Computational Physics - Project 4

Johannes Scheller, Vincent Noculak, Lukas Powalla, Richard Asbah November 10, 2015

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Introduction

In project 4 we are dealing with the Ising model in two dimensions without an external magnetic field. We are looking at a lattice of L times L particles, where each particle has a spin value of ± 1 . In order to compute different interesting properties of the system, we want to use the metropolis algorithm. With our computations, we want to calculate the Energy, the absolute value of the magnetisation, the heat capacity and susceptibility of the system as a function of temperature in order to study phase transitions of the system. In our case, we want to study the phase transition from a ordered phase for low temperatures to a disordered phase for high temperatures above the critical Temperature T_c (in a canonical ensemble this means a second order phase transition and therefore a divergence of the heat capacity at the critical Temperature). In order to get familiar with all the quantities, we first wan to study the case of a 2 times 2 lattice. For This lattice-size, we want to find a analytical expression for all interesting physical properties of the system and compare them with the computed values. However, researcher have solved the two dimensional Ising model a long time ago for any and even for infinite size. The Ising model in two dimensions has been solved first for any "fixed" size by Kaufman in 1949 and in the end even for a infinite size by Onsager in 1944. (compare with "Statistical Mechanics: Algorithms and Computations; Werner Krauth; published 2006") This project may also show the link from statistical physics to macroscopic properties of a given physical system, which is a interesting relation.

1 Theory

1.1 General properties of physical systems and their link to statistical physics

1.1.1 physical ensembles

Let us now look at a physical system and its surroundings. In principle, it is necessary to describe the relation of the physical system and its surroundings in order to determine the properties of the system. (sometimes this relations are related to physical boundary conditions) It is necessary to know whether we want to allow for instance particle/heat exchange or not. How we set up our system also defines us the thermodynamic potential, which can be used to describe the system. (e.g. Entropy, Helmholtz, Gibbs) All in all, we have the Microcanonical ensemble, the canonical ensemble and the pressure canonical ensemble. (in this case an ensemble means a collection of mircoscopic systems, compare with the lecture notes Computational physics 2015 at University of Oslo by Morten Hjorth-Jensen page 417) In the following, we will always deal with the canonical ensemble. This means that we don't allow particle exchange from the system with its surroundings, but we allow exchange of heat with the environment. Fixed variables are in this case the temperature, the total volume and the total particle number. The total energy of the canonical ensemble is not constant, because there can be heat exchange with the surroundings. The system, which does not allow heat exchange and does not allow particle exchange is called micro-canonical ensemble. (compare "Statistical Mechanics An Intermediate Course; 2nd Edition; G. Morandi/F. Napoli/ E. Ercolessi; page 94ff")

1.1.2 General properties of canonical ensembles

The canonical ensemble can be expressed by Helmholtz' free energy. The system strives to a minimum of Helmholtz' free energy, which is defined as follows:

$$F = -k_B T \ln Z = \langle E \rangle - T S \tag{1}$$

where the entropy S is given by

$$S = -k_B \ln Z + k_B T \frac{\partial \ln Z}{\partial T} \tag{2}$$

We can see that F depends on the expectation value of the Energy and on -TS. Hence, the canonical ensemble pursuits towards an energy minimum and higher entropy. This can be interpreted as a "struggle between two important principles in physics" (lecture notes Computational physics 2015 at University of Oslo by Morten Hjorth-Jensen page 419) The probability distribution for a canonical ensemble is given by the Boltzmann distribution.

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

 $\beta = 1/k_BT$ where T is the temperature, k_B is the Boltzmann constant, E_i is the energy of micro state i and Z is the partition function for the canonical ensemble is the sum over all the micro states M.

