

UNIVERSITÀ DEGLI STUDI DI TORINO TANS Course A.A. 2017/2018 Prof. Massimo Masera Ising Model Simulation with MC methods

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Abstract

A computational solution for the Ising model is proposed using Metropolis-Hastings algorithm. Results show critical temperature for 2D and 3D models under periodic boundary conditions. Different sizes and temperatures are studied in agreement with algorithm performances and sensibilities to the parameters. We plotted energy, susceptibility and magnetization vs temperature and we fitted it in order to find critical exponent.

Introduction

The Ising model is a well studied model in Statistical Mechanics which describes ferromagnetic phenomena. It is characterized by a microscopic configuration space based on D-dimensional lattice, that brings to macroscopic statistical quantities. Moreover it can easily generalize the concept of collective effects caused by binary valued points interacting in pairs; for this reason the Ising model became the core of the physics of complex systems. In the last decades computational methods have been applied to search a numerical solution for the 3D Ising model in order to fill the lack of an analytical solution. Alongside Metropolis algorithm became the most popular method of important sampling in MC. The basic idea of a weighted sampling based on the importance of a region determined a great improvement for the numerical solutions in general. Under these premises we want to outline the purpose this work wants to pursuit: finding numerical solution of the D-dimensional Ising model with Metropolis algorithm.

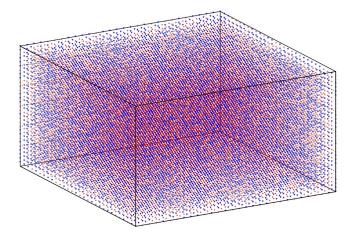


Figure 1: High temperature simulation of 3D Ising model

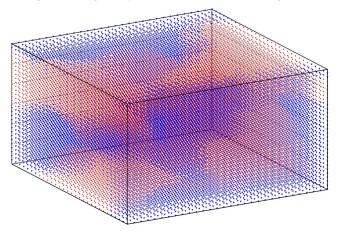


Figure 2: Low temperature simulation of 3D Ising model

1 Theory

1.1 Ising model and importance sampling

The Ising model is a physical-mathematical model characterized by a lattice of local spins in the [-1;+1] domain. The correspondent energy is proportional to the quadratic interaction of near spins (σ_i) :

$$H(q,\sigma) = \sum_{\langle l,m \rangle}' -J\sigma_i \sigma_j \tag{1}$$

Where J is the energy interaction term and we sum over the neighbours.

The system has a macroscopic equilibrium state (macrostate) described by statistical quantities: number of spins N, lattice volume V, energy E, magnetization M. According to the Statistical Mechanics theory we consider the thermodynamic limit, i.e. $N, V \to \infty$ and N/V = cost in order to have negligible fluctuations on energy and magnetisation: $(\Delta E)^2/\langle E \rangle = O(1/\sqrt{N})$. From a theoretical point of view we consider an ensemble of lattices, each one with a particular configuration of spins (microstate) that corresponds to the same macrostate; we point out that all microstates are independent and equiprobable. The configuration space is described through the Hamiltonian formalism indeed, under ergodic hypothesis, the system visits all microstates. Moreover stationary hypothesis implies the existence of at least an equilibrium state in which the probability distribution of the microstates satisfies the Liouville theorem. These hypotheses bring respectively to the following statements:

- 1. In the $t \to \infty$ limit: the average (over all the ensemble) of a physical quantity is equal to the temporal average of that quantity.
- 2. The probability distribution of the microstates depends only on the Hamiltonian:

$$\rho(q,\sigma) \propto \exp\{-\beta H(q,\sigma)\} \tag{2}$$

Where q is the spin position on the lattice and β is the product of the Boltzmann constant and the temperature T of the system.

In a finite case the Ising model system is characterized by magnetization and energy fluctuations that brings us to consider it under canonical formalism. We define the partition function as follows: $Z = \sum_q exp[-\beta H(q)]$. The model simulation can be done through Monte Carlo method, i.e. we generate a pseudo-random chain of numbers in order to extract a microstate sample (from all the configuration space) distributed by the following equilibrium distribution of probability:

$$P_{eq}(q) = \exp\{-\beta H(q)\}/Z \tag{3}$$

In practice we cannot compute the partition function, moreover an algorithm that searches over all the phase space has not a great performance.

