

Presentation Outline

- 1 Introduction
- 2 Theoretical Basis
 - The Ising model
 - Physical observables
- 3 Methodology
 - Metropolis-Hastings Algorithm
 - The Worm Algorithm
- 4 Results
 - Algorithm behavior
 - Susceptibility and Heat Capacity
 - Autocorrelation time - Dynamical Exponent
- 5 Discussion

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Introduction

- The Metropolis algorithm is a widely used Monte Carlo method for the Ising model.
- The problem of critical slowing down
- Prokof'ev and Svistunov proposed an alternative update algorithm called the Worm Algorithm (WA)
- WA preserves the local nature of the update step, but achieves a very small dynamical exponent.

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- A mathematical model to understand the behaviour of systems phase transitions, like ferromagnetic materials

- 2D Ising model: Magnetic system consisting of interacting spins on a two-dimensional lattice

The Ising model

Hamiltonian:

$$H = J \sum_{\langle i,j \rangle} s_i s_j - h \sum_i s_i$$

Partition function:

$$Z = \sum_s e^{-\beta H} = \sum_s e^{-[-\beta J \sum_{\langle i,j \rangle} s_i s_j - \beta h \sum_i s_i]}$$

Physical observables

describe the properties of the system change at phase transition

- Magnetization

$$M = \frac{1}{N} \sum_i^N \sigma_i \quad (1)$$

with N being the amount of sites. For a square lattice, one simply replaces N by L^2 , where L is the lattice length and σ the spin.

- Energy
- Autocorrelation - Dynamical Exponent z $\tau \approx L^z$ for large J and β
- Susceptibility

$$\chi = (k_B \beta) \cdot (\langle M^2 \rangle - \langle M \rangle^2) \quad (2)$$

- Specific heat

$$C = (k_B \beta)^2 \cdot (\langle E^2 \rangle - \langle E \rangle^2) \quad (3)$$

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Metropolis-Hastings Algorithm

- Monte Carlo simulation method
- generate samples from a probability distribution
- iterative update of the system with the Accept-Reject method
- Implement Metropolis-Hastings method
 - 1 Random configuration of an $N \times N$ lattice
 - 2 Flip the spin at the site
 - 3 Calculate the energy cost ΔE of the flip
 - 4 The reject/accept step

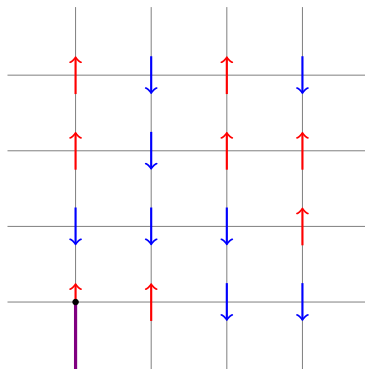
The Worm Algorithm

- The worm algorithm is an alternative to the standard Metropolis algorithm.

- There are many variations of the worm algorithm in literature.
- Our first implementation of the worm algorithm is as follows:
 - 1 Start with an arbitrary lattice configuration.¹
 - 2 Select an arbitrary point to create the worm.
 - 3 Grow the worm:
 - 1 Choose a random direction to move.
 - 2 If the new point is of the same spin as the old point, add it to the worm with probability 1.
 - 3 If not, perform a Metropolis-like check. If a flip is favourable, flip and add
 - 4 This way, we create a worm with equal spins.
 - 5 Break when the worm head meets its tail or if a flip is no longer favourable.
 - 4 Measure the observables.
 - 5 Carry out the desired number of iterations, with a new worm every time.
- There are a few caveats.

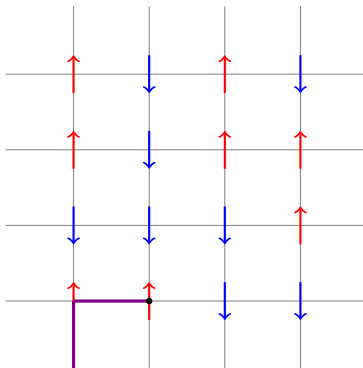
¹We started with an “almost” cold start.

- Given below is a worm at an intermediate step of its growth.
- The lattice below is a small part of the total lattice.

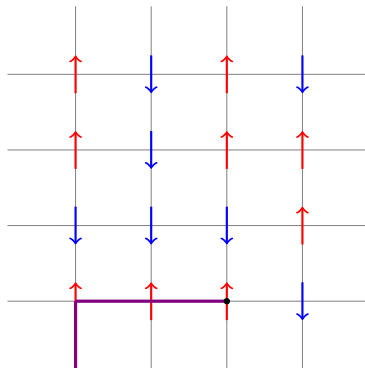


The Worm Algorithm

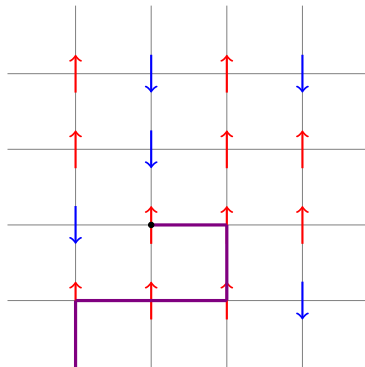
- The worm decides to move right.
- The new site has the same spin as the old site.



- The worm decides to move right.
- The new site has the opposite spin as the old site.
- Now we perform a metropolis-like check and decide whether to flip or not.

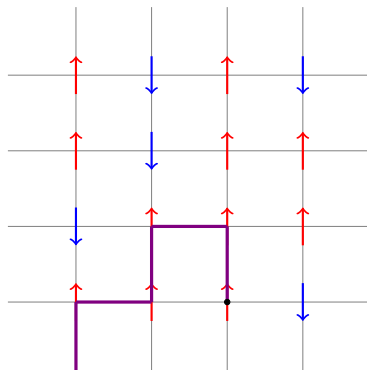
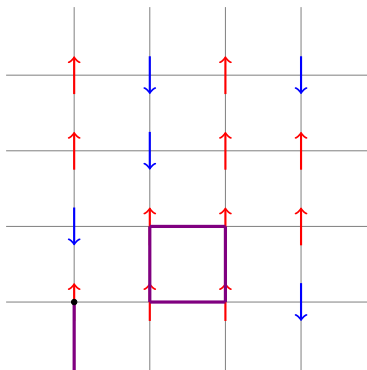


- The worm after a few more steps.



The Worm Algorithm

- Caveat: If the worm tries to move to a point that is already a part of the worm, we need to update the head accordingly.
- We want only an even number of bonds between the sites.
- We ensure this by choosing the new head and breaking an old bond appropriately.



- Caveat: If the worm tries to leave the lattice, let it choose a different direction.

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Algorithm behavior - Metropolis

Algorithm behavior - Worm Algorithm

Susceptibility and Heat Capacity - Metropolis

Susceptibility and Heat Capacity - Worm Algorithm

Dynamical Critical Exponent - Metropolis

Dynamical Critical Exponent - Worm Algorithm

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