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

After running the system for long time the canonical ensemble is uniquely determined and does not depend on the arbitrary choices of the initial temperature. The system uncertainty due the Energy fluctuations in the canonical ensemble gives the variance of the energy.

from equation 1, 2 and probability distribution P_i

$$\langle E \rangle = k_b T^2 \frac{\partial \ln Z}{\partial T} = \sum_{i=1}^{M} E_i P_i(\beta) = \frac{1}{Z} \sum_{i=1}^{M} E_i e^{-\beta E_i}$$
 (5)

The heat capacity is how much the energy change due to the change in the temperature. The heat capacity C_V can be defined as

$$C_V = \frac{\partial E}{\partial T} \tag{6}$$

$$\frac{\partial}{\partial T} \frac{1}{Z} = \frac{\partial}{\partial T} \frac{1}{\sum_{i=1}^{M} e^{-\frac{1}{k_B T} E_i}} = -\frac{1}{k_B T^2} \frac{\sum_{i=1}^{M} E_i e^{-E_i \frac{1}{k_B T}}}{\left(\sum_{i=1}^{M} e^{-E_i \frac{1}{k_B T}}\right)^2} = -\frac{1}{k_B T^2} \frac{\sum_{i=1}^{M} E_i e^{-E_i \frac{1}{k_B T}}}{(Z)^2}$$
(7)

$$\frac{\partial}{\partial T} \sum_{i=1}^{M} E_i e^{-\frac{1}{k_B T} E_i} = \frac{1}{k_B T^2} \sum_{i=1}^{M} E_i^2 e^{-\frac{1}{k_B T} E_i}$$
(8)

$$C_{V} = \frac{\partial \langle E \rangle}{\partial T} = \frac{\partial}{\partial T} \left(\frac{1}{Z} \sum_{i=1}^{M} E_{i} e^{-\frac{1}{k_{B}T} E_{i}} \right) = -\frac{1}{k_{B}T^{2}} \frac{\sum_{i=1}^{M} E_{i} e^{-E_{i} \frac{1}{k_{B}T}}}{Z^{2}} \sum_{i=1}^{M} E_{i} e^{-\frac{1}{k_{B}T} E_{i}} + \frac{1}{Z} \frac{1}{k_{B}T^{2}} \sum_{i=1}^{M} E_{i}^{2} e^{-\frac{1}{k_{B}T} E_{i}}$$
(9)

$$= -\frac{1}{k_B T^2} \left(\frac{\sum_{i=1}^M E_i e^{-E_i \frac{1}{k_B T}}}{Z} \right)^2 + \frac{1}{Z} \frac{1}{k_B T^2} \sum_{i=1}^M E_i^2 e^{-\frac{1}{k_B T} E_i} = \frac{1}{k_B T^2} \left(\langle E \rangle^2 - \langle E^2 \rangle \right)$$
 (10)

The magnetic susceptibility is a measurable quantity, which indicates if the material is attracted or repelled of a magnetic field. Magnetic materials can be classified as paramagnetic, diamagnetic or ferromagnetic based on their susceptibility.

$$\chi = \frac{\partial < M >}{\partial H} \tag{11}$$

We can evaluate the mean magnetization through:

$$\langle M \rangle = \sum_{i}^{M} M_{i} P_{i}(\beta) = \frac{1}{Z} \sum_{i}^{M} M_{i} e^{-\frac{E_{i}}{k_{B}T}}$$
 (12)

The total energy of the system in addition of external magnetic field H can be described with:

$$E = -\sum_{i,j} J_{s_i s_j} - H \sum_{i} s_i$$
 (13)

The magnetization is the sum of all spin for a given configuration:

$$\frac{\partial E_i}{\partial H} = -\sum_i s_i = -M_i \tag{14}$$

$$\frac{\partial}{\partial H} \frac{1}{Z} = \frac{\partial}{\partial H} \frac{1}{\sum_{i=1}^{M} e^{-\frac{1}{k_B T} E_i}} = -\frac{1}{k_B T} \frac{\sum_{i=1}^{M} M_i e^{-E_i \frac{1}{k_B T}}}{\left(\sum_{i=1}^{M} e^{-E_i \frac{1}{k_B T}}\right)^2} = -\frac{1}{k_B T} \frac{\sum_{i=1}^{M} M_i e^{-E_i \frac{1}{k_B T}}}{(Z)^2}$$
(15)

$$\frac{\partial}{\partial H} \sum_{i=1}^{M} M_i e^{-\frac{1}{k_B T} E_i} = \frac{1}{k_B T} \sum_{i=1}^{M} M_i^2 e^{-\frac{1}{k_B T} E_i}$$
(16)

