

# Project 4, deadline April 30

## Computational Physics PHY480/905

Department of Physics and Astronomy, Michigan State University

Spring semester 2017

### Studies of phase transitions in magnetic systems

**Introduction.** The aim of this project is to study a widely popular model to simulate phase transitions, the so-called Ising model in two dimensions. At a given critical temperature, this model exhibits a phase transition from a magnetic phase (a system with a finite magnetic moment) to a phase with zero magnetization.

In its simplest form the energy of the Ising model is expressed as, without an externally applied magnetic field,

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

with  $s_k = \pm 1$ . The quantity  $N$  represents the total number of spins and  $J$  is a coupling constant expressing the strength of the interaction between neighboring spins. The symbol  $\langle kl \rangle$  indicates that we sum over nearest neighbors only. We will assume that we have a ferromagnetic ordering, viz  $J > 0$ . We will use periodic boundary conditions and the Metropolis algorithm only. The material on the Ising model can be found in chapter 13 of the lecture notes. The Metropolis algorithm is discussed in chapter 12.

**For this project you can collaborate with fellow students but you have to hand in an individual report.** This project (together with projects 3 and 5) counts 1/3 of the final mark.

**Project 4a): A simple  $2 \times 2$  lattice, analytical expressions.** Assume we have only two spins in each dimension, that is  $L = 2$ . Find the analytical expression for the partition function and the corresponding expectations values for the energy  $E$ , the mean absolute value of the magnetic moment  $|M|$  (we will refer to this as the mean magnetization), the specific heat  $C_V$  and the susceptibility  $\chi$  as functions of  $T$  using periodic boundary conditions.

**Project 4b): Writing a code for the Ising model.** Write now a code for the Ising model which computes the mean energy  $E$ , mean magnetization  $|M|$ , the specific heat  $C_V$  and the susceptibility  $\chi$  as functions of  $T$  using periodic boundary conditions for  $L = 2$  in the  $x$  and  $y$  directions. Compare your results with the expressions from a) for a temperature  $T = 1.0$  (in units of  $kT/J$ ).

How many Monte Carlo cycles do you need in order to achieve a good agreement?

**Project 4c): When is the most likely state reached?** We choose now a square lattice with  $L = 20$  spins in the  $x$  and  $y$  directions.

In the previous exercise we did not study carefully how many Monte Carlo cycles were needed in order to reach the most likely state. Here we want to perform a study of the time (here it corresponds to the number of Monte Carlo sweeps of the lattice) one needs before one reaches an equilibrium situation and can start computing various expectations values. Our first attempt is a rough and plain graphical one, where we plot various expectations values as functions of the number of Monte Carlo cycles.

Choose first a temperature of  $T = 1.0$  (in units of  $kT/J$ ) and study the mean energy and magnetisation (absolute value) as functions of the number of Monte Carlo cycles. Let the number of Monte Carlo cycles (sweeps per lattice) represent time. Use both an ordered (all spins pointing in one direction) and a random spin orientation as starting configuration. How many Monte Carlo cycles do you need before you reach an equilibrium situation? Repeat this analysis for  $T = 2.4$ . Can you, based on these values estimate an equilibration time? Make also a plot of the total number of accepted configurations as function of the total number of Monte Carlo cycles. How does the number of accepted configurations behave as function of temperature  $T$ ?

**Project 4d): Analyzing the probability distribution.** Compute thereafter the probability  $P(E)$  for the previous system with  $L = 20$  and the same temperatures, that is at  $T = 1.0$  and  $T = 2.4$ . You compute this probability by simply counting the number of times a given energy appears in your computation. Start the computation after the steady state situation has been reached. Compare your results with the computed variance in energy  $\sigma_E^2$  and discuss the behavior you observe.

**Studies of phase transitions.** Near  $T_C$  we can characterize the behavior of many physical quantities by a power law behavior. As an example, for the Ising class of models, the mean magnetization is given by

$$\langle M(T) \rangle \sim (T - T_C)^\beta,$$

where  $\beta = 1/8$  is a so-called critical exponent. A similar relation applies to the heat capacity

$$C_V(T) \sim |T_C - T|^\alpha,$$

and the susceptibility

$$\chi(T) \sim |T_C - T|^\gamma, \quad (1)$$

with  $\alpha = 0$  and  $\gamma = 7/4$ . Another important quantity is the correlation length, which is expected to be of the order of the lattice spacing for  $T \gg T_C$ . Because the spins become more and more correlated as  $T$  approaches  $T_C$ , the correlation length increases as we get closer to the critical temperature. The divergent behavior of  $\xi$  near  $T_C$  is

$$\xi(T) \sim |T_C - T|^{-\nu}. \quad (2)$$

A second-order phase transition is characterized by a correlation length which spans the whole system. Since we are always limited to a finite lattice,  $\xi$  will be proportional with the size of the lattice. Through so-called finite size scaling relations it is possible to relate the behavior at finite lattices with the results for an infinitely large lattice. The critical temperature scales then as

