Project 4 FYS4150

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Abstract

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1 Introduction

2 Theory

2.1 Ising model

The two dimensional Ising model is a statistical model that allows us to investigate the temperature dependence of different properties of a magnet. The model consists of a two dimensional lattice of spins that can be in two different states, spin up, \uparrow , or spin down, \downarrow [1]. In our system we do not have an applied field, $B_a = 0$.

The energy of our system is then:

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

where in our system J > 0, giving a ferromagnetic ordering, and s_k , $s_l = \pm 1$.

2.2 Introduction to statistics

The methods in this project require some statistics. Equation 2 shows the probability density and in the Ising model, the probability distribution is Boltzmann distribution (Equation 3) where Z is the partition function that normalizes the probability.

$$P(a \le X \le b) = \int_{b}^{a} p(x) dx \tag{2}$$

$$P_i(\beta) = \frac{e^{-\beta E_i}}{Z} \tag{3}$$

$$Z = \sum_{i=1}^{M} e^{-\beta E_i} \tag{4}$$

$$\langle h \rangle_X = \int h(x) p(x) dx$$
 (5)

$$\langle x^n \rangle = \int x^n \, p(x) \, dx \tag{6}$$

$$\langle x \rangle = \int x \, p(x) \, dx$$

Partition function

Boltzman

2.3 Magnetic properties

Energy, magnetic moment, susceptibility, heat capacity.

$$T_C(L) - T_C(L = \infty) = aL^{-1/\nu}$$
 (8)

Markhow chain - convergance

2.4 Analytical solutions for L=2

L=2 case:

Table 2.1: text

No spin up	Deg	Energy	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

Error Random number

The partition function:

$$Z = \sum_{i}^{M} e^{-\beta E_i} = e^{-\beta 8J} + e^{-\beta 8J} + e^{\beta 8J} e^{\beta 8J} + 12$$

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

The energy:

$$\langle E \rangle = k_B T^2 \left(\frac{\partial Z}{\partial T} \right)_{V,N}$$

$$= k_B T^2 \frac{\partial}{\partial T} \left[\ln \left(4 \cosh \left(\frac{8J}{k_B T} \right) + 12 \right) \right]$$

$$\frac{\partial \ln Z}{\partial T} = \frac{\partial Z}{\partial \beta} \frac{\partial \beta}{\partial T} = \frac{\partial \ln Z}{\partial \beta} \left(\frac{-1}{k_B T^2} \right)$$

(7)
$$\langle E \rangle = -\left(\frac{\partial Z}{\partial \beta}\right)_{VN} = -\frac{\partial}{\partial \beta} \ln\left[4\cosh\left(8J\beta\right) + 12\right]$$

$$= \frac{-1}{4\cosh(8J\beta) + 12} 4\sinh(8J\beta)8J\beta$$
$$= \frac{-8J\sinh(8J\beta)}{3\cosh((J\beta) + 4}$$

Following the same method, we found that:

$$\langle |M| \rangle = \frac{1}{Z} \sum_{i}^{M} M_{i} e^{\beta E_{i}} = \frac{(8J)^{2} \cosh(8J\beta)}{\cosh(8J\beta) + 3}$$

$$\langle M \rangle = 0$$

$$\langle E^2 \rangle = \frac{8 \left(e^{8J\beta} + 1 \right)}{\cosh(8J\beta) + 3}$$

$$\left\langle M^{2}\right\rangle =\frac{1}{Z}\left(\sum_{i}^{M}M_{i}^{2}e^{\beta E_{i}}\right)=\frac{2\left(e^{8J\beta}+2\right)}{\cosh(8J\beta)+3}$$

We can use these to calculate the rest:

$$C_V = k\beta^2 \left(\left\langle E^2 \right\rangle - \left\langle E \right\rangle^2 \right)$$

$$\chi = \beta \left(\left\langle M^2 \right\rangle - \left\langle M \right\rangle^2 \right)$$

3 Method

In this project we tried out many new concepts in our algorithm. We used Monte Carlo method.

3.1 Monte Carlo cycles

In Monte Carlo methods, the goal is to

- 3.2 Metropolis algorithm
- 3.3 Random number generator
- 3.4 Parallelizing

3.5 Unit tests

Check the L=2 result with the analytical one - We did it visually

Table 4.1: This table compares the analytical values for L=2 with the numerical ones after 10^6 Monte Carlo cycles. The values are in units per spin.