$$\chi = \frac{\partial \langle M \rangle}{\partial H} = \frac{\partial}{\partial H} \left(\frac{1}{Z} \sum_{i=1}^{M} M_i e^{-\frac{1}{k_B T} E_i} \right) = -\frac{1}{k_B T} \frac{\sum_{i=1}^{M} M_i e^{-E_i \frac{1}{k_B T}}}{Z^2} \sum_{i=1}^{M} M_i e^{-\frac{1}{k_B T} E_i} + \frac{1}{Z} \frac{1}{k_B T} \sum_{i=1}^{M} M_i^2 e^{-\frac{1}{k_B T} E_i}$$
(17)

$$= -\frac{1}{k_B T} \left(\frac{\sum_{i=1}^{M} M_i e^{-E_i \frac{1}{k_B T}}}{Z} \right)^2 + \frac{1}{Z} \frac{1}{k_B T} \sum_{i=1}^{M} M_i^2 e^{-\frac{1}{k_B T} E_i} = \frac{1}{k_B T} \left(\langle M \rangle^2 - \langle M^2 \rangle \right)$$
(18)

1.1.3 Ferromagnetic order

A ferromagnet has a spontaneous magnetic moment even with the absence of an external magnetic field. Due the existence of a spontaneous moment the electron spin and magnetic moments must be arranged in a regular manner. In a ferromagnet most of the spins are aligned and in an anti ferromagnet the most of neighbouring spins are pointing in opposite directions. (compare "Introduction to Solid States Physics- 8th edition, by Charles kittel, page 323) In general -in a ferromagnet- spins up want to be next to spin up and spin downs want to be next to spin downs. At low temperature, the spin system is magnetised (either mostly up or mostly down). However, at high temperatures above the critical temperature, up and down spins are equally likely and the spins will in total cancel each other out. (total magnetisations is zero)

1.1.4 link from the Macroscopic values to statistical physics

In chapter 1.1.2, we derived the expression of the heat capacity and of the magnetic susceptibility of a canonical ensemble. We didn't care about statistical properties. However, what strikes the eye is that in the expression of the heat capacity as well as in the expression of the magnetic susceptibility, we see that they both depend on the variance of the energy of the magnetisation. this means that they can be written as:

$$C_{\nu} = \frac{1}{k_B T^2} \cdot \text{Var(E)} \tag{19}$$

$$\chi = \frac{1}{k_B T} \cdot \text{Var}(M) \tag{20}$$

The variance measures how far away a set of numbers is spread out. A huge variance means then that the values of the quantity fluctuate a lot around the expectation value. Now, we have linked a statistical property to a thermodynamic and macroscopic such as heat capacity.

1.2 theoretical numerical solutions

1.2.1 Ising model

Ising model is a mathematical model for ferromagnetism studies of phase transitions for magnetic system at given a temperature. The model consists the interaction between two neighbouring spins is related by the interaction energy

$$-Js_ks_l \tag{21}$$

where the sin s can be in two states +1 or -1,where s_k and s_l are the nearest neighbors. Which give a low energy (-J) if the two spin aligned and high energy (j) for spin pointing in opposite direction. The total energy to a system with N number of spins and with the absence of magnetic field can be expressed as

$$E = -J \sum_{\langle kl \rangle}^{N} s_k s_l \tag{22}$$

1.2.2 Periodic boundary conditions

Periodic boundary conditions is used for approximating a large or infinite system by using smaller repeating system, we will impose PBCs on our spin lattice in x and y directions.

$$s(L+1,y) = s(1,y)$$

 $s(x,L+1) = s(X,1)$

1.2.3 Metropolis algorithm in the two dimensional Ising model

The Ising model with Metropolis algorithm generates a sequence of states with Monte Carlo path, where the transition between states depends on the transition probability between the next and current state. The probability distribution is given by the Boltzmann distribution which is the probability for finding the system in a state s.

$$Ps = \frac{e^{-\beta E_i}}{\sum_{i=1}^{M} e^{-\beta E_i}}$$
 (23)

It is difficult to compute since we need the sum over all states. If we have a 10×10 spin lattice interacting in our Ising model, there are 2^{100} possible states. Computing the sum seems to be not that efficient, but luckily the Metropolis algorithm needs only the ratios between the state probabilities and we do not need to compute the sum of all the states after all. The Metropolis algorithm in this case can be implement by establishing two dimensional Ising model with random lattice configuration. Then we flip a randomly chosen spin and compute the energy difference ΔE . If

 $\Delta E \leq 0$ we accept the flip, otherwise we compute the transition probability $w = e^- \beta \Delta E$ and compare with a random number r.

