

2D and 3D Ising model implementation and Analysis

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(Dated: June 1, 2016)

We will simulate the 2d ising model and 3d ising model. From the result, we observe that the critical temperature is 2.3K. blabla, Chen will finish it later.

I. Overview

1. Introduction of Ising model

1.1. Ising Model

Ising model is a mathematical model to describe the ferromagnetism property of materials. It is first invented by the physicist Wilhelm Lenz, and is then gave to his student Ernst Ising, after whom the model is named, as a problem.

1.2. Ferromagnetism

Ferromagnetism is a basic mechanism of certain materials (such as iron), that forms permanent magnets attracted to magnets. Ferromagnetism can be divided into several distinguished types, such as ferrimagnetism, which is the strongest one, and some types response weakly, such as paramagnetism, diamagnetism and anti-ferromagnetism. Ferromagnetism describes the chemical make-up, crystalline structure and also microstructure of materials, and it arises due to two effects from quantum mechanics: spin and the Pauli Exclusion Principle. Generally speaking, the ferromagnetism of materials come from the spin property of electrons. Electrons has a quantum mechanical spin, which arises the magnetic dipole moment of it. The spin of the electron can only be in two states, either with magnetic field pointing "up" or "down", and it is the mainly source of ferromagnetism. When these magnetic dipoles pointing in the same direction, then the tiny magnetic fields add together to a much larger macroscopic field. And for materials made of atoms with filled electron shells, the magnetic moment of every electron is cancelled by the opposite moment of the second electron in the pair, such result in a total dipole moment of zero. So, only atoms with unpaired spins can have a net magnetic moment. So only materials with partially filled shells have ferromagnetism.

II. 2D Ising model

1. Basic Idea

Consider a d-dimensional lattice, each lattice site $k \in \Lambda$ is a discrete variable which indicate the spin state of the site. There is an interaction J_{ij} between any two adjacent sites $i, j \in \Lambda$, and for each site $j \in \Lambda$, there is an external magnetic field h_j interacting with it. The energy is approximated using the Hamiltonian equation:

$$\mathcal{H}(\sigma) = - \sum_{\langle i, j \rangle} J_{ij} \sigma_i \sigma_j - \mu \sum_j h_j \sigma_j \quad (1)$$

Here we ignore the external magnetic term, so the Hamiltonian equation becomes

$$\mathcal{H}(\sigma) = - \sum_{\langle i, j \rangle} J_{ij} \sigma_i \sigma_j \quad (2)$$

2. 2D Model overview

2.1. Hamiltonian

Consider the lattice Λ , denote the lattice site σ_{ij} to be the site in location (i, j) , and the Hamiltonian is

$$\mathcal{H}(\sigma) = - \sum_{\langle (i, j), (i', j') \rangle} J_{(i, j), (i', j')} \sigma_{ij} \sigma_{i'j'} \quad (3)$$

Updating for flipping the spin in site (i, j) will involve sites around it, which are $(i-1, j)$, $(i+1, j)$, $(i, j-1)$, $(i, j+1)$. So when updating one single site $\sigma_{i,j}^{new} = -\sigma_{i,j}$, the change in Hamiltonian will be

$$\Delta \mathcal{H} = 2(\sigma_{i-1,j} + \sigma_{i+1,j} + \sigma_{i,j-1} + \sigma_{i,j+1}) \sigma_{i,j} \quad (4)$$

2.2. Updating Algorithm - One Sweep

1. For each site in the lattice, calculate the $\Delta \mathcal{H}$ according to flipping the spin of this site.
2. If $\Delta \mathcal{H} < 0$, accept the flip, update the spin state.
3. If $\Delta \mathcal{H} > 0$, flip the spin state of site (i, j) with probability $\exp(\Delta \mathcal{H}/kT)$. Which is to generate a random number between $[0, 1]$. If the random number

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if less than the probability, then accept the update, if not, reject the update.

Note: Using periodic boundary conditions.

2.3. Updating Algorithm - Monte Carlo Process

1. Perform N_{therm} sweeps for warm up, to reach a relatively stable status
2. Perform N_{meas} sweeps for measurement, take each 100 record as a measurement.
3. Output the data and perform the calculation (some expectation values).

2.4. Critical Temperature

For 2D Ising model, there is a critical temperature, at which the material will have a transition in phase. We are using two approaches to obtain the critical temperature here:

1. Energy - Temperature figure
Observe the expectation value of energy according to the temperature, and the critical temperature should locate in where the curve has the greatest slope. In which

$$\langle E \rangle = \frac{1}{2} \langle \sum_i^N \mathcal{H}_i \rangle = \frac{1}{2} \langle -J \sum_i^N \sum_{j_{nn}} s_i s_j \rangle$$

2. Heat Capacity - Temperature plot
Observe the heat capacity according to the temperature, and the critical temperature should locate in where the heat capacity has the greatest value. In which

$$C = \frac{\partial E}{\partial T} = \frac{\Delta E^2}{k_b T} = \frac{\langle E^2 \rangle - \langle E \rangle^2}{k_b T^2}$$

3. Analytic solution

3.1. Partition Function

Now discuss the Analytic solution of the Ising model in a square lattice Λ with N sites, and using periodic boundary condition, which is exactly what we are using in our experiment.