The Boltzmann's factor as probability of choosing a configuration gives a great improvement for the simulation. In this way we are not sampling all the configuration space but a selected and uniform distributed microstates collection. This approach is called *importance sampling* and it allows us to calculate physical quantities as the average number over the extracted configuration:

$$\langle E \rangle_N = \frac{1}{N} \sum_{i=1}^N E(q_i)$$

This intuition can be proved under ergodic hypothesis: in fact we can consider an evolving Ising system in which there is a spin flip at each time step. The temporal collection of microstates is a Markov chain with a transition probability (W) that leads the system to the unique equilibrium probability in the $t \to \infty$ limit. Indeed the ratio of the transition probability is:

$$\frac{W(q \to q')}{W(q' \to q)} = \frac{P_{eq}(q')}{P_{eq}(q)} = \exp\{\beta [E(q) - E(q')]\}$$
(4)

Where q and q' are respectively the configuration before the spin-flip and after. This implies the equivalence between importance sampling and random extraction. The arbitrary choose on $W(q \to q')$ determines the particular algorithm; for the Metropolis-Hastings algorithm:

$$W(q \to q') = min \left\{ 1, \frac{P_r(q')P_{eq}(q')}{P_r(q)P_{eq}(q)} \right\}$$
 (5)

Where $P_r(q)$ is the probability to be in the q configuration; we consider symmetric probability: $P_r(q) = P_r(q')$. In this way the configuration q' is more probable than q if $P_{eq}(q') > P_{eq}(q)$ and the step transition could occur with the following probabilities and conditions:

$$\begin{cases} W(q \to q') = 1 & \Delta E \leq 0 \\ W(q \to q') = \exp\{-\beta \Delta E\} & \Delta E > 0 \end{cases}$$

Phase transition in Ising model

Magnetization in Ising model shows phase transition at a certain temperature called T_c . This means that the logarithm of the partition function has a critical point of the first order at this temperature and two different statistical descriptions concurrently exist at T_c . Indeed the magnetization curve has two different fits before this point and after, these fits correspond to inner configuration of spins: random for $\langle M \rangle = 0$ and ordered for $\langle M \rangle \to +1/-1$.

2 Model

2.1 Lattice Implementation

A D-dimensional Lattice with N spins along each edge is described by:

$$\mathbf{L} = (L_0, L_1, ..., L_{N^D-1})$$

Let i denote the index of array L: $i=\{0,1,...,N^D-1\}.$

 $L_i = \{0, 1\}$ is a boolean entry representing down and up spin by convention.

Let *i*-spin denote the spin represented by the L_i boolean entry.

The cartesian coordinates $\mathbf{a} = (a_0, a_1, ..., a_{D-1})$ of the *i*-spin can be computed from index *i* if a convention is set on the Lattice arrangement, which in our model is the following:

- Origin O of the D-dimensional space is a spin-vertex of the cubic Lattice, each edge starting from O is aligned along one of the D positive directions.
- Unitary distance between each couple of adjacent spins.
- Spins are indexed starting from increasing the first dimension's coordinate, than the second and so forth.

n.b. This rules imply that, given a specific spin, each coordinate a_j takes values in range $\{0, 1, \ldots, N-1\}$.

Index i can be expressed as a power series of N with coefficients a_i :

$$i = \sum_{j=1}^{D-1} a_j N^j \tag{6}$$

The a_i coordinates can then be computed from i as:

$$a_j = [i\% N^{j+1}]/N^j (7)$$

2.2 Energy Computation

The Lattice's energy is computed according to equation (1). The Lattice class provides a method which returns the energy of the system.

The implemented algorithm takes the lattice array **L** and performs two for loops: one over the system's dimension d = (0, 1, ..., D - 1) and the other over the index $i = (0, 1, ..., N^D - 1)$.

The key idea is that, given a fixed *i*-spin, along each dimension d, two neighbours $i_{d\pm}$ -spin are found on the increasing and decreasing d-coordinate respectively.

So each single spin has 2D interacting neighbours in total, by summing up their energy interaction terms the contribute of the single spin to the total energy is obtained. So the energy can be rewritten as:

$$H = \frac{1}{2} \sum_{i=0}^{N^{D}-1} \sum_{d=0}^{D-1} \sum_{\pm} -J\sigma_{i}\sigma_{i_{d_{\pm}}}$$
 (8)

Performing \sum_i , double countings of couples occur and a factor 1/2 is required. For each fixed index i, the algorithm takes into account only the interaction term with the i_{d_+} -spin so that the number of operation is halved. So the algorithm computes:

$$H = \sum_{i=0}^{N^{D}-1} \sum_{d=0}^{D-1} -J(L_i \oplus L_{i_{d_+}})$$
(9)

 $\sigma_i\sigma_{id_+}$ has been replaced by $L_i\oplus L_{id_+}$, the former can take values ± 1 wether the two spins are aligned or not. Since spins are represented by boolean entries of array L_i , the XOR bitwise operator \oplus is applied on couple (L_i,L_{id_+}) so that it returns the same value of $\sigma_i\sigma_{id_+}$.