If $r \le w$ we accept the flip otherwise we keep the old configuration. We can keep choosing new random spins until we are satisfied with a good representation of the states.(compare to "lecture notes Computational physics 2015 at University of Oslo by Morten Hjorth-Jensen page 435)"

1.2.4 Critical temperature (Lars Onsager)

in 1944 the Norwegian chemist Lars Onsager made very important discovery in theoretical physics, namely the exact solution of the Ising spin model in two dimensions. His work is up to now a valid theoretical description of the two dimensional Ising model. Onsager's solutions achieved the thermodynamic properties of interaction systems and phase transitions at T_c . However in 1942 Lars Onsager solved the two dimensional model for zero field energy, which has been published two years after. In 1948, he wrote the solution for the zero field magnetization in a conference at Cornell. Onsager showed how to derive the partition function for the canonical ensemble with zero external magnetic field Z(B=0,T) with N spins.

$$Z_N = \left(2\cosh(\beta J)e^I\right)^N \tag{24}$$

where $I = \frac{1}{2\pi} \int_0^{\pi} d\phi ln \left[\frac{1}{2} \left(1 + \sqrt{1 - \kappa^2 sin^2 \phi} \right) \right]$ where $\kappa = \frac{2 sinh(2\beta J)}{cosh^2(2\beta J)}$ and the energy is given by

$$\langle E \rangle = -J coth(2\beta J) \left[1 + \frac{2}{\pi} (2 tanh^2 (2\beta J) - 1) K_1(q) \right]$$
 (25)

where $q = sinh(2\beta J)/cosh^2(2\beta J)$ and the complete elliptic integral of the first kind is:

$$k_1(q) = \int_0^{\frac{\pi}{2}} \frac{d\phi}{\sqrt{1 - q^2 \sin^2 \phi}}$$
 (26)

and differentiating the energy with the respect to temperature we obtain the specific heat:

$$C_{v} = \frac{\partial \langle E \rangle}{\partial T} = \frac{4K_{B}}{\pi} (\beta J coth(2\beta J))^{2} \left\{ K_{1}(q) - K_{2}(q) - (1 - tanh^{2}(2\beta J)) \left[\frac{\pi}{2} + (2tanh^{2}(2\beta J) - 1)K_{1}(q) \right] \right\}$$
(27)

where

$$k_2(q) = \int_0^{\frac{\pi}{2}} d\phi \sqrt{1 - q^2 \sin^2 \phi}$$
 (28)

Near the critical temperature T_c the specific heat behaves as:

$$C_{\nu} \approx -\frac{2}{\pi} \left(\left(\frac{2J}{K_B T_c} \right)^2 ln \left[1 - \frac{T}{T_c} \right] + const.$$
 (29)

$$C_{\nu} \sim \left[1 - \frac{T}{T_{c}}\right]^{\alpha} \tag{30}$$

the limiting form of the function

$$\lim_{\alpha \to 0} \frac{1}{\alpha} (Y^{-\alpha} - 1) = -\ln Y \tag{31}$$

can be used to infer that closed-form result in low singularity with $\alpha = 0$. Onsager's result can be written as:

$$<\frac{M(T)}{N}> = \left[1 - \frac{(1 - \tanh^2(\beta J))^4}{16 \tanh^4(\beta J)}\right]^{\frac{1}{8}}$$
 (32)

for $T < T_c$. otherwise the magnetization is zero "lecture notes Computational physics 2015 at University of Oslo by Morten Hjorth-Jensen page 435)"