Denote J_h and J_v to be number of horizontal and vertical coupling respectively. And denote

$$K = \beta J_h \quad L = \beta J_v$$

In which, $\beta = 1/(kT)$, and T is the absolute temperature and k is the Boltzmann's constant. Thus the partition

function can be written as

$$Z_N(K, L) = \sum_{\{\sigma\}} \exp \left(K \sum_{\langle i, j \rangle_H} \sigma_i \sigma_j + L \sum_{\langle i, j \rangle_V} \sigma_i \sigma_j \right) \quad (5)$$

3.2. Critical Temperature

The analytical critical temperature T_c is obtained using the Kramers-Wannier duality relation, which in this case gives

$$\beta F(K^*, L^*) = \beta F(K, L) + \frac{1}{2} \log[\sinh(2K) \sinh(2L)]$$

In which, $F(K, L)$ is the free energy per site, and also have the following relations

$$\begin{aligned} \sinh(2K^*) \sinh(2L) &= 1 \\ \sinh(2L^*) \sinh(2K) &= 1 \end{aligned}$$

And because there is only one critical temperature, so that use the assumption that there is only one critical line in the (K, L) plane, the duality relation above implies that

$$\sinh(2K) \sinh(2L) = 1$$

And for our square lattice, $J_h = J_v$, which is the isotropic case, the relation for critical temperature T_c is

$$\frac{kT_c}{J_h} = \frac{2}{\ln(1 + \sqrt{2})} \approx 2.26918531421$$

4. Numerical result

————— - TODO —————

III. 3D Ising model

In previous section, we introduced the general idea of 2D Ising model, which is based on a 2D plane. In this section, we are going to talk about 3D Ising model, which as the name suggested, is based on 3D space. We will perform our experiment in a square cube lattice, which contain N sites. Ising model is originally designed to describe the magnetism of materials, so we'll also assuming our system is a magnetization system, and examine some related physical.

1. 3D Model overview

1.1. Hamiltonian

In 3D case, the Hamiltonian can be written as

$$\begin{aligned} \mathcal{H} = -J \sum_{i,j,k=1}^N & (S_{i-1,j}S_{i,j} + S_{i,j}S_{i+1,j} \\ & + S_{i,j-1}S_{i,j} + S_{i,j}S_{i,j+1} + S_{i-1,k}S_{i,k} \\ & + S_{i,k}S_{i+1,k} + S_{i,k-1}S_{i,k} + S_{i,k}S_{i,k+1} \\ & + S_{j-1,k}S_{j,k} + S_{j,k}S_{j+1,k} + S_{j,k}S_{j,k-1} \\ & + S_{j,k}S_{j,k+1}) - \mathbf{H} \sum_i^N \mathbf{s}_i \end{aligned} \quad (6)$$

Where \mathbf{H} is the external force, for simplicity, we ignore the external force in our experiment. So that $\mathbf{H} = 0$.

1.2. Magnetic Property

The magnetization is denoted as M , which of a given configuration can be given by

$$M = \frac{1}{N} \sum_i^N S_i \quad (7)$$

The magnetic susceptibility is denoted as χ , and is calculated as

$$\chi = \frac{1}{kT} (\langle M^2 \rangle - \langle M \rangle^2) \quad (8)$$

In which, k is Boltzmann's constant, and T is the temperature of the system.

2. Numerical result

—————- TODO —————

IV. parallel(cuda, matlab)

1. CUDA

Numerical approach of Ising model requires computation on each lattice site, which when lattice becomes really large scaled will become very computationally heavily. So we want to find a way to improve the computation efficiency and reduce the computation time.

Because calculation for one site only involves adjacent sites, so the basic idea of improve computation efficiency is to separate the whole space into two parts, and calculation for one part depend on another part, and do not depend on sites inside this part, and calculate each parts in turn. For example in 2D case, we can calculate for odd sites using the previous values of even sites, and then calculate for even sites using the updated odd sites, iteratively perform the steps until convergence.

—————- figures here for explanation —————

Following this idea, when computing for one part, if the computation can be paralleled, this will largely decrease computation time, so we are using CUDA to enable parallel computation for different sites. The updating algorithm is exactly the same as previously stated, the only difference is that without parallelization, each iteration requires N sweeps for each sites in the lattice, while using parallelization will compress number of sweeps each iteration.

2. Matlab

—————- TODO —————

V. Conclusion

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