

A study of phase transition in the 2D Ising model through Markov Chain Monte Carlo simulation

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NB! Abstract here

I. INTRODUCTION

This report will explore temperature-dependent behavior in ferromagnetism using the two-dimensional **Ising model**. The main purpose is determining a numerical estimation of the **critical temperature** at which the system transitions from a magnetized to a non-magnetized phase.

The Ising model is a widely studied model in statistical physics. Work on the Ising model has shown to be useful for the analysis of several complex systems

... more about Ising model and phase transitions + Monte Carlo methods and Markov chains. [?]

II. METHODS

The square 2D lattices for our Ising model will have a length of L containing N spins with the relation $N = L^2$. Each spin s_i will have two possible states of

$$s_i = -1 \text{ or } s_i = +1.$$

The total spin state or **spin configuration** of a lattice will be represented as $\mathbf{s} = (s_1, s_2, \dots, s_N)$. In its simplest form the total energy of the system is expressed as

$$E(\mathbf{s}) = -J \sum_{\langle kl \rangle}^N s_k s_l - \mathcal{B} \sum_k^N s_k.$$

Here \mathcal{B} is an external magnetic field. Since we will be looking at the Ising model without an external magnetic field the equation will be simplified to

$$E(\mathbf{s}) = -J \sum_{\langle kl \rangle}^N s_k s_l, \quad (1)$$

where $\langle kl \rangle$ denotes the sum going over all *neighboring pairs* of spins avoiding double-counting. J is the **coupling constant** simply setting the energy associated with spin interactions. *Periodic boundary conditions* will be implemented allowing all spins to have four neighbors.

$$M(\mathbf{s}) = \sum_i^N s_i \quad (2)$$

is the magnetization of the entire system expressed as a sum over all spins. The energy per spin is

$$\epsilon(\mathbf{s}) = \frac{E(\mathbf{s})}{N} \quad (3)$$

and the magnetization per spin is given by

$$m(\mathbf{s}) = \frac{M(\mathbf{s})}{N}. \quad (4)$$

These will be used to compare and analyze results.

$$\beta = \frac{1}{k_B T} \quad (5)$$

describes the “inverse temperature” with the systems’ temperature T and the Boltzmann constant k_B .

$$Z = \sum_{\text{all possible } \mathbf{s}} e^{-\beta E(\mathbf{s})} \quad (6)$$

represents the partition function. This, the “inverse temperature” and the total energy of the system appear in the *Boltzmann distribution*,

$$p(\mathbf{s}; T) = \frac{1}{Z} e^{-\beta E(\mathbf{s})}. \quad (7)$$

This will be the probability distribution used for random sampling in our Monte Carlo approach.

For comparison with early numerical implementations we will first consider an analytical solution. The following table I summarizes all sixteen possible **spin configurations** of a 2×2 lattice with *periodic boundary conditions*.

TABLE I. Analytic values for the sixteen **spin configurations** of the 2×2 Ising model lattice.

Nr. of spins in state +1	Degeneracy	Total energy	Total magnetization
0	1	-8J	-4
1	4	0	-2
2	4	0	0
2	2	8J	0
3	4	0	2
4	1	-8J	4

Based on the values in table I we derive the specific analytical expressions for the 2×2 lattice case. The calculations of these analytic solutions can be found in appendix A and are used to test early versions of the code.

We will study properties of the system at equilibrium for different lattice sizes as a function of temperature T . The analysis will focus on four main properties. The

mean energy (eq. 3), mean magnetization (eq. 4), specific heat capacity

$$C_V = \frac{1}{N} \frac{1}{k_B T^2} (\langle E^2 \rangle - \langle E \rangle^2), \quad (8)$$

and the susceptibility

$$\chi = \frac{1}{N} \frac{1}{k_B T^2} (\langle M^2 \rangle - \langle |M| \rangle^2). \quad (9)$$

Periodic boundary conditions

The Ising model simulations we will be implemented using periodic boundary conditions. This way, all spins will have four neighbors as seen in table II, also at the boundaries of the lattice.



TABLE II. An example of a spin and its neighbors. Here all spins are in a positive state (+1).

Markov chains

Monte Carlo methods

The Metropolis algorithm

The Monte Carlo method will repeatedly require the Boltzmann factor $e^{-\beta \Delta E}$. The energy shift induced by

flipping a single spin

$$\Delta E = E_{\text{after}} - E_{\text{before}} \quad (10)$$

can only take five possible values in a 2D-lattice of arbitrary size. This is shown in appendix B. These values are

$$\Delta E = 0, -4J, 4J, -8J, 8J. \quad (11)$$

To reduce computational cost we will avoid repeatedly calling the exponential function. This is done by pre-computing the five possible Boltzmann factors in an array. NB MORE HERE.

III. RESULTS

IV. DISCUSSION

V. CONCLUSION

Appendix A: Analytical solutions for a 2×2 lattice

The specific partition function becomes

$$Z = 2e^{\beta 8J} + 2e^{-\beta 8J} + 12 = 4(\cosh(8\beta J) + 3) \quad (\text{A1})$$

Additionally we will calculate a few expectation values for which the general formula is given as

$$\langle A \rangle = \sum_s A_s p(s). \quad (\text{A2})$$

This is a sum over all spin states s_i . Here $p(s)$ is a chosen probability distribution, in our this case the Boltzmann distribution in eq. 7.

$$\langle E \rangle = \frac{16J}{Z} (e^{-\beta 8J} - e^{\beta 8J}) \quad (\text{A3})$$

$$\langle E^2 \rangle = \frac{128J^2}{Z} (e^{-\beta 8J} + e^{\beta 8J}) \quad (\text{A4})$$

$$\langle \epsilon \rangle = \frac{4J}{Z} (e^{-\beta 8J} - e^{\beta 8J}) \quad (\text{A5})$$

$$\langle \epsilon^2 \rangle = \frac{32J^2}{Z} (e^{-\beta 8J} + e^{\beta 8J}) \quad (\text{A6})$$

$$\langle |M| \rangle = \frac{8J}{Z} (e^{\beta 8J} + 2) \quad (\text{A7})$$

$$\langle M^2 \rangle = \frac{8J}{Z} (e^{\beta 8J} + 1) \quad (\text{A8})$$

$$\langle |m| \rangle = \frac{2}{Z} (e^{\beta 8J} + 2) \quad (\text{A9})$$

$$\langle m^2 \rangle = \frac{2}{Z} (e^{\beta 8J} + 1) \quad (\text{A10})$$

NB! Missing analytic C_V and X

Appendix B: Possible ΔE values

$$\begin{array}{c} \uparrow \\ \uparrow \uparrow \uparrow \\ \uparrow \end{array}$$

Appendix C: Metropolis algorithm flow chart