

PHYS 516: Methods of Computational Physics  
 ASSIGNMENT 3- MC Simulation of the Ising Model

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## 1 Theoretical Foundation of Metropolis Algorithm

Consider a set of  $N$  states,  $\Gamma_1, \Gamma_2, \dots, \Gamma_N$  and let the probability to find the system in the  $m$ -th state,  $\Gamma_m$ , be  $\rho_m$ . Here, we prove that the probability distribution is a fixed point of the metropolis transition matrix defined below, i.e.,  $\Pi\rho = \rho$ .

$$(\text{Metropolis transition matrix})\pi_{mn} = \begin{cases} \alpha_{mn} & \rho_m \geq \rho_n, m \neq n \\ \frac{\rho_m}{\rho_n} \alpha_{mn} & \rho_m < \rho_n, m \neq n \\ 1 - \sum_{m' \neq n} \pi_{m'n} & m = n \end{cases}$$

Here,  $\pi_{mn}$  are elements of the matrix  $\Pi$ ,  $\rho_m$  are the elements of vector  $\rho$ , and  $\alpha_{mn}$  are elements of a symmetric attempt matrix, i.e.,  $\alpha_{mn} = \alpha_{nm}$ .

This can be proved by enforcing the Detailed Balance condition. Consider a pair of states  $m$  and  $n$ . Assuming  $\rho_m < \rho_n$ ,

$$\pi_{nm}\rho_m = \alpha_{nm}\rho_m = \alpha_{mn}\rho_m \quad (1)$$

where the second equality holds because of the symmetric attempt. For the same case,

$$\pi_{mn}\rho_n = \frac{\rho_m}{\rho_n} \alpha_{mn}\rho_n = \alpha_{mn}\rho_m \quad (2)$$

Using equations 1 and 2,

$$\pi_{nm}\rho_m = \pi_{mn}\rho_n \quad (3)$$

The left-hand side is a flux of probability that the current state is  $n$  and that the next state is  $m$ , and the right-hand side is a flux from  $m$  to  $n$ .

In order to show that above detailed balance condition is sufficient for the unit-eigenvalue relation, sum the both side over  $n$ .

$$\sum_{n=1}^N \pi_{mn}\rho_n = \underbrace{\sum_{n=1}^N \pi_{nm} \rho_m}_{=1} \quad (4)$$

$$\implies \sum_{n=1}^N \pi_{mn}\rho_n = \rho_m \quad (5)$$

$$\therefore \Pi\rho = \rho \quad (6)$$

## 2 2D Ising Model

Ising model is used for modelling ferromagnetic materials. This model represents a lattice consisting of atoms which have quantum mechanical spin, which can be  $\pm 1$ . When these individual magnetic fields are aligned in the same direction, it gives rise to macroscopic magnetic field. This strong alignment arises from exchange interactions between electrons.

### 2.1 Computer Simulation

A program was written in C to simulate the 2D Ising model.

```
1  /* Monte Carlo Simulation of 2D Ising Model */
2  #include <stdio.h>
3  #include <stdlib.h>
4  #include <time.h>
5  #include <math.h>
6
7  // Define global variables
8  #define L 20 //lattice size
9  int s[L][L]; //Spins s[i][j]=+-1
10 double exp_dV[2][5];
11 double JdivT; // J/kBT
12 double HdivT; // H/kBT
13
14 void table_set(){ //function to set up the table
15     int sDash,sneighbor,k,l;
16     for (k=0;k<2;k++){
17         sDash = 2*k-1;
18         for (l=0;l<5;l++){
19             sneighbor = 2*l-4;
20             exp_dV[k][l] = exp(2*sDash*(JdivT*sneighbor + HdivT)); // check
                formula
21         }
22     }
23 }
24
25 int main() {
26     double runM;
27     double sumM=0.0, sumM2=0.0;
28     double exp_val, avgM, sigM;
29     int snew,sneighbor,Sta_step;
30     int i,j,step,k,l,im,ip,jm,jp;
31     FILE *f = fopen("Magnetization_data.txt", "w");
32
33     printf("Input J/kBT\n");
34     scanf("%le",&JdivT);
35
36     HdivT = 0.0;
37     Sta_step = 2000000;
```

```

38
39 table_set(); // Set up the look-up table for the exponent
    calculation
40
41 for(i=0;i<L;i++) { //Cold start- start with all spins up
    configuration
42     for(j=0;j<L;j++) {
43         s[i][j] = 1;
44     }
45 }
46 runM=1.0*L*L;
47
48 for(step=0; step<Sta_step; step++) {
49     i=rand()%L; j=rand()%L;
50     snew=s[i][j];
51
52     // Figure out which element of the table is to be looked up
53     im = (i + L - 1) % L;
54     ip = (i + 1) % L;
55     jm = (j + L - 1) % L;
56     jp = (j + 1) % L;
57     k = (snew+1)/2;
58     sneighbor = s[im][j] + s[ip][j] + s[i][jm] + s[i][jp];
59     l = (sneighbor+4)/2;
60
61     //Change in Pot Energy wth flip
62     exp_val=exp_dV[k][l];
63     // Accept or reject flip conditionally
64     if (exp_val>1.0) {
65         s[i][j] = snew;
66         runM += 2*snew; //update value of magnetization
67     }
68     else if(rand()/(double)RAND_MAX < exp_val) {
69         s[i][j] = snew;
70         runM += 2*snew; //update value of magnetization
71     }
72     sumM += runM;
73     sumM2 += runM*runM;
74     fprintf(f, "%f\n", runM);
75 }
76 fclose(f);
77 avgM = sumM/Sta_step; //Mean Magnetization
78 sigM = sqrt(sumM2/Sta_step-avgM*avgM); //Standard deviation
79 printf("Mean Magnetization %le, Standard Deviation %le \n",
    fabs(avgM), sigM);
80 }