From the *i* index coordinates $\mathbf{a} = (a_0, a_1, ..., a_{D-1})$ the i_{d_+} index coordinates

$$\mathbf{a^{d_+}} = (a_0^{d_+}, a_1^{d_+}, .., a_{D-1}^{d_+})$$
 can be computed.

The vector $\mathbf{a}^{\mathbf{d}_{+}}$ differs from \mathbf{a} only on the d entry $a_{d}^{d_{+}}$ which is the only one to be increased, this is computed as $(a_{d}+1)\%N$ since we are applying periodic boundary conditions on the Lattice.

From this formulas we can express the $i_{d_{+}}$ index as the sum of two terms:

$$i_{d+} = \left\{ \sum_{j=0}^{d-1} a_j N^j + [(a_d+1)\%N]N^d \right\} + \left\{ \sum_{j=d+1}^{D-1} a_j N^j \right\}$$
$$= \left\{ (i+N^d)\%N^{d+1} \right\} + \left\{ (i/N^{d+1})N^{d+1} \right\}$$

n.b.: if d = D - 1 the second term equals 0, so it is not computed by the algorithm

2.3 Metropolis Algorithm

At each step Metropolis Algorithm proposes a Lattice configuration q' which differs from the initial configuration q only by one spin value.

The number of possible q' configurations is equal to the total number of spins N^D ; q' is chosen with uniform probability.

Defining the ratio $a = P_{eq}(q')/P_{eq}(q)$

- 1. If $a \ge 1$ than q' is accepted
- 2. If a < 1:

```
q' is accepted with probability a if q' is rejected (with probability 1-a) configuration q is kept
```

That means q' is always accepted when its probability is greater than q, otherwise it is accepted with probability $P_{eq}(q')/P_{eq}$.

According to formula (3) $a = \exp\{-\beta \Delta E\}$; it is easy to observe that when $a \geq 1$ than $\Delta E \leq 0$. The latter inequality is taken as the condition of point 1.

The Algorithm's single step is implemented as following:

- 1. Random integer i is extracted with uniform probability from $\{0, 1, ..., N^D 1\}$
- 2. The energy variation ΔE resulting from flipping the i-spin is computed
- 3. If $\Delta E \leq 0$ than the *i*-spin is flipped
- 4. If $\Delta E > 0$:

A random double number u is extracted with uniform probability from [0,1] if $u<\exp\{-\beta\Delta E\}$ i-spin is flipped if $u>\exp\{-\beta\Delta E\}$ i-spin is not flipped

3 Results

This section wants to summarize the simulation results. Under this purpose we underline two main effects:

- Finite size and finite volume of the lattice imply magnetization and energy fluctuations; periodic conditions reduce this phenomena. Moreover finite size leads to no residual magnetization: this effect is out of the temperature's range studied.
- Importance sampling algorithms can produce correlated configurations corresponding to different steps of the simulation. This phenomena is in contrast with theoretical hypothesis of microstates independence and can be correctly treated through binning technique.

Calibrating the model

Simulations are done with 10^8 steps for thermalization (10^9 for the 3D lattice) and more than 10^7 steps with useful data. We varied the lattice size from 40x40 to 100x100 number of spins. Simulation time is really sensible to the number of temperature steps. We choose 100 steps of temperature from 0.5K to 3.5K centred around theoretical critical temperature (2.27K for 2D, 4K for 3D). (See Figure 12 in Appendix)

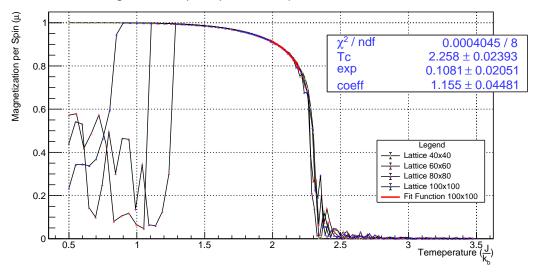
-2000 -4000 -8000 -10000 -14000 -16000 -18000 -18000 -18000 -18000 -20000 -1500

Energy on Temperature for Different Size

Figure 3: Energy-Temperature graph for 2D Lattices with I=10,000,000 and 100 steps of T

The binning method can take into account the correlation between different configurations. Measures corresponding to different steps of the simulation can have a correct error valuation through this technique: we divide a set of N ordered configurations in $N/2^n$ groups (each one with n elements), we calculate the mean for each bin and we iterate this process until the desired level of binning; then we calculate the global mean

Magnetization per Spin on Temperature for Different Size



 $\textbf{Figure 4:} \ \ \text{Magnetization-Temperature graph for 2D Lattices with I=10,000,000 and 100 steps of T}$