	Numerical:	Analytical:
$\overline{\langle E \rangle}$	-1.9958	-1.9960
$\langle E^2 \rangle$	15.9664	15.9679
$\langle M \rangle$	0.0451	0
$\langle M^2 \rangle$	3.9930	3.9933
$\langle M \rangle$	0.9986	0.9987
χ	3.9849	3.9933
C_V	0.0335	0.0321

Make a small matrix (ordered initial), calculate energy - flip one spin, calculate energy - is the change in energy what we expect?

- Should have had it underveis - Kan neste gang lage tester som gjør at når vi implementerer nye ting som for eksempel classes eller paralelleisering, vet at det ikke har skjedd noe galt.

Metropolis (T,A,...) Stochastic matrix - convergences (forhold eigenvalue).

Hvilken random number engine

MPI:

- Develop codes locally, run with some few processes and test your codes. Do benchmarking, timing and so forth on local nodes, for example your laptop or PC. - When you are convinced that your codes run correctly, you can start your production runs on available supercomputers.

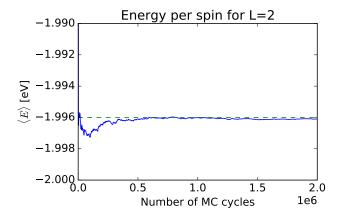
MPI functions:

4 Result

4.1 Matrix dimension L=2

OBS! Need an number of MC cycles necessary!

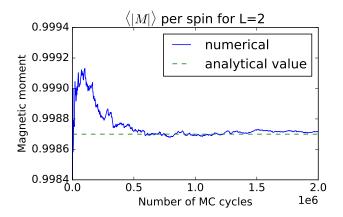
All calculations in this subsection are at T = 1.0 K.



Heat capacity per spin for L=2 0.038 0.036 0.034 0.032 $\sum_{0.032}^{0.032}$ 0.030 0.028 0.026 numerical 0.024 analytical value 0.022 1000000 1500000 2000000 500000 Number of MC cycles

Figure 4.1: This is a plot of the expectation value of the energy per spin verus number of Monte Carlo cycles. The plot shows that we have a good agreement after $5 \cdot 10^5$ MC cycles.

Figure 4.3: This is a plot of the heat capacity per spin verus number of Monte Carlo cycles. The plot shows that we have a good agreement after $5 \cdot 10^5$ MC cycles.



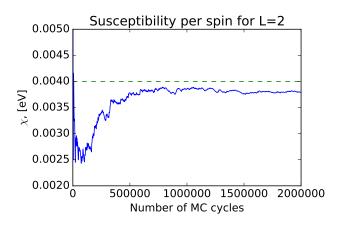


Figure 4.2: This is a plot of the expectation value of the mean absolute value of the magnetic moment per spin verus number of Monte Carlo cycles. The plot shows that we have a good agreement after $5 \cdot 10^5$ MC cycles.

Figure 4.4: This is a plot of the susceptibility per spin verus number of Monte Carlo cycles. The plot shows that we have a good agreement after $5 \cdot 10^5$ MC cycles.

4.2 Matrix dimension L = 20

HMM: Should define an area that is enough for equilibrium!

OBS: Need the number of MC cycles to reach equilibrium!

OBS: Need equilibration time! (5 1e5?)

OBS: Comment accepted configs T dependency

4.2.1 Different temperatures

$_{-1}$ Mean energy and magnetic moment at T = 1.0_{11} -1.961.00 Energy (eV) 0.99 -1.97 $\langle M \rangle$ $\langle E \rangle$ -1.98-1.99____0.96 1.0 0.2 0.4 0.6 0.8 Monte Carlo cycles 1e6

Figure 4.5: This is a plot of both the expectation value of the energy and absolute magnetic moment per spin verus number of Monte Carlo cycles at T=1.0 K. The plot shows that an equilibrium is reached already at $2 \cdot 10^5$ MC cycles.

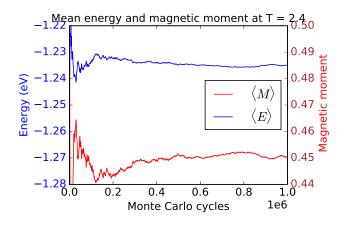


Figure 4.6: This is a plot of both the expectation value of the energy and absolute magnetic moment per spin verus number of Monte Carlo cycles at T=2.4 K. The plot shows that an equilibrium is reached at around $5 \cdot 10^5$ MC cycles.