$$T_C(L) - T_C(L = \infty) = aL^{-1/\nu}, \quad (3)$$

with  $a$  a constant and  $\nu$  defined in Eq. (2). We set  $T = T_C$  and obtain a mean magnetisation

$$\langle \mathcal{M}(T) \rangle \sim (T - T_C)^\beta \rightarrow L^{-\beta/\nu}, \quad (4)$$

a heat capacity

$$C_V(T) \sim |T_C - T|^{-\gamma} \rightarrow L^{\alpha/\nu}, \quad (5)$$

and susceptibility

$$\chi(T) \sim |T_C - T|^{-\alpha} \rightarrow L^{\gamma/\nu}. \quad (6)$$

**Project 4e): Numerical studies of phase transitions.** We wish to study the behavior of the Ising model in two dimensions close to the critical temperature as a function of the lattice size  $L \times L$ . Calculate the expectation values for  $\langle E \rangle$  and  $\langle |M| \rangle$ , the specific heat  $C_V$  and the susceptibility  $\chi$  as functions of  $T$  for  $L = 40$ ,  $L = 60$ ,  $L = 100$  and  $L = 140$  for  $T \in [2.0, 2.3]$  with a step in temperature  $\Delta T = 0.05$  or smaller. You may find it convenient narrow the domain for  $T$ .

Plot  $\langle E \rangle$ ,  $\langle |M| \rangle$ ,  $C_V$  and  $\chi$  as functions of  $T$ . Can you see an indication of a phase transition? Use the absolute value  $\langle |M| \rangle$  when you evaluate  $\chi$ . For these production runs you should parallelize the code using MPI (recommended). Alternatively OpenMP can be used. Use optimization flags when compiling. Perform a timing analysis of some selected runs in order to see that you get an optimal speedup when parallelizing your code.

**Project 4f): Extracting the critical temperature.** Use Eq. (3) and the exact result  $\nu = 1$  in order to estimate  $T_C$  in the thermodynamic limit  $L \rightarrow \infty$  using your simulations with  $L = 40$ ,  $L = 60$ ,  $L = 100$  and  $L = 140$ . The exact result for the critical temperature (after Lars Onsager) is  $kT_C/J = 2/\ln(1 + \sqrt{2}) \approx 2.269$  with  $\nu = 1$ .

## Background literature

If you wish to read more about the Ising model and statistical physics here are three suggestions.

- M. Plischke and B. Bergersen, *Equilibrium Statistical Physics*, World Scientific, see chapters 5 and 6.
- D. P. Landau and K. Binder, *A Guide to Monte Carlo Simulations in Statistical Physics*, Cambridge, see chapters 2,3 and 4.
- M. E. J. Newman and T. Barkema, *Monte Carlo Methods in Statistical Physics*, Oxford, see chapters 3 and 4.

## Introduction to numerical projects

Here follows a brief recipe and recommendation on how to write a report for each project.

- Give a short description of the nature of the problem and the eventual numerical methods you have used.
- Describe the algorithm you have used and/or developed. Here you may find it convenient to use pseudocoding. In many cases you can describe the algorithm in the program itself.
- Include the source code of your program. Comment your program properly.
- If possible, try to find analytic solutions, or known limits in order to test your program when developing the code.
- Include your results either in figure form or in a table. Remember to label your results. All tables and figures should have relevant captions and labels on the axes.
- Try to evaluate the reliability and numerical stability/precision of your results. If possible, include a qualitative and/or quantitative discussion of the numerical stability, eventual loss of precision etc.
- Try to give an interpretation of your results in your answers to the problems.

- Critique: if possible include your comments and reflections about the exercise, whether you felt you learnt something, ideas for improvements and other thoughts you've made when solving the exercise. We wish to keep this course at the interactive level and your comments can help us improve it.
- Try to establish a practice where you log your work at the computerlab. You may find such a logbook very handy at later stages in your work, especially when you don't properly remember what a previous test version of your program did. Here you could also record the time spent on solving the exercise, various algorithms you may have tested or other topics which you feel worthy of mentioning.

### Format for electronic delivery of report and programs

The preferred format for the report is a PDF file. You can also use DOC or postscript formats or as an ipython notebook file. As programming language we prefer that you choose between C/C++, Fortran2008 or Python. The following prescription should be followed when preparing the report:

- Use your github repository to upload your report. Indicate where the report is by creating for example a **Report** folder. Please send us as soon as possible your github username.
- Place your programs in a folder called for example **Programs** or **src**, in order to indicate where your programs are. You can use a README file to tell us how your github folders are organized.
- In your git repository, please include a folder which contains selected results. These can be in the form of output from your code for a selected set of runs and input parameters.
- In this and all later projects, you should include tests (for example unit tests) of your code(s).
- Comments from us on your projects, with score and detailed feedback will be emailed to you.

Finally, we encourage you to work two and two together. Optimal working groups consist of 2-3 students. You can then hand in a common report.