Magnetic Susceptibility on Temperature for Different Size

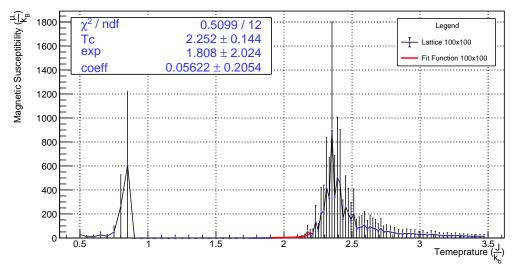


Figure 5: Susceptibility-Temperature graph for 2D Lattices with I=10,000,000 and 100 steps of T

and its error. In this way we correlate the measure at a particular level of binning (higher level \rightarrow correlation length << size of the bins \rightarrow lower correlation \rightarrow higher error). We chose the eighth level of binning for which we found a plateau in the error-level graph.

Energy on Temperature for Different Size

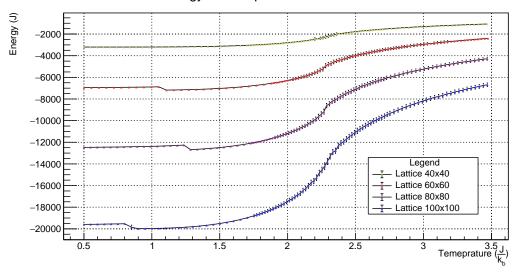


Figure 6: 10,000,000 steps of cooling for the 2D Lattice with L=100

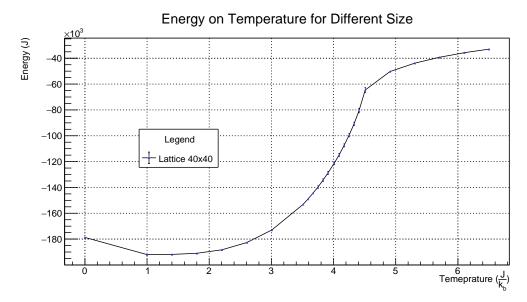


Figure 7: Energy-Temperature graph for 3D Lattices with I=40,000,000 and 25 steps of T

4 Conclusions

Metropolis-Hastings algorithm has been implemented in Root (C++) in order to show phase transitions in the 2D and 3D Ising model. Simulation parameters are the following:

Magnetization per Spin on Temperature for Different Size

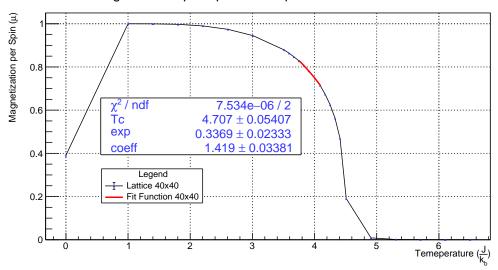


Figure 8: Magnetization-Temperature graph for 3D Lattices with I=40,000,000 and 25 steps of T

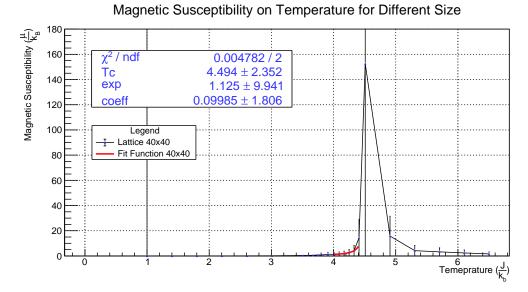


Figure 9: Susceptibility-Temperature graph for 3D Lattices with I=40,000,000 and 25 steps of T

number of spins for side: 40-60-80-100
thermalization steps: 10⁸-10⁹

• data collection steps: $> 10^7$

• temperature measures: 25 - 100

Energy on Temperature for Different Size

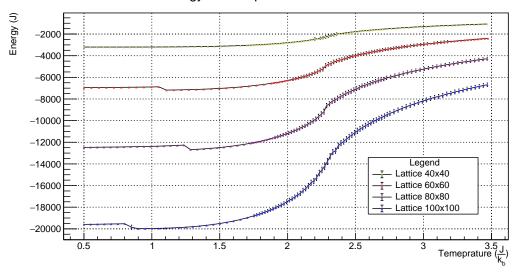


Figure 10: 1,000,000,000 steps of cooling for the 3D Lattice with L=40

We recall the theoretical critical temperature of the 2D Ising model with infinite size: 2.27K. The 3D Ising model does not have an analytical solution so we present a result in harmony with the results accepted by scientific community.