4.2.2 Initial state

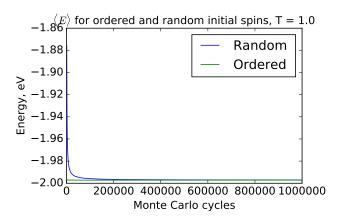


Figure 4.7: This is a plot of both the expectation value of the energy and absolute magnetic moment per spin verus number of Monte Carlo cycles at $T=1.0~\mathrm{K}$. The plot shows the difference in the behaviour of the ordered initial state and a random initial state.

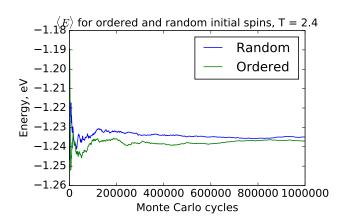


Figure 4.8: This is a plot of both the expectation value of the energy and absolute magnetic moment per spin verus number of Monte Carlo cycles at $T=2.4~\mathrm{K}$. The plot shows the difference in the behaviour of the ordered initial state and a random initial state.

4.2.3 Accepted configurations

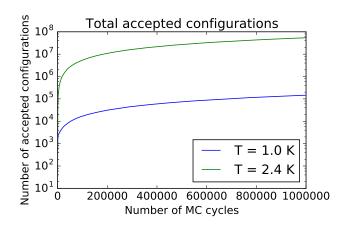


Figure 4.9: This is a plot of the total number of accepted configurations versus number of Monte Carlo cycles with random initial state.

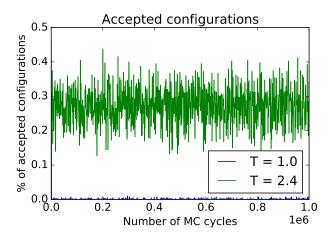


Figure 4.10: This is a plot of the percentage accepted of attempted configurations versus Monte Carlo cycles with random initial state.

4.3 Energy probability

OBS: Compare result with computed variance!

OBS: Discuss behavior (In Discussion - maybe just merge result and discussion?)

Computed variance (from same dataset?):

$$\sigma_E^2 = \left\langle E^2 \right\rangle - \left\langle E \right\rangle^2$$

$$\text{FWHM} = 2\sqrt{2ln2}\sigma \approx 2.355\sigma$$

T = 1.0 K:

$$\sigma_E^2 = 638181 - (-798.855)^2 = 11.69$$

$$\sigma = 3.42$$

$$\text{FWHM} \approx 2.355 \cdot 3.24 = 7.63$$

T = 2.4 K:

$$\sigma_E^2 = 247886 - (-494.628)^2 = 3229.14$$

$$\sigma = 56.8$$
EVALUATE: 2.255, 56.0, 122.76

 $FWHM \approx 2.355 \cdot 56.8 = 133.76$

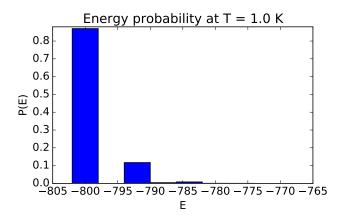


Figure 4.11: This is a plot of the energy probability when $T=1.0~\mathrm{K}$. The energy is the total energy of the 2D lattice with $20\times20~\mathrm{spins}$.

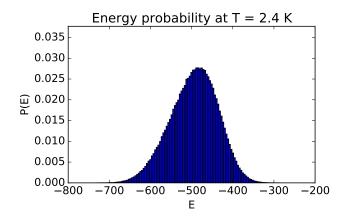


Figure 4.12: This is a plot of the energy probability when T=2.4 K.The energy is the total energy of the 2D lattice with 20×20 spins.

4.4 Increasing dimensionality/ Critical temperature

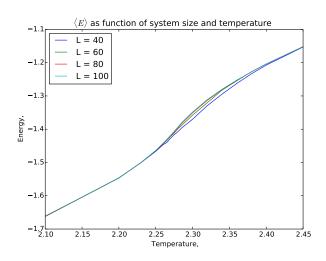


Figure 4.13: This is a plot if the energy versus temperature around the critical temperature for the different lattice sizes with $L=40,\,L=60,\,L=80$ and L=100.

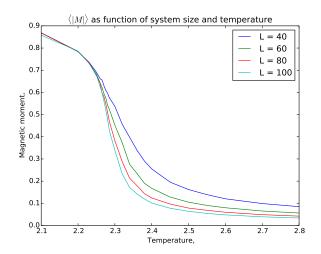


Figure 4.14: This is a plot if the absolute magnetic moment versus temperature around the critical temperature for the different lattice sizes with $L=40,\,L=60,\,L=80$ and L=100.

