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Tutorial: Monte Carlo Simulation of the 2D Ising Model

August 26, 2019

In this tutorial, we will study the phase transition in the classical two-dimensional Ising model, with Hamiltonian

$$H = -J \sum_{\langle ij \rangle} \sigma_i \sigma_j,$$

where $\sigma_i = \pm 1$, J is the coupling strength and $\sum_{\langle ij \rangle}$ denotes a sum over nearest neighbours. We will consider simulations on a square lattice with periodic boundaries. In the thermodynamic limit, the critical temperature is known to be $T_c/J \approx 2.269$.

We will use and modify the two Python programs ising_mc.py and plot_ising.py throughout this tutorial in order to implement Monte Carlo (MC) methods that estimate T_c and compare with this known exact solution.

1 Monte Carlo algorithm

Consider the Monte Carlo program $ising_mc.py$, which is designed to perform a Monte Carlo simulation (using the single-spin-flip Metropolis algorithm) and record measurements of the system's energy E and magnetization M.

a) Examine the section of the code that computes the two-dimensional neighbours array, which is used when calculating the system's energy. The code is already written such that neighbours[i,0] and neighbours[i,1] store the lattice location of spin i's rightward and upward neighbours, respectively. Modify the code such that it will also store spin i's leftward neighbour in neighbours[i,2] and its downward neighbour in neighbours[i,3].

Hint: Don't forget to consider the periodic boundary conditions.

- b) Examine the sweep() function, which proposes a number N_spins single spin-flip Monte Carlo updates. Convince yourself that this code is implementing the single-spin-flip Metropolis algorithm.
- c) Implement a more efficient way of calculating the energy difference deltaE within the sweep() function. The given implementation calculates this energy difference by using the getEnergy() function, which involves iterating a loop N_spins times. However, you should be able to calculate deltaE by summing only four terms.

d) Run your code with n_eqSweeps=1000 and n_measSweeps=10000. The code will generate files in a directory called Data that will store the energy and magnetization corresponding to each of your sampled system configurations. (In Question 2, we will analyze and plot the resulting data.)

Hint: Set animate = False for this part so that the code runs faster.

2 Estimating the critical temperature

Recall from lecture that the specific heat C_V and susceptibility χ can be expressed as

$$C_V = \frac{\langle E^2 \rangle - \langle E \rangle^2}{T^2} \,, \qquad \qquad \chi = \frac{\langle M^2 \rangle - \langle M \rangle^2}{T} ,$$

where E is the energy and $M = \sum_i \sigma_i$ is the magnetization. For our Monte Carlo calculations on finite lattices, there is no spontaneous symmetry breaking and therefore $\langle M \rangle = 0$ at all T. As a result, we instead examine $\langle |M| \rangle$ and calculate the susceptibility as

$$\chi = \frac{\langle M^2 \rangle - \langle |M| \rangle^2}{T}.$$

The quantities C_V/N versus T and χ/N versus T both diverge at the critical temperature T_c in the thermodynamic limit $N \to \infty$. On a finite lattice, these quantities do not diverge but will acquire peaks near T_c .

- a) Use the code plot_ising.py to read in the Monte Carlo data from Question 1e and plot the estimators for $\langle E \rangle/N$ and $\langle |M| \rangle/N$. Consider the values you find for these estimators in the limit of small T; do they match your theoretical expectations?
- b) Modify plot_ising.py to calculate C_V and χ . Plot C_V/N and χ/N versus T and verify that there are peaks in these quantities near T_c .
- c) Use ising_mc.py to generate additional data for higher L and for more temperatures close to $T_{\rm c}$. Modify plot_ising.py to plot your results for several different values of L and confirm that the peaks in C_V/N and χ/N get closer to $T_{\rm c}$ as L increases.