Me	odel	Critical T	Critical Exp
	2D	Numb. Users	Max Delivery Cost
3	BD	700	1.75 €

5 Appendix

Code

This section's aim is to introduce the reader to the C++ implementation of the model. The code is based on TObject class of Root and its scheme is the following:

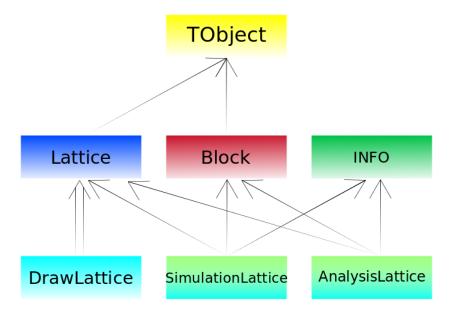


Figure 11: Class scheme

Classes

Lattice

The Lattice class permits to construct a Lattice, compute its termodynamic quantities and change its state with the Metropolis Algorithm.

Computation details of functions such as **energy**() and **cooling**() are thoroughly explained in Model section.

PRIVATE DATA MEMBERS

const uint	${f N}$	number of spins along one edge
const uint	dim	dimension
const uint	$\mathbf{num_spin}$	total number of spins : N^{dim}
bool *	lattice	boolean array of size num_spin
static double	${f T}$	temperature

PUBLIC MEMBER FUNCTIONS

CONSTRUCTORS

$\mathbf{Lattice}()$

Default Constructor

Creates Lattice object with N=1, dim=1

The single entry of lattice is set to 0 or 1 with 0.5 probability

Lattice(const uint& _N , const uint& _dim)

Standard Constructor

Creates Lattice object with N = N, $\dim = \dim$

Sets each lattice entry to 0 or 1 with 0.5 probability

Lattice(const Lattice& obj)

Copy Constructor

Creates Lattice object with obj's data members

DESTRUCTOR

\sim Lattice

Frees up memory allocated by lattice

PHYSICAL AND NUMERICAL FUNCTIONS

```
bool flipSpin(const uint& n)
    If n < num\_spin
    Sets lattice[n] to !lattice[n] and returns true
    Else it doesn't change lattice and returns false
int dE(const uint& n) const
    Returns energy variation resulting from applying flipSpin(n)
    n.b.: it's a const method, it doesn't apply flipSpin(n)
int energy() const;
    Returns total Energy
float magnetization() const
    Returns Magnetization per site
void cooling(const uint& iter)
    Applies one step of Metropolis Algorithm
void cooling(const uint& iter)
    Applies iter steps of Metropolis Algorithm
double * coolingPar()
    Applies cooling() and returns a four-dimensional array arr
    arr are respectively variations of :
    Temperature, Energy, Magnetization, Energy per site
OVERLOADED OPERATORS
Lattice& operator=(const Lattice& obj)
    Assignment operator
friend std::ostream& operator<<(std::ostream& out, const Lattice& lat)
    Taking lat as a reference to a Lattice object
    Prints lattice member of lat by typing the command cout << lat;
bool operator==(const Lattice& obj)
```

Taking lat1 and lat2 as references to Lattice objects

lat1 == lat2 returns true if lat1 and lat2 have same data members

GETTERS AND SETTERS

```
uint getN() const ; uint getDim() const; uint getNumSpin() const
    returns respectively N , dim , num_spin
bool getSpin(const uint & n) const
    returns lattice[n] entry
double getT() ; static void setT(const double& _T)
    respectively :
    Returns T ; Sets T = _T

OTHERS

void printLatticeCSV(const TString& name) const
    prints lattice in a csv file saved in the working directory
void printLatticeROOT(const TString& name , const TString& ln = "lat") const
    saves Lattice object in a root file
```

DrawLattice

DrawLattice class permits to draw a 2D or 3D Lattice inside a TCanvas

The Lattice object to be drawn can be imported in two ways with two Constructors The first constructor takes a Lattice object as argument

The second constructor takes the name of a root file and the name of a Lattice saved with the **printLatticeROOT** method

The Lattice is displayed in a TMultiGraph which has two TGraphs for up and down spins, with respectively kRed and kBlue color.

The computation of the spin coordinates is explained in the Model Section.

