FYS4150: Project 4

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1 GJENSTÅR

- Finn histogram av ENERGI, ikke forventningsverdi!!! 4d
- Test verdiene med variansen; hva slags underliggende sannsynlighetsfordeling er dette?
- parallelliser!

2 Introduction

The aim of this project is to study the Ising model in two dimensions.

3 The Ising model (ch. 13)

The Ising model describes palse transitions in two dimensions. At a given critical temperature the model exhibits a phase transition from one magnetic moment (spin) to a phase with zero magnetization. Without an externally applied field, this is described as

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

where $s_k = \pm 1$, N is the total number of spins and J is a coupling constant.

4 The Metropolis algorithm (ch. 12)

5 2x2 lattice

We consider a 2×2 -lattice, meaning L = 2. The spins have 16 configurations. The probability distribution is given by

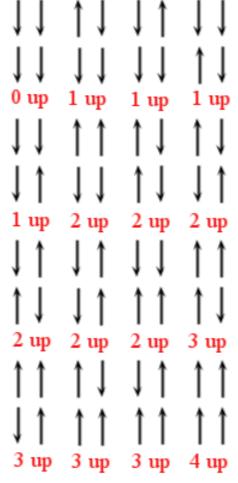
$$P_i(\beta) = \frac{e^{-\beta E_i}}{Z} \tag{2}$$

where $\beta = \frac{1}{kT}$ is the inverse temperature, k is the Boltzmann constant, E_i is the energy of a state i. The partition function is

$$Z = \sum_{i=1}^{M} e^{-\beta E_i} \tag{3}$$

This fuction is difficult to compute since we need all states. The Metropolis algorithm only considers *ratios* between probabilities, so we don't need to calculate this (luckily!)

Metropolis
$$\rightarrow \frac{P_s}{P_t} = \frac{e^{-(\beta E_s)}}{e^{-(\beta E_k)}} = e^{-\beta (E_s - E_k)} = e^{-\beta \Delta E}$$



$\mathcal{N}_{\underline{0}}$ of spins	Degeneracy	E	M
pointing up			
0	1	-8J	-4
1	4	0	-2
2	4	0	0
2	2	8J	0
3	4	0	2
4	1	-8J	4

(b) Possible energies for a 2×2 -lattice of spin particles.

(a) Different spin configurations

Figure 1: Spins for a 2×2 -lattice with periodic boundary conditions.

The partition function in our case is

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

$$= 2e^{8J\beta} + 2e^{-8J\beta} + 12$$

$$= 4(3 + \cosh(8J\beta))$$

The energy expectation value

$$\begin{split} \langle E \rangle &= -\frac{\partial \ln Z}{\partial \beta} = -\frac{\partial \ln Z}{\partial Z} \frac{\partial Z}{\partial \beta} = -\frac{1}{Z} \frac{\partial Z}{\partial \beta} \\ &= -\frac{4 \cdot 8J \sinh(8J\beta)}{4(3 + \cosh(8J\beta))} \\ &= -8J \frac{\sinh(8J\beta)}{3 + \cosh(8J\beta)} \end{split}$$

with the corresponding variance

$$\sigma^{2} = \langle E^{2} \rangle - \langle E \rangle^{2} = \frac{1}{Z} \sum_{i=1}^{M} E_{i}^{2} e^{-\beta E_{i}} - \left(\frac{1}{Z} \sum_{i=1}^{M} E_{i} e^{-\beta E_{i}}\right)^{2}$$

$$= \frac{8^{2} J^{2} 2 e^{8J\beta} + 8^{2} J^{2} 2 e^{-8J\beta} + 0 \cdot 12}{4(3 + \cosh(8J\beta))} - \left(-8J \frac{\sinh(8J\beta)}{3 + \cosh(8J\beta)}\right)^{2}$$

$$= 32J^{2} \frac{2 \cosh(8J\beta)}{(3 + \cosh(8J\beta))} - 64J^{2} \frac{\sinh^{2}(8J\beta)}{(3 + \cosh(8J\beta))^{2}}$$

$$= J^{2} 64 \frac{\cosh(8J\beta)(3 + \cosh(8J\beta)) - \sinh^{2}(8J\beta)}{(3 + \cosh(8J\beta))^{2}}$$

$$= J^{2} 64 \frac{3 \cosh(8J\beta) + \cosh^{2}(8J\beta) - \sinh^{2}(8J\beta)}{(3 + \cosh(8J\beta))^{2}}$$

$$= J^{2} 64 \frac{3 \cosh(8J\beta) + 1}{(3 + \cosh(8J\beta))^{2}}$$

The specific heat capacity

$$C_V = \frac{1}{k_B T^2} (\langle E^2 \rangle - \langle E \rangle^2) \tag{4}$$

$$C_V = \frac{64J^2}{k_B T^2} \frac{3\cosh(8J\beta) + 1}{(3 + \cosh(8J\beta))^2}$$

The mean absolute value of the magnetic moment (mean magnetization)

$$\langle |\mathcal{M}| \rangle = \frac{1}{Z} \sum_{i}^{M} |\mathcal{M}_{i}| e^{-\beta E_{i}}$$

$$= \frac{1}{4(3 + \cosh(8J\beta))} (4e^{8J\beta} + 4 \cdot 2e^{0} + 2 \cdot 4e^{0} + 4e^{8J\beta})$$

$$= 2 \frac{e^{8J\beta} + 2}{3 + \cosh(8J\beta)}$$

The susceptibility

$$\chi = \frac{1}{k_B T} (\langle \mathcal{M}^2 \rangle - \langle \mathcal{M} \rangle^2) \tag{5}$$

$$\chi = \frac{1}{k_B T} \left(\frac{1}{Z} \sum_{i}^{M} \mathcal{M}_i^2 e^{-\beta E_i} - \left(\frac{1}{Z} \sum_{i}^{M} \mathcal{M}_i e^{-\beta E_i} \right)^2 \right)$$

$$= \frac{1}{k_B T} \frac{\left(16 e^{8J\beta} + 4 \cdot 4 e^0 + 4 \cdot 4 e^0 + 16 e^{8J\beta} \right)}{4(3 + \cosh(8J\beta))}$$

