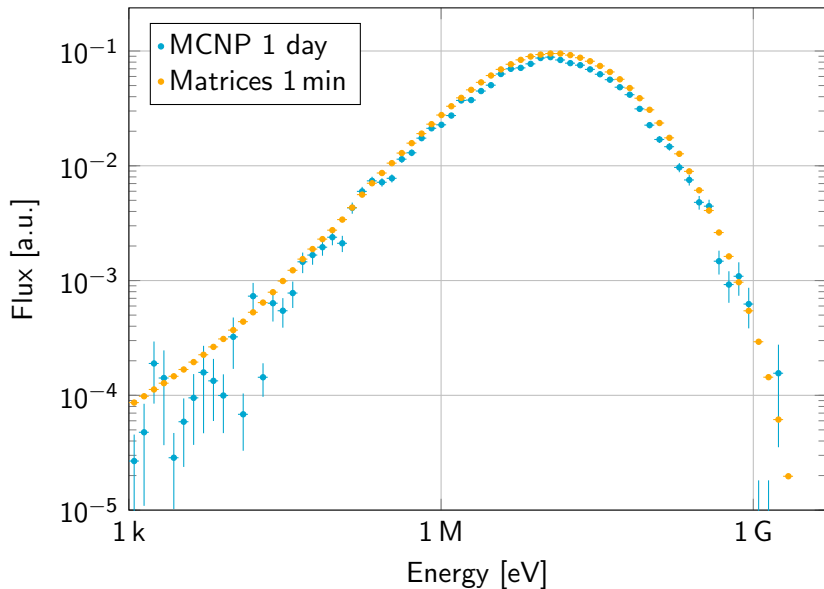
Transmission matrices

Results — Photons



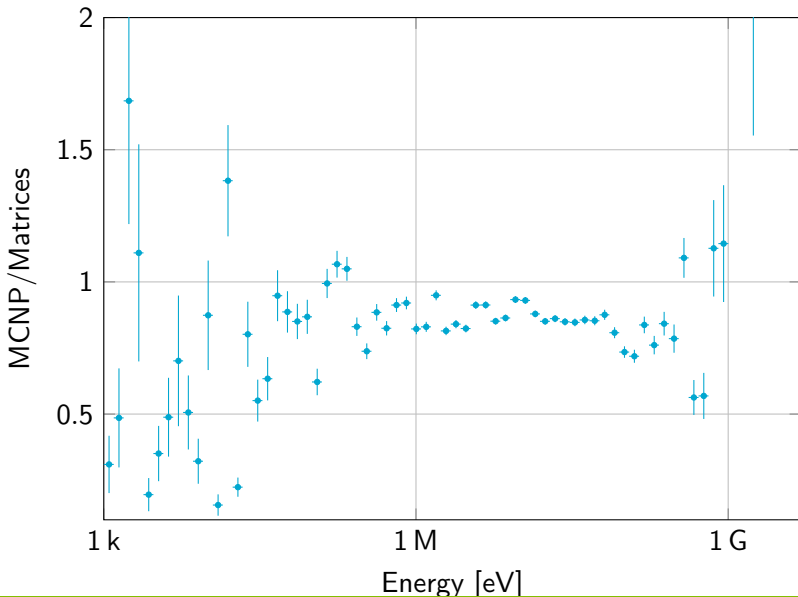
Transmission matrices

Results — Photons



Transmission matrices

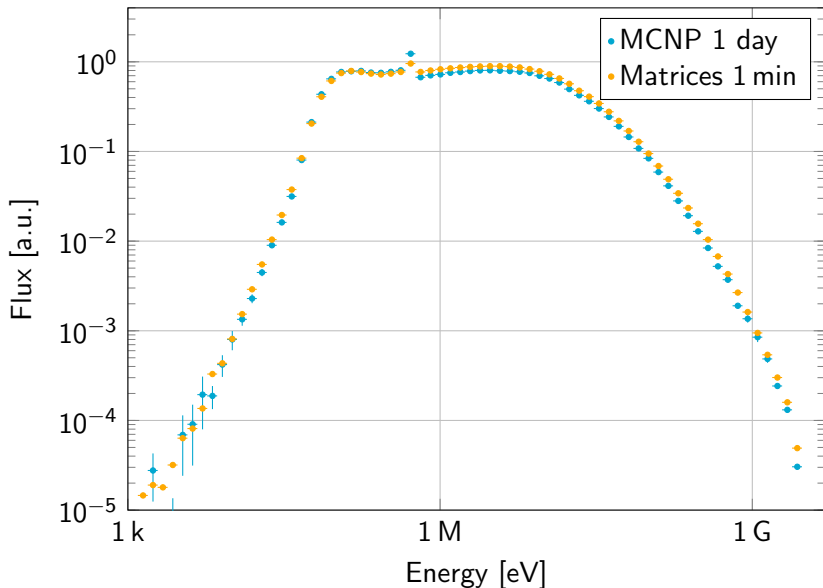
Results — Photons



Electrons

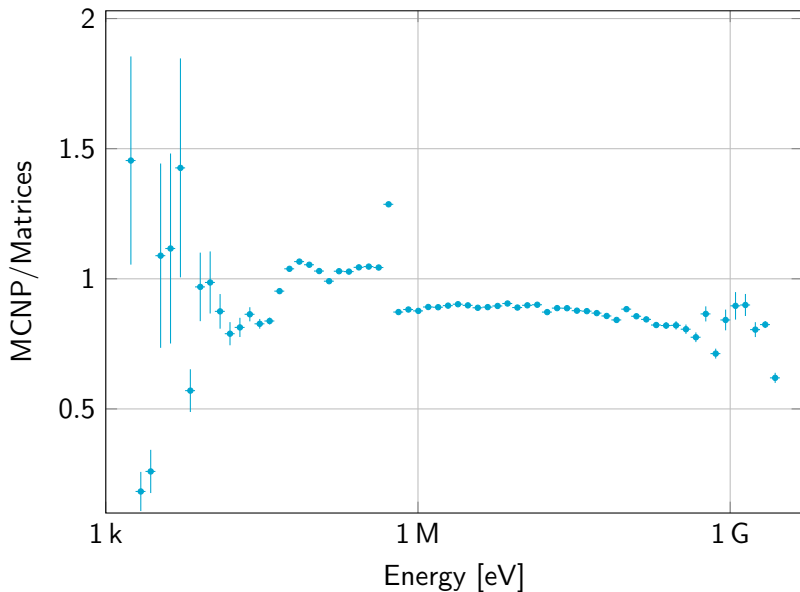
Transmission matrices

Results — Electrons



Transmission matrices

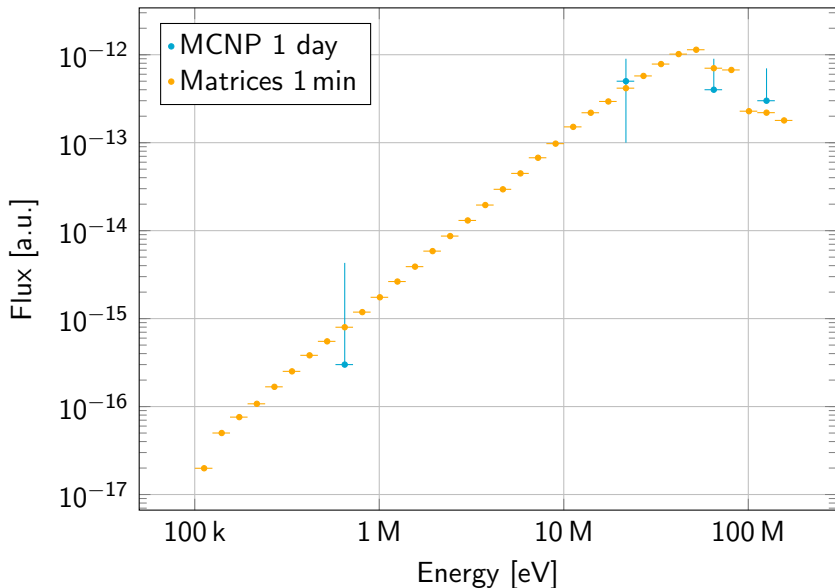
Results — Electrons



Muons

Transmission matrices