PRIVATE DATA MEMBERS

Lattice	lattice	Lattice object
uint	N , dim, num_spin	N, dim, num_spin of lattice
TString	fname	name of root file
TString	lname	name of Lattice saved in root file fname
TString	${\bf cname}\;,{\bf ctitle}$	Tcanvas name and title
TString	\mathbf{gname} , \mathbf{gtitle}	TMultiGraph name and title

cname, **ctitle**, **gname**, **gtitle** are assigned the same way in each of the three Constructors as : cname("cv"), ctitle("default canvas"), gname("gr"), gtitle("lsing")

PUBLIC MEMBER FUNCTIONS

CONSTRUCTORS

DrawLattice()

Default Constructor

Creates lattice with Lattice Default Constructor

DrawLattice(const Lattice& lat)

First Standard Constructor

Sets lattice = lat

DrawLattice(const TString& _fname, const TString& _lname)

Second Standard Constructor Searches a Tfile with name _fname

Sets lattice with readFile()

void readFile()

Gets Lattice object with name _lname from the TFile and sets it to lattice

DRAW FUNCTION

```
void draw()
         performs a switch statement on dim
         if dim = 2 applies draw2D()
         if dim = 3 applies draw3D()
         else returns without applying a draw Function
    GETTERS
    uint getN(); uint getDim(); uint getNumSpin()
         returns respectively N, dim, num_spin
PRIVATE MEMBER FUNCTIONS
    DRAW FUNCTIONS
    void draw2D()
         draws a bidimensional TMultiGraph
    void draw3D()
         draw a threedimensional TMultiGraph
    SETTERS
    void \ \mathbf{setN}() \ ; \ void \ \mathbf{setDim}() \ ; \ void \ \mathbf{setNumSpin}()
         Sets N, dim, num_spin respectively to lattice.N, lattice.dim, lattice.num_spin
         n.b. applied in First and Second Standard Constructor after lattice is set
    void setN(_N); void setDim(_dim); void setNumSpin(_num_spin )
         Sets N, dim, num_spin respectively to _N, _dim, _num_spin
```

```
st This macro explains how to construct handle and draw a Lattice st
        -----*/
7 #include "Lattice.h"
8 #include "DrawLattice.h"
10 {
11
    //LATTICE CLASS
12
13
   Lattice::setT(0.5); //Set static member Temperature to 0.5
14
                         //Default is T = 0.
15
16
    int N = 10;
17
    int dim = 2;
18
19
    Lattice lat(N,dim); //Construct a bi-dimensional Lattice
20
                         //with 10 spins along each edge
    cout << "lat visualization :" << endl;</pre>
22
    cout << lat << endl; //Print Lattice spins on the console</pre>
23
                          //The lattice spins will be disposed randomly
24
25
    cout << "energy E of lat : " << lat.energy() << endl;</pre>
26
    cout << "magnetization M of lat : " << lat.magnetization() << endl;</pre>
27
28
    lat.cooling(); //Perform a Metropolis step
29
30
    double * data = new double[4];
    data = lat.coolingPar(); //perform a Metropolis step collecting data
33
    cout << "T = " << data[0] << " : temperature" << endl;</pre>
34
    cout << "dE = " << data[1] << " : energy variation" << endl;
35
    \verb|cout| << "dM = " << data[2] << " : magnetization variation" << endl; \\
36
    cout << "dS = " << data[3] << " : energy per site variation" << endl;
37
    delete[] data;
38
39
    /*
40
41
     * According to Metropolis algorithm if dE <0
42
     * there's a probability [1 - exp(-dE/T)] (K=1)
43
     * that the step will not change lat state with a spinFlip
44
     * in that case dE=0; dM=0; dS=0;
45
46
     */
47
48
49
50
51
```

```
54
     const unsigned iter = 10000;
55
     lat.cooling(iter); // Perform 10000 Metropolis steps
56
57
     cout << lat << endl; //Print Lattice spins on the console</pre>
58
                            //The lattice will now be ordered
59
60
     cout << "energy E of lat : " << lat.energy() << endl;</pre>
61
     // Energy will be around -200
62
     cout << "magnetization M of lat : " << lat.magnetization() << endl;</pre>
63
     // Magnetization per site will be about 1 or -1
64
66
67
      st If T is set to a value greater than the critic temperature
68
      * (for 2D Ising Model Tc = 2.27)
69
      * and the same Macro is executed,
70
      * lat will not thermalize.
71
72
73
      */
74
     //DRAWLATTICE
75
77
     lat = Lattice(10,2);
78
79
     DrawLattice drawLat1(lat); // Create a DrawLattice object
80
                                   // lat is passed in order to draw it
81
82
83
     cout << "lat with dim = " << lat.getDim() << endl;</pre>
84
     cout << "and edge = " << lat.getN() << "will be drawn" << endl;</pre>
85
86
     drawLat1.draw(); // Draws the three-dimensional Lattice lat
89
     Lattice lat3D(20,3); // Construct a three-dimensional Lattice
90
                            // with 20 spins along each edge
91
92
     lat = lat3D;
                   // Use assignment operator to reassign lat
93
94
95
     cout << "lat with dim = " << lat.getDim() << endl;</pre>
96
     cout << "and edge = " << lat.getN() << "will be drawn" << endl;</pre>
99
100
     DrawLattice drawLat2(lat);
     drawLat2.draw();
101
102
103 }
```

Listing 1: Caption

SimulationLattice

Simulation Lattice permits to run a simulation and collect data during runtime. The simulation is performed on a set of Lattice objects over a range of temperatures All Lattices have same edge length and same dimension For each temperature :

each Lattice is submitted to **I0** cooling iterations in order to reach thermal equilibrium.