Closed solution for a 2 dimensional 2 x 2 lattice

We want now to look at a 2 x 2 lattice and we want to calculate the partition function, the energy, magnetisation, heat capacity and susceptibility of the system dependent of T. The partition function for a canonical ensemble with periodic boundary conditions can be computed by:

$$Z = \sum_{i=1}^{M} e^{-\beta E_i} \tag{33}$$

Here, β is $\frac{1}{k_b \cdot T}$, where k_b is the Bolzmann constant. In this expression we sum over all microstates m. The Energy of the system in configuration i is then:

$$E_i = -J \sum_{\langle kl \rangle}^N s_k s_l \tag{34}$$

The sum over < kl > means that we only sum over nearest neighbours. In our 2 x 2 case, we have for each "particle" two possible values ± 1 . This means that we have all in all $2^{2\cdot 2} = 2^4 = 16$ micro states. We have to compute the Energy of the micro states in order to compute the partition function. We also want to introduce the magnetisation, which is simply the sum over all the spins of the system:

$$M_i = \sum_{j=1}^{N} s_j \tag{35}$$

We want also to introduce the so called degeneracy, which counts the number of micro states for a given micro energy. We get the following table: We can now write the expression of the partition function as in equation 38. We used the

Figure 1: Energy of the different micro states

Number of spins up (+1)	Degeneracy	Energy	Magnetization
4	1	-8 <i>J</i>	4
3	4	0	2
2	4	0	0
2	2	8 <i>J</i>	0
1	4	0	-2
0	1	-8J	-4

Table 1 to calculate the sum over the micro states.

$$Z = \sum_{i=1}^{M} e^{-\beta E_i} = 12 \cdot e^{-\beta \cdot 0} + 2 \cdot e^{-8J\beta} + 1 \cdot e^{8J\beta} + 1 \cdot e^{8J\beta}$$
 (36)

$$=12+2\cdot e^{-8J\beta}+2\cdot e^{8J\beta}\tag{37}$$

$$= 12 + 4 \cdot \cosh\left(8J\beta\right) \tag{38}$$

We can now calculate the expectation value of the energy. There are two possible ways of calculating it. the first way of calculating the expectation value of the energy can be seen in equation 40.

$$\langle E \rangle = -\frac{\partial ln(Z)}{\partial \beta} = -\frac{1}{Z} \cdot 32J \cdot \sinh(8J\beta)$$
 (39)

$$= -\frac{32J \cdot \sinh((8J\beta))}{Z} \tag{40}$$

$$= -\frac{32J \cdot \sinh((8J\beta))}{Z}$$

$$= -\frac{8 \cdot J \cdot \sinh(8J\beta)}{3 + \cosh(8J\beta)}$$
(40)

Alternatively, we can calculate the expectation value of the Energy by looking at the micro states:

$$\langle E \rangle = \frac{1}{Z} \sum_{i=1}^{M} E_i e^{-\beta E_i} = -\frac{8 \cdot J \cdot \sinh(8J\beta)}{3 + \cosh(8J\beta)}$$
 (42)

Both expressions are equal. Next, we want to determine the expectation value of the magnetisation. We use the formula 44. We can see that we get 0 for the expectation value of the magnetisation.

$$\langle M \rangle = \frac{1}{Z} \sum_{i}^{M} M_i \cdot e^{-\beta E_i} \tag{43}$$

$$= \frac{1}{Z} \cdot \left(4 \cdot 1 \cdot e^{-8J\beta} + 2 \cdot 4 + (-2) \cdot 4 + (-4) \cdot 1 \cdot e^{-8J\beta} \right) \tag{44}$$

$$=0 (45)$$

However, we are interested in the expectation value of the absolute value of magnetisation, which is < |M| >. This expression can be determined as follows:

$$<|M|> = \frac{1}{Z} \sum_{i}^{M} |M_{i}| \cdot e^{-\beta E_{i}}$$
 (46)

$$= \frac{1}{Z} \cdot \left(|4| \cdot 1 \cdot e^{8J\beta} + |2| \cdot 4 + |(-2)| \cdot 4 + |(-4)| \cdot 1 \cdot e^{8J\beta} \right) \tag{47}$$

$$=\frac{1}{Z}\cdot\left(8\cdot e^{8J\beta}+16\right)\tag{48}$$

$$=\frac{2 \cdot e^{8J\beta} + 4}{3 + \cosh(8J\beta)} \tag{49}$$

In order to describe how the temperature will change when thermal energy is added to the system, we want to look at a quantity called heap capacity. (C_v) The bigger this quantity is the less heats the system up by a given amount of thermal energy, which is added to the system.