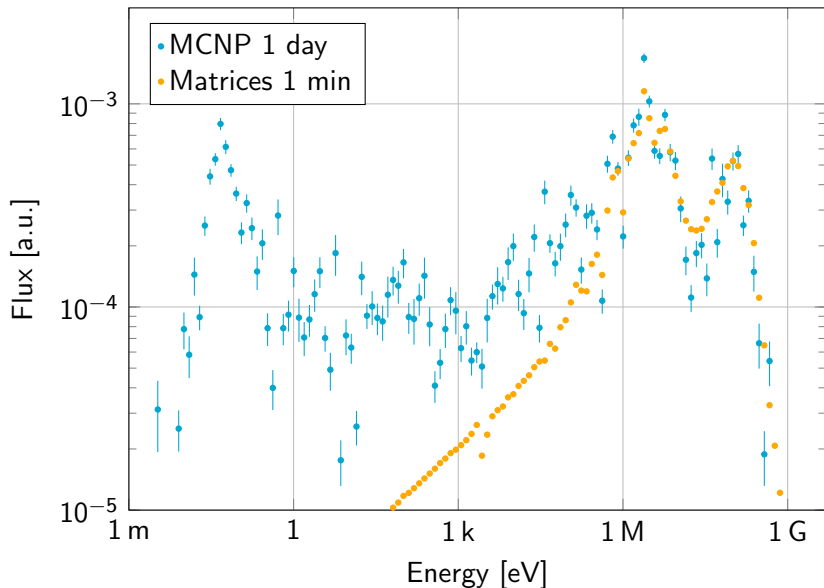
Results — Muons



Neutrons

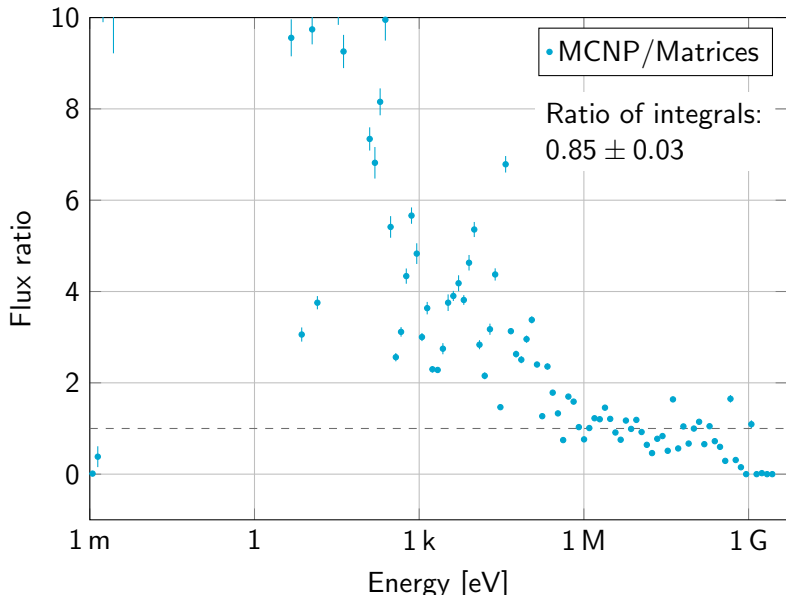
Transmission matrices

Results — Neutrons

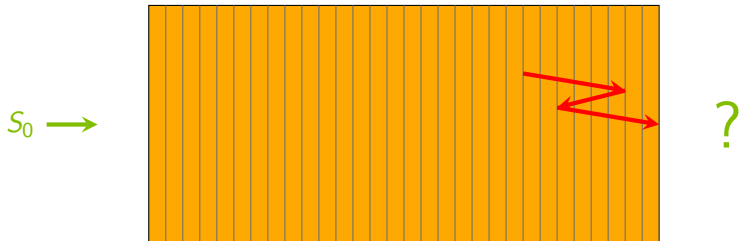


Transmission matrices

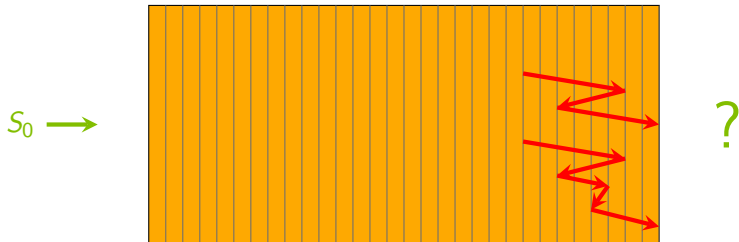
Results — Neutrons



Motivation for using Markov chains to improve neutron spectrum



Motivation for using Markov chains to improve neutron spectrum