In this step data are not collected

each Lattice is submitted to **iter** cooling iterations, during each iteration measures of temperature, energy, magnetization and susceptibility are collected.

PRIVATE DATA MEMBERS

Lattice *	${\bf lattice_vector}$	vector storing a Lattice for each entry
const uint	$\mathbf N$	edge length of each Lattice
const uint	$\operatorname{\mathbf{dim}}$	dimension of each Lattice
const uint	$\mathbf{dim_vector}$	dimension of lattice_vector
const TString	file	root file name in which data are collected
uint	I 0	number of iterations not collecting data
uint	iter	number of iterations collecting data
double	$\mathbf{tempmin}$	minimum range's temperature
double	tempmax	maximum range's temperature
uint	$\mathbf{tempstep}$	number of temperatures in the range

n.b.: **tempstep** should be a multiple of 4 for implementation reasons

PUBLIC MEMBER FUNCTIONS

CONSTRUCTORS

SimulationLattice()

Default Constructor

SimulationLattice(const uint& _N, const uint& _dim, const uint& _dim_vector)

3-Parameters Constructor

The other data members can be assigned with the setter methods

SimulationLattice(const uint& _N, const uint& _dim, const uint& _dim_vector, const TString& _file, const uint& _i0, const uint& _iter, const double& _tempmin, const double& _tempmax, const uint& _tempstep)

Full Parameter Constructor

SimulationLattice(const Lattice& lat, const uint& _dim_vector, const TString& _file, const uint& _i0, const uint& _iter, const double& _tempmin, const double& _tempmax, const uint& _tempstep)

Constructor based on a Lattice Sets N and dim as those of lat

SimulationLattice(const SimulationLattice& obj)

Copy Constructor

SimulationLattice& operator=(const SimulationLattice& obj)

Assignment Operator

DESTRUCTOR

\sim SimulationLattice

Frees up memory allocated by lattice_vector

RUN FUNCTION

void run() const

Performs the simulation

GETTERS AND SETTERS

Getters methods are defined for all the Private Data Members except for lattice_vector, but its entries can be obtained with:

Lattice **getLattice**(const uint& i)

returns lattice_vector[i]

Setters methods are defined for all the Private Data Members except for $lattice_vector$ N, dim and dim_vector

AnalysisLattice

AnalysisLattice performs an analysis on raw data collected on a simulation input-file The analysis creates an output-file which stores :

mean and standard deviation of physical quantities for each Lattice and temperature of the simulation.

mean and standard deviation performed over the Lattices for each temperature of the simulation

Once the output file is created, the physical functions can be plotted into a graph The curves of Magnetization and Susceptibility vs Temperature can be fitted in order to estimate the Critical Temperature and Exponenents of the phase transition.

PRIVATE DATA MEMBERS

```
const TString file_in input root file name const TString file_out output root file name static double TempCritic critic temperature
```

PUBLIC MEMBER FUNCTIONS

CONSTRUCTOR

AnalysisLattice(const TString& file_input, const TString& file_output)

Parametric Constructor Sets input and output file names

RUN FUNCTION

void run()

Performs the analysis of the previous simulation.

Recreates output file

DRAW FUNCTIONS

TGraphErrors * **drawLattice**(cuint& lattice_number, cuint& x_axis, cuint& y_axis)

returns a TGraphErrors of x and y, integers representing physical qunatities

See the following macro for define statements of x and y

Usually x is TEMPERATURE while y could be ENERGY, MAGNETI-ZATION or SUSCEPTIBILITY

The graph represent a single Simulation's Lattice identified by lattice_number

TGraphErrors * drawLatticeMean(cuint& x_axis, cuint& y_axis)

Returns graph of x and y of physical quantities averaged over all Simulation's Lattices

TGraphErrors * draw(cuint& x_axis, cuint& y_axis)

draw a TMultiGraph containing the graphs for all Lattices obtained with the drawLattice() method