```

IsingModel.c

## 2.2 Results

The plots below show the Mean magnetization and its standard deviation as a function of the exchange coupling, and a histogram showing the number of MC samples for every value of Magnetization. The matlab script used to generate these plots is shown below.

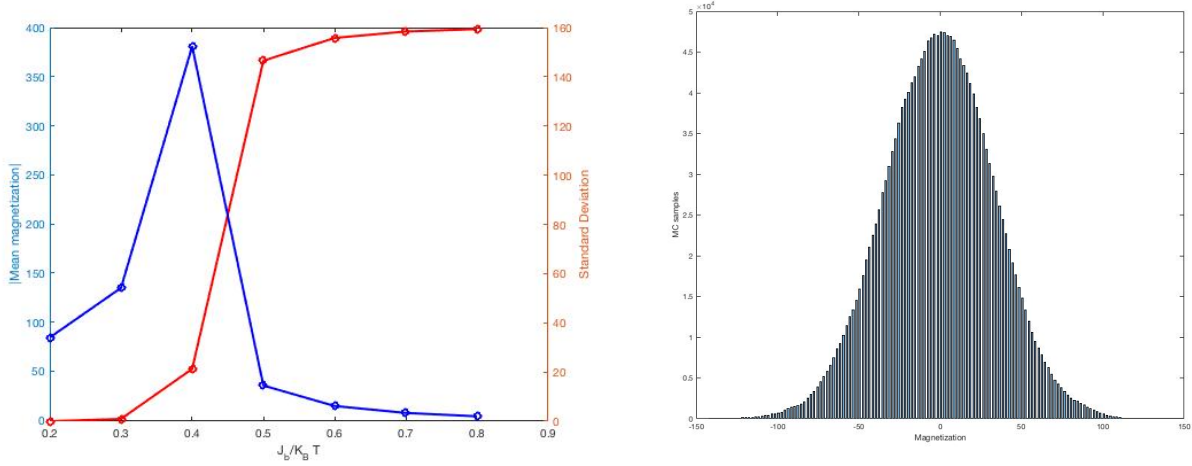


Figure 1: Plot of Magnetization and S.D vs  $J/K_B T$ , and Histogram for  $J/K_B T = 0.2$

```

1 close all; clear all;
2
3 %% Scatter Plots
4 JDivT = [0.2:0.1:0.8];
5 M = [-9.427350e-01,1.361261e+00 ,5.233627e+01,...
6      3.662830e+02,3.894057e+02,3.959868e+02,3.984030e+02];
7 yyaxis left;
8 plot(JDivT, M, 'r-o','LineWidth', 2);
9 xlabel('J_b/K_B T'); ylabel('|Mean magnetization|');
10
11 sd = [3.403888e+01, 5.418066e+01,1.523472e+02,...
12      1.449735e+01,6.180052e+00,3.360630e+00,1.993231e+00];
13 yyaxis right;
14 plot(JDivT, sd, 'b-o','LineWidth', 2);
15 xlabel('J_b/K_B T'); ylabel('Standard Deviation');
16
17 %% Histogram
18 figure()
19 A = importdata('Magnetization_data.txt', '\n', 0);
20 histogram(A);
21 ylabel('MC samples'); xlabel('Magnetization');
22 xlim([-150 150]);

```

Ising\_MagnetizationPlots.m