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Future depends on today but not on yesterday

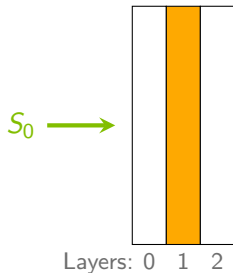
Typical examples

- Brownian motion
- Poisson processes
- Stock market
- Google's PageRank
- Monopoly game
 - search: "Dominating Monopoly using Markov chain"

Markov chains

Block matrix definition for 1 layer

	0	1	2
0	[0]	[E]	[0]
1	[R ₁]	[0]	[T ₁]
2	[0]	[0]	[0]

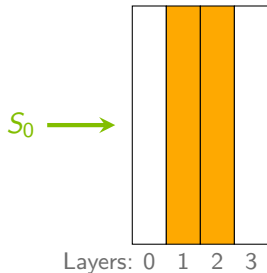


■ Matrix shape: $((N_{layers} + 2) \cdot N_E \cdot N_\Omega \cdot N_{particle\ types}) \times ((N_{layers} + 2) \cdot N_E \cdot N_\Omega \cdot N_{particle\ types})$

Markov chains

Block matrix definition for 2 layers

	0	1	2	3
0	[0]	[E]	[0]	[0]
1	[R ₁]	[0]	[T ₁]	[0]
2	[0]	[R ₂]	[0]	[T ₂]
3	[0]	[0]	[0]	[0]



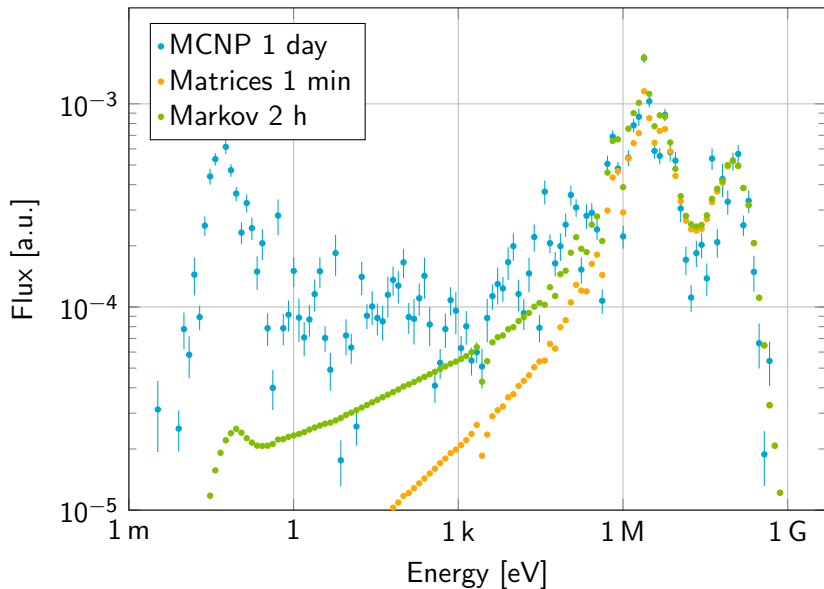
■ Matrix shape: $((N_{layers} + 2) \cdot N_E \cdot N_{\Omega} \cdot N_{particle\ types}) \times ((N_{layers} + 2) \cdot N_E \cdot N_{\Omega} \cdot N_{particle\ types})$