FIT FUNCTIONS

static double analiticX(double * x, double * par)

function used for fitting SUSCEPTIBILITY

has parameters : Critic Temperature , Critical Exponent and a proportional constant

static double **analiticM**(double * x, double * par)

function used for fitting MAGNETIZATION

has parameters : Critic Temperature , Critical Exponent and a proportional constant

void findTcritic()

sets TempCritic to the temperature for which suscpetibility is maximum in the TGraphErrors returned by **drawLatticeMean**(TEMPERATURE,SUSCEPTIBILIT void **fitLattice**(bool mean, cuint& x_axis, cuint& y_axis, double fit_temp_min, double fit_temp_max, int lat_number);

Performs fit on MAGNETIZATION or SUSCEPTIBILITY vs TEMPER-

ATURE graph

respectively with fit funcion analiticM or analiticX

if mean==true

fits TGraphErrors returned by drawLatticeMean(x_axis,y_axis)

if mean==false

fits TGraphErrors returned by drawLattice(lat_number.x_axis,y_axis)

The fit is performed in [temp_min,temp_max] range

BIN FUNCTIONS

std::vector<double> bin(const std::vector<double>& vec)

returns std::vector with half size of vec according to binning procedure std::vector<double> binN(cuint& num_bin, const std::vector<double>& vec)

returns std::vector from applying bin on vec num_times

TGraph * evalBinning(cuint& nb)

returns a graph representig energy standard deviation versus bin level nb $\operatorname{GETTERS}$ AND SETTERS

Getters and Setters are implemented for all the private Data Members

```
* This macro explains how to perform a complete simulation
  * using th libraries.
4
5
7 #include "SimulationLattice.h" // Lattice.h included
8 #include "AnalysisLattice.h" //
9 #include "TString.h"
                                 // Other useful root inclusions
10 #include "TStopwatch.h"
                                  11
11
12 {
   const unsigned int lattice_size(10);
13
   const unsigned int lattice_dimension(2);
14
    // this way a 10x10 lattice is created
15
    const unsigned int number_of_lattices(5);
16
    // simulation performed simultaneously on 5 lattices
17
    TString simulation_file("simulation_file.root");
18
19
    TString analysis_file("analysis_file.root");
20
    unsigned int iter_pre_simulation(1000000);
    unsigned int iter_for_simulation(500000);
22
    double min_temperature(0.5);
    double max_temperature(3.5);
23
    double steps_of_temperature(30);
24
25
    TStopwatch timer;
26
27
    SimulationLattice s(lattice_size, lattice_dimension,
28
29
                         number_of_lattices, simulation_file,
                         iter_pre_simulation, iter_for_simulation,
30
31
                         min_temperature, max_temperature,
32
                         steps_of_temperature);
33
    * Alternative way to call it:
34
35
    * SimulationLattice s(lattice_size, lattice_dimension,
36
       number_of_lattices);
     * s.setFile(simulation_file);
37
     * s.setIO(iter_pre_simulation);
38
39
     * ...
40
     * s.run();
     */
41
                     // Simulation started with parameters set
42
    timer.Start();
                     // simulation_file always recreated
43
    s.run();
    timer.Stop();
                     11
44
                     11
    timer.Print();
45
46
    AnalysisLattice a(simulation_file, analysis_file);
47
48
    * input and output file must be provided even
49
     * if the output file already exists.
50
     * The output file is recreated only by run()
    * but is called in reading mode by the other
```

```
* data members
53
    */
54
    timer.Start();
                    // Analysis started: parameters got
55
                    // from input file.
    a.run();
56
                    // Output file Recreated.
    timer.Stop();
57
    timer.Print();
58
                     //
59
60
    /*
    * Once the analysis is performed:
61
    * a.draw(TEMPERATURE, MAGNETIZATION);
62
63
    * defined name defined name defined value
64
65
    * ENERGY
                                    1
66
     * TEMPERATURE TEMP
                                     2
67
     * MAGNETIZATION MAG
                                     3
68
     * SITE_ENERGY SENERGY
                                    4
69
     * SUSCEPTIBILITY SUSC
70
                                     5
71
72
73 }
```

Listing 2: Caption

Performance

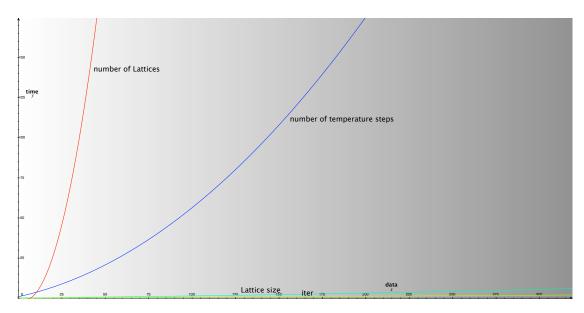


Figure 12: Time scaling for different data member initialization