FYS4150 Project 4: The Ising model in 2 dimensions

Marie Foss (# 56), Maria Hammerstrøm (# 59)

Abstract

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Github: https://github.com/mariahammerstrom/Project4

1 Introduction

The project deals with the Ising model in two dimensions, without an external magnetic field. The Ising model is a model to study phase transitions at finite temperature for magnetic systems. In its simplest form the **energy** for a specific microstate i is expressed as

$$E_i = -J \sum_{\langle kl \rangle}^N s_k s_l,\tag{1}$$

with $s_k = \pm 1$, N is the total number of spins and J is a coupling constant expressing the strength of the interaction between neighboring spins. The symbol < kl > indicates that we sum over nearest neighbors only. We will assume that we have a ferromagnetic ordering, meaning J > 0. We will use periodic boundary conditions and the Metropolis algorithm only.

We will be using the **Boltzmann probability distribution** (PDF) defined as

$$w_i = \frac{1}{Z} e^{-\beta E_i},\tag{2}$$

where $\beta = 1/kT$ with k the Boltzmann constant and T the temperature, and with Z as the **partition function** for the canonical ensemble defined as

$$Z = \sum_{i=1}^{\infty} e^{-\beta E_i}.$$
 (3)

summing over all micro states i. w_i expresses the probability of finding the system in a given microstate i.

The magnetic moment of a given microstate is

$$M_i = \sum_j s_j. (4)$$

Some quantities we are interested in calculating, are the **expection values** for the energy $\langle E \rangle$ and magnetic moment $\langle M \rangle$:

$$\langle E \rangle = \frac{1}{Z} \sum_{i} E_{i} e^{-\beta E_{i}}$$

$$\langle M \rangle = \frac{1}{Z} \sum_{i} M_{i} e^{-\beta E_{i}},$$
(5)

as well as the variances for the energy σ_E^2 and for the magnetic moment σ_M^2 :

$$\sigma_E^2 = \langle E^2 \rangle - \langle E \rangle^2 \sigma_M^2 = \langle M^2 \rangle - \langle M \rangle^2,$$
 (6)

where

$$\sigma_E^2 = \frac{1}{Z} \sum_{i} E_i^2 e^{-\beta E_i} - \left[\frac{1}{Z} \sum_{i} E_i e^{-\beta E_i} \right]^2$$
 (7)

First we will assume that we only have two spins in each dimension, that is L=2, where L is the lattice length. In this case, we can find a closed form expression for the partition function in Eq. (3) and the corresponding expectation values for E, $|\mathcal{M}|$, the specific heat C_V and the susceptibility χ as functions of T using periodic boundary conditions. **Periodic boundary conditions** means that the neighbor to the right of a given spin s_N in Eq. (1) takes the value of s_1 . Similarly, the neighbor to the left of s_1 takes the value of s_N .

In the case L = 2 we can write Eq. (1) as:

$$E_{i} = -J \sum_{\langle kl \rangle}^{N} s_{k} s_{l} = -J \sum_{k=1}^{2} \sum_{l=1}^{2} s_{k} s_{l} = -J(s_{1} s_{1} + s_{1} s_{2} + s_{2} s_{1} + s_{2} s_{2})$$
(8)

... EXPRESSIONS!

2 Methods

We write a code for the Ising model which computes the mean energy E, mean magnetization $|\mathcal{M}|$, the specific heat C_V and the susceptibility χ as functions of T using periodic boundary conditions for in the x and y directions. Using the Ising model in two dimensions, the number of configurations is given by 2^N with $N = L \times L$ number of spins for a lattice of length L.

We will use the **Metropolis algorithm**. The algorithm goes as follows:

- 1. Generate a random configuration in the lattice to create an initial state with energy E_b .
- 2. Change the initial configuration by flipping for example one spin only. Compute the energy of this state E_t .
- 3. Calculate $\Delta E = E_t E_b$.
- 4. The Metropolis test: If $\Delta E \leq 0$ the new configuration is accepted, meaning the energy is lowered and we are moving towards the energy minimum at a given temperature. If $\Delta E > 0$, calculate $w = e^{-\beta \Delta E}$. If w < r where r is a random number, accept the new configuration. If else, keep the old configuration.
- 5. Compute the new energy $E' = E_t 0 \Delta E$. Calculate various expectation values using this energy.
- 6. Repeat steps 2-5 for the chosen number of Monte Carlo cycles, meaning the number of times we should sweep through the lattice and summed over all spins.
- 7. Compute the various expectation values by dividing by the total number of cycles (and possibly the number of spins).

3 Results

Our numerical results from computing the mean energy E, mean magnetization $|\mathcal{M}|$, the specific heat C_V and the susceptibility

 χ as functions of T using periodic boundary conditions, can be compared with expressions given in Eq. (8) - (?) for a temperature T=1.0 (in units of kT/J). This gives the following results:

Quantity	Closed form	Numerical
Mean energy E	-	-
Mean magnetization $ \mathcal{M} $	-	-
Specific heat C_V	-	-
Susceptibility χ	-	-

The number of Monte Carlo cycles needed to achieve good agreement is XXX.

Next we looked at the case of L=20 spins in the x and y directions. In this case we wanted to perform a study of the time (or number of Monte Carlo cycles) we need before reaching an equilibrium situation and can start computing the various expectation values.

A rough and study is done by plotting the various expectation values as a function of number of Monte Carlo cycles, for both T = 1.0 kT/J and T = 2.4 kT/J, as shown in Fig. XXX. PLOT + COMMENTS!

We also study the total number of accepted configurations as a function of the total number of Monte Carlo cycles, shown in Fig. XXX. PLOT + COMMENTS!

4 Conclusions

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5 List of codes

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