$$S_n \leftarrow S_0 \times (M_n)^k$$

- S_0 — source term (mixed particle spectra allowed)
- S_n — spectra exiting n layers
- $(M_n)^k$ — Markov chain matrix for n layers in the power of k
 - The value of k defines the reflection orders to be included in the result
 - The process quickly converges $\Rightarrow k$ does not need to be large

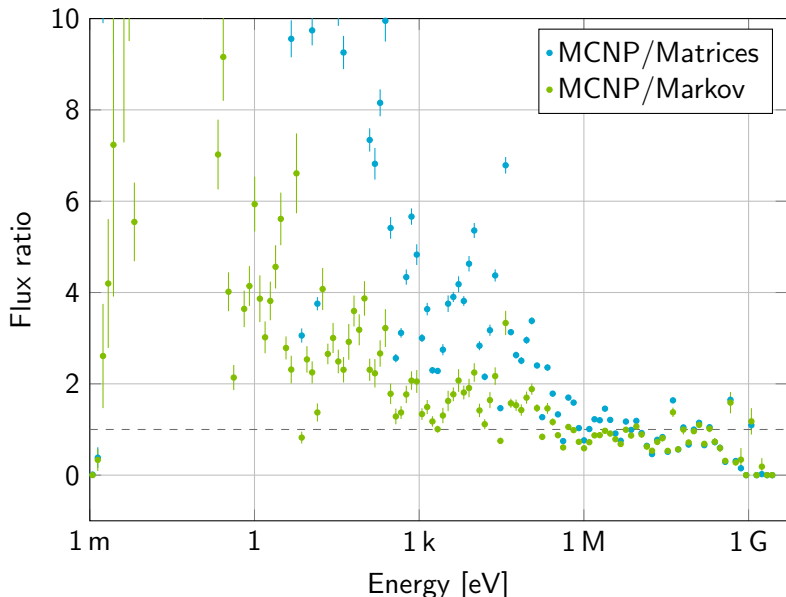
Markov chains

Results — Neutrons



Markov chains

Results — Neutrons



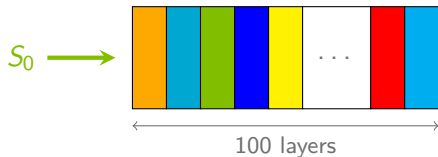
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Genetic algorithms

Motivation

- At this point we can quickly calculate particle spectra beyond thick shielding
- This allows us to test many heterogeneous beam dump configurations in 1D



- How many configurations to test?
 - Number of layers: $n=100$
 - Number of materials, e.g. $k=7$
 - ⇒ **brute force is still too time expensive**

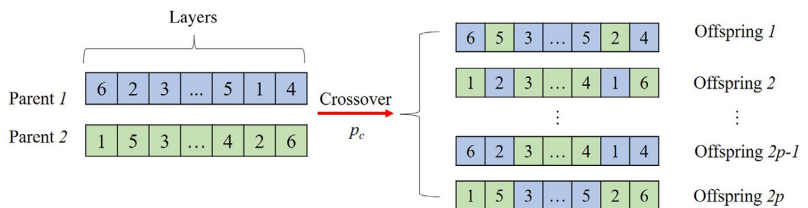
Genetic algorithms

Quick introduction

- Genetic algorithms implement natural selection principles
 - Crossover
 - Mutation
- They induce evolution of population to a state that maximises the specified figure-of-merit
- We can define a figure-of-merit to reflect our boundary conditions:
 - Dose rate
 - Size
 - Mass
 - Cost
 - Complexity