$$C_{\nu} = \frac{1}{k_b T^2} \left(\frac{1}{Z} \sum_{i=1}^{M} E_i^2 e^{-\beta E_i} - \left(\frac{1}{Z} \sum_{i=1}^{M} E_i e^{-\beta E_i} \right)^2 \right)$$
 (50)

$$= \frac{1}{k_b T^2} \left(\frac{1}{Z} \left(2 \cdot (8J)^2 \cdot e^{8J\beta} + 2 \cdot (-8J)^2 \cdot e^{-8J\beta} \right) - \left(\frac{8 \cdot J \cdot \sinh(8J\beta)}{3 + \cosh(8J\beta)} \right)^2 \right)$$
 (51)

$$= \frac{1}{k_B T^2} \left(\frac{64 \cdot J \cdot \cosh(8J\beta)}{3 + \cosh(8J\beta)} - \left(\frac{8 \cdot J \cdot \sinh(8J\beta)}{3 + \cosh(8J\beta)} \right)^2 \right)$$
 (52)

$$= \frac{1}{k_b T^2} \left(\frac{64 \cdot J + 3 \cdot J \cdot 64 \cosh(8J\beta)}{\left(3 + \cosh(8J\beta)\right)^2} \right)$$
 (53)

$$= \frac{64}{k_b T^2} \left(\frac{J + 3J \cdot \cosh(8J\beta)}{\left(3 + \cosh(8J\beta)\right)^2} \right) \tag{54}$$

(55)

At last, we want to have a look at the magnetic susceptibility. This quantity is a magnetic property of the material. The magnetic susceptibility describes the response of the material to an applied magnetic field.

$$\chi = \frac{1}{k_b T} \cdot \left(\frac{1}{Z} \sum_{i=1}^{M} M_i^2 e^{-\beta E_i} - \left(\frac{1}{Z} \sum_{i=1}^{M} M_i e^{-\beta E_i} \right)^2 \right)$$
 (56)

$$= \frac{1}{k_b T} \cdot \left(\frac{1}{Z} \cdot \left(4^2 \cdot 1 \cdot e^{-8J\beta} + 2^2 \cdot 4 + (-2)^2 \cdot 4 + (-4)^2 \cdot 1 \cdot e^{-8J\beta} \right) - (0)^2 \right)$$
 (57)

$$= \frac{1}{k_b T} \cdot \frac{32e^{-8J\beta} + 32}{12 + 4 \cdot \cosh(8J\beta)} \tag{58}$$

$$=\frac{1}{k_b T} \cdot \frac{8e^{-8J\beta} + 8}{3 + \cosh(8J\beta)} \tag{59}$$

(60)

$$\chi_{abs} = \frac{1}{k_b T} \cdot \left(\frac{1}{Z} \sum_{i=1}^{M} M_i^2 e^{-\beta E_i} - \left(\frac{1}{Z} \sum_{i=1}^{M} |M_i| e^{-\beta E_i} \right)^2 \right)$$
 (61)

$$= \frac{1}{k_b T} \cdot \left(\frac{1}{Z} \cdot \left(4^2 \cdot 1 \cdot e^{-8J\beta} + 2^2 \cdot 4 + (-2)^2 \cdot 4 + (-4)^2 \cdot 1 \cdot e^{-8J\beta} \right) - \left(\frac{2 \cdot e^{8J\beta} + 4}{3 + \cosh(8J\beta)} \right)^2 \right)$$
 (62)

$$= \frac{1}{k_b T} \cdot \left(\frac{32e^{-8J\beta} + 32}{12 + 4 \cdot \cosh(8J\beta)} - \left(\frac{2 \cdot e^{8J\beta} + 4}{3 + \cosh(8J\beta)} \right)^2 \right)$$
 (63)

$$= \frac{1}{k_b T} \cdot \left(\frac{8e^{-8J\beta} + 8}{3 + \cdot \cosh\left(8J\beta\right)} - \left(\frac{2 \cdot e^{8J\beta} + 4}{3 + \cosh(8J\beta)} \right)^2 \right) \tag{64}$$

$$= \frac{1}{k_b T} \cdot \frac{4(3 + 4\cosh(8J\beta) - 10\sinh(8J\beta) - 2\sinh(16J\beta))}{(3 + \cosh(8J\beta))^2}$$
(65)

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