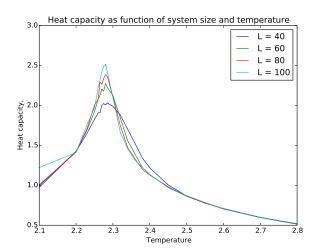


Figure 4.15: This is a plot if the heat capacity versus temperature around the critical temperature for the different lattice sizes with L=40, L=60, L=80 and L=100.

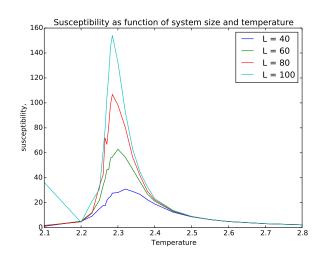


Figure 4.16: This is a plot if the susceptibility versus temperature around the critical temperature for the different lattice sizes with L = 40, L = 60, L = 80 and L = 100. The exact value is $T_C = kTC/J = 2/\ln(1+\sqrt{2}) \approx 2.269 \, k_B \, K$ [2].

OBS: Indication of phase transition? (Peak - at least for Cv and X)

OBS: Use Equation 8 to extract T_C .

Getting these equations from 8 where $\nu = 1$:

$$T_C(40) - T_C(\infty) = a \cdot 40^{-1}$$

$$T_C(60) - T_C(\infty) = a \cdot 60^{-1}$$

$$T_C(80) - T_C(\infty) = a \cdot 80^{-1}$$

$$T_C(100) - T_C(\infty) = a \cdot 100^{-1}$$

Table 4.2: L=80 and MC cycles is 1e6.

Number of processors:	CPU time [s]:
1	513.069
2	306.975

(Sett inn tall!)

$$T_C(\infty) = -a \cdot 40^{-1} + 2.28 \tag{9}$$

$$T_C(\infty) = -a \cdot 60^{-1} + 2.27 \tag{10}$$

$$T_C(\infty) = -a \cdot 80^{-1} + 2.28 \tag{11}$$

$$T_C(\infty) = -a \cdot 100^{-1} + 2.27 \tag{12}$$

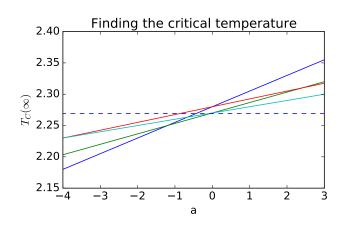


Figure 4.17: This is a plot of Equation 8 with different values of L (See Equations 9 - 12). The intersections represent the solution. They should have all had a cross section in the same place, and the y-value of the intersection would have been the critical temperature when $L \to \infty$.

Exact
$$T_C = kTC/J = 2/\ln(1+\sqrt{2}) \approx 2.269$$
 [2]

5 Discussion

6 Conclusion

References

- [1] Ising model. https://en.wikipedia.org/wiki/Ising_model. Accessed: 2017-11-17.
- [2] Lars Onsager. Crystal statistics. i. a two-dimensional model with an order-disorder transition. *Phys. Rev.*, 65:117–149, Feb 1944.

Appendix

State	Spinn	Energi	Magnetization
0	$\downarrow\downarrow\downarrow\downarrow\downarrow$	-8J	$\overline{-4}$
1	$\downarrow\downarrow\downarrow\downarrow\uparrow$	0	-2
2	$\downarrow\downarrow\uparrow\uparrow\downarrow$	0	-2
3	$\downarrow \uparrow \downarrow \downarrow$	0	-2
4	$\uparrow\downarrow\downarrow\downarrow$	0	-2
5	$\downarrow\downarrow\uparrow\uparrow\uparrow$	0	0
6	$\downarrow\uparrow\downarrow\uparrow$	0	0
7	$\downarrow\uparrow\uparrow\downarrow$	8J	0
8	$\uparrow\downarrow\downarrow\uparrow\uparrow$	8J	0
9	$\uparrow\downarrow\uparrow\downarrow$	0	0
10	$\uparrow \uparrow \downarrow \downarrow$	0	0
11	$\downarrow\uparrow\uparrow\uparrow$	0	2
12	$\uparrow\downarrow\uparrow\uparrow$	0	2
13	$\uparrow \uparrow \downarrow \uparrow$	0	2
14	$\uparrow\uparrow\uparrow\downarrow$	0	2
15	$\uparrow\uparrow\uparrow\uparrow$	-8J	4