Genetic algorithms

Quick introduction — Crossover

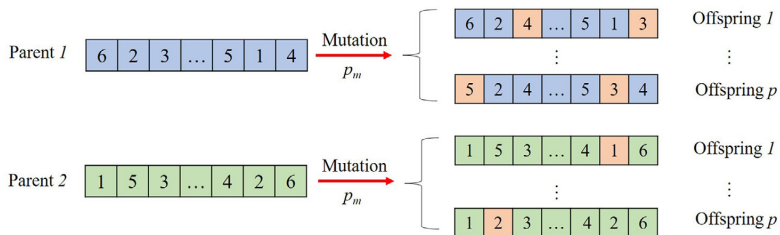


[R. Zhang et al]

- p_c — crossover probability
- Elements of both parents are randomly exchanged to form a new offspring

Genetic algorithms

Quick introduction — Mutation



[R. Zhang et al]

- p_m — mutation probability
- The offspring inherits the elements of its parent but some randomly chosen elements are randomly changed

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- Steel
- Lead
- Tungsten
- Concrete
- Boron carbide (B_4C)
- Water
- Polyethylene

Optimised beam dump layout

Results

Optimal beam dump configuration

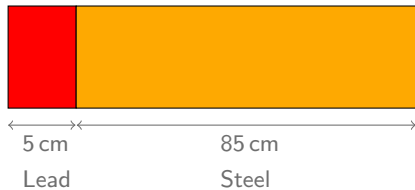


↔
5 cm

Lead

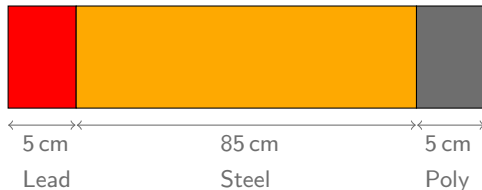
Results

Optimal beam dump configuration



Results

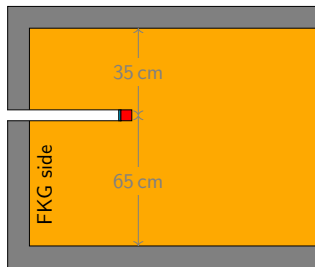
Optimal beam dump configuration



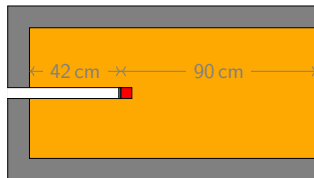
Almost the same as our initial guess!

Results

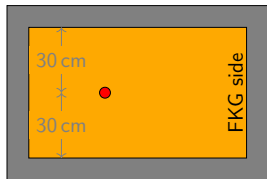
Final geometry (fine-tuned with FLUKA)







(a) Horizontal section



(b) Vertical section



(c) Vertical section

-  Kapton tape: ~0.1 cm thick.
-  Lead: 3 cm radius, 5 cm length.
-  Steel: S235JR.
-  Borated polyethylene: 5 cm thick.

Results

Final geometry (as built without Polyethylene)



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MAG — an open source tool for 1D shielding calculations:

<https://github.com/kbat/mag>

Solver

```
$ mag-solve -layers 5 Lead 100 Concrete 5 Poly  
            -source e 3e3
```

Dose rates [pSv/primary]:

e: 1.13

n: 0.04

γ : 1.42

μ : 1.49e-11

total: 2.56

+ data file with spectra for each particle type

MAG — an open source tool for 1D shielding calculations:

<https://github.com/kbat/mag>

Optimiser

```
$ mag-optimise -nlayers 100 -source e 3e3
```

- Optimises a slab made of 100 layers filled with materials from the existing database
- Arbitrary figure-of-merit can be specified
- Complex sources are supported

- Simple shielding calculations
 - e.g. non-neutron applications
 - e.g. for non-FLUKA users
- General particle transport through matter
- Layer/mesh-based variance reduction generation



Zhang R, Yu Y, Zhang Z, Yang Y.

An iterative prediction method for designing the moderator used for the boron neutron capture therapy

Med Phys. 2022; **49**:598–610.

<https://doi.org/10.1002/mp.15339>

2022