

# The Ising model

## Computational Physics

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## 1 Introduction

In this paper we will investigate the different properties of the Ising model in 2D and 3D using Monte Carlo simulation methods. The canonical Metropolis single spin flip method as well as the Wolff cluster flip algorithm are used to obtain values for the critical exponents  $\chi$ ,  $\gamma$  and  $\nu$  in both 2D and 3D.

## 2 Theory

### 2.1 Ising model

### 2.2 Monte Carlo approximation

## 3 Algorithms

### 3.1 Metropolis

The Metropolis algorithm [1] utilizes single spin flips to transition between states. By selecting a random spin at each step we make sure that the transition chance between any two neighbouring states  $\mu$ ,  $\nu$  is equal:

$$g(\mu, \nu) = g(\nu, \mu) = \frac{1}{N}$$

### 3.2 Wolff

The Metropolis algorithm has as major disadvantage that the thermalization slows down exponentially around the critical point. As this is the area in parameter space we are most interested in, another algorithm is needed. We use an algorithm proposed by Wolff in 1989 [3]. In essence, this algorithm stochastically builds up a cluster which it will flip at once, thus negating the critical slowdown.

## 4 Implementation

Both the Metropolis and the Wolff algorithm were implemented using Python and NumPy. While this choice of language will not result in the fastest implementation it yields easily readable and adaptable code. Moreover, simple libraries exist to parallelise the different runs to keep the total wall clock computation time reasonable. The other major benefit of using python is of course easy access to extensive plotting libraries.

## 5 Results

### 5.1 Thermalization& correlation

To visualize the thermalisation we do several runs for each value of  $\beta$ , calculate the magnetisation for each and take the mean across different runs. This way, the plot will be a flat line when thermalized. Results for a  $40 \times 40$  grid and 100 runs per value of beta is given in figure 1. The obvious conclusion here is that the thermalisation time depends quite strongly on  $\beta$ , especially around the critical point, which corresponds with the interesting part of the simulations. An algorithm which would bypass this behaviour would thus be beneficial. The Wolff algorithm is such an algorithm, as although it still needs to thermalize, the thermalization time does not diverge around the critical point. The

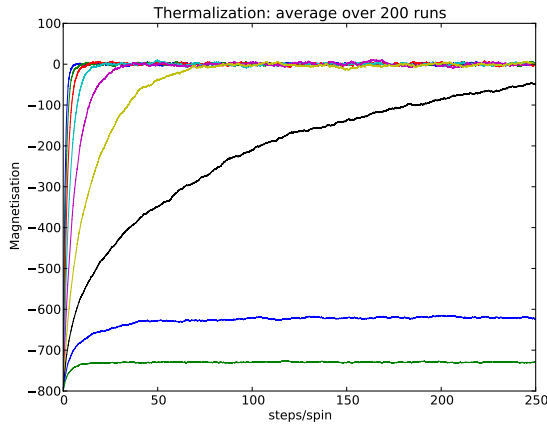


Figure 1: Thermalization

normalized autocorrelation of the Magnetisation and the energy for diverse values of  $\beta J$  is given in figures ?? and ??.

### 5.2 Finite size scaling

## 6 Data collapse

### 6.1 Rectangular 2D lattice

To obtain a data collapse we use the values as found by Newman and Barkema [2]:  $\gamma = 1.76$ ,  $\nu = 1.00$  and  $T_c = 2.269J$ . For the exponent  $\beta$  we used the exact value  $\frac{1}{8}$ . The result is shown in figure ??

### 6.2 Hexagonal 2d lattice

To check the universality of the critical exponents we also ran the simulations for a hexagonal lattice. All values are kept the same as compared to the previous section, except for  $T_c$ , which we determine by ...**TODO** The result can be seen in figure ??.

### 6.3 3 dimensions

## References

- [1] N. Metropolis, A. W. Rosenbluth, M. N. Rosenbluth, A. H. Teller, and E. Teller. Equation of State Calculations by Fast Computing Machines. *jcp*, 21:1087–1092, June 1953.
- [2] M. E. J. Newman and G. T. Barkema. *Monte Carlo Methods in Statistical Physics*. Clarendon Press, 1999.
- [3] Ulli Wolff. Collective monte carlo updating for spin systems. *Phys. Rev. Lett.*, 62:361–364, Jan 1989.