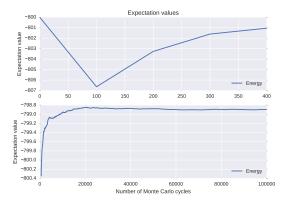
$$= \frac{1}{k_B T} \frac{8 e^{8J\beta} + 8}{(3 + \cosh(8J\beta))}$$

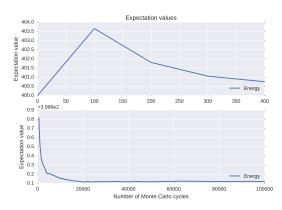
The analytical values (calculated in Python) are found in table (??).

Z	$\langle E \rangle$	$ \langle \mathcal{M} \rangle$	C_{V}	\sim
	\2/	(\cup_{V}	Λ
5973.917	-7.9839	3 9946	0.1283	15 9732
0010.011	1.0000	0.0010	0.1200	10.0102

Figure 2: Analytical values of the expectation values. Here $J=1,\,k=1,\,T=1.0$ K.

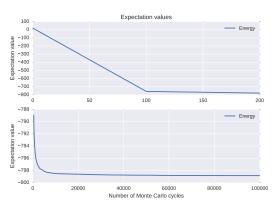
For mcs = 1000000 cycles the results are very similar to the analytic results.

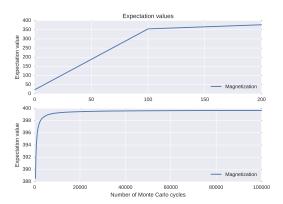




- (a) Expectation value energy.
- (b) Expectation value magnetization.

Figure 3: 20×20 -lattice with temperature T = 1, ordered initial configuration.





- (a) Expectation value energy.
- (b) Expectation value magnetization.

Figure 4: 20×20 -lattice with temperature T = 1, random initial configuration. Used 10000 points.

6 20x20 lattice

Setting temperature to T=1 and running a 20×20 lattice for a different number of Monte Carlo cycles ranging from 1 to 10^{-7} . We see from figure (3) that for 1 cycle energy and magnetization are exactly -800 and 400. This is because when one random spin is flipped the probability of keeping this energy configuration is proportional to $e^{-8}\simeq 0.0003$. As we add more cycles, however, the probability of some spins using their new configuration becomes larger. Both expectation values seem to stabilize at around 10^5 .

We also look at a system that begins with a random distribution of spins. For a 20×20 -lattice this will probably begin with energy 0 and magnetization 0. It will the 'tip' to one of the sides, and converge towards -800 and 400. The results are shown in fig (4).

We also run the program for temperature T=2.4. The stable energy and magnetization are now smaller in norm.

7 Analyzing the probability distribution

We find the probability of a state by counting the number of states in an energy E, and dividing by the total number of configurations

$$P(E) = \frac{N(E)}{cms} \tag{6}$$

where cms is the number of Monte Carlo cycles used. We see from the results in Fig. (??) that the probability distribution is much narrower for T = 1.1.

After running 10^5 monte carlo cycles, the variance is for 1T: $\sigma_E = -3.58476$, for T = 2.4: $\sigma_E = 3161.7$.

8 Sources

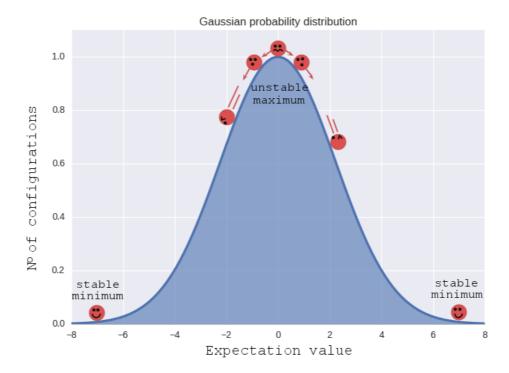


Figure 5: Expectation configuration distribution

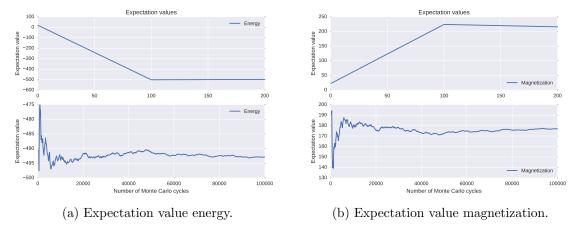


Figure 6: 20×20 -lattice with temperature T=2.4, random initial configuration. Used 10000 points.

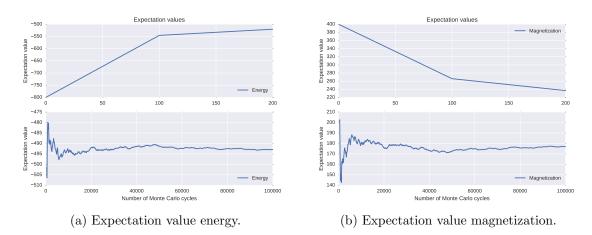


Figure 7: 20×20 -lattice with temperature T=2.4, ordered initial configuration. Used 10000 points.

