

Monte Carlo Methods Example: The Ising Model

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The Model I



The Ising model (1) is defined as follows:

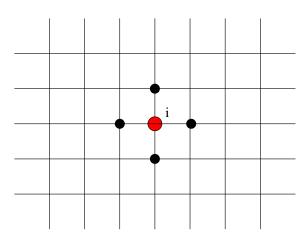
- Let $G = L^d$ be a d-dimensional lattice.
- Associated with each lattice site i is a spin s_i which can take on the values +1 or -1.
- The spins interact via an exchange coupling *J*. In addition, we allow for an external field *H*.
- The Hamiltonian reads

$$\mathcal{H} = -J \sum_{\langle i,j \rangle} s_i s_j + \mu H \sum_i s_i \tag{1}$$

- The first sum on the right-hand side of the equation runs over nearest neighbours only.
- lacktriangleright The symbol μ denotes the magnetic moment of a spin. If the exchange constant J is positive, the Hamiltonian is a model for ferromagnetism, i.e., the spins tend to align parallel.
- For J negative the exchange is anti ferromagnetic and the spins tend to align antiparallel. In what follows we assume a ferromagnetic interaction J > 0.

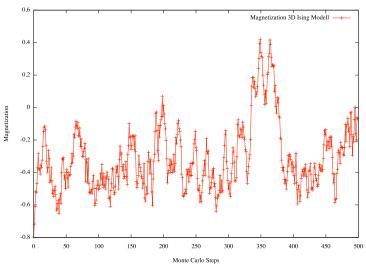
The Model II





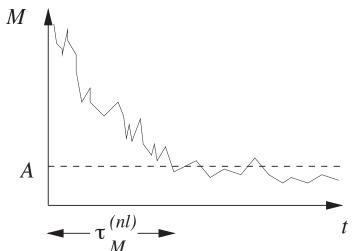
The Model III



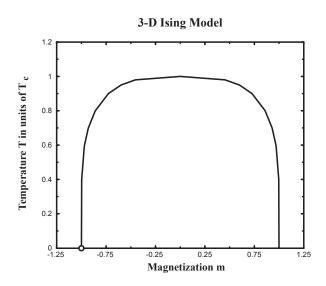


The Model IV









Fixed Energy Monte Carlo I



- Let E be the fixed energy and suppose that a spin configuration $s = (s_1, ..., s_N)$ was constructed with the required energy.
- We set the demon energy to zero and let it travel through the lattice.
- At each site the demon attempts to flip the spin at that site.
- If the spin flip lowers the system energy, then the demon takes up the energy and flips the spin.
- On the other hand, if a flip does not lower the system energy the spin is only flipped if the demon carries enough energy.
- A spin is flipped if

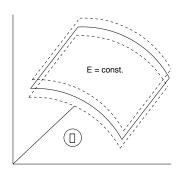
$$E_D - \Delta \mathcal{H} > 0 \tag{2}$$

and the new demon energy is

$$E_D = E_D - \Delta \mathcal{H} \tag{3}$$

Fixed Energy Monte Carlo II





- After having visited all sites one time unit has elapsed and a new configuration is generated.
- In Monte-Carlo method language the time unit is called the MC step per spin.
- After the system has relaxed to thermal equilibrium, i.e., after n_0 Monte-Carlo Steps (MCS), the averaging is started. For example, we might be interested in the magnetization.

Fixed Energy Monte Carlo III



■ Let *n* be the total number of MCS, then the approximation for the magnetization is

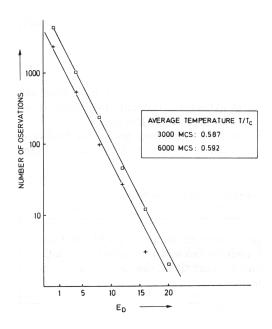
$$\langle m \rangle = \frac{1}{n - n_0} \sum_{i > n_0}^{n} m(s_i)$$
 (4)

where s_i is the ith generated spin configuration. Since the demon energies ultimately become Boltzmann distributed, it is easy to show that

$$\frac{J}{k_B T} = \frac{1}{4} \ln \left(1 + 4 \frac{J}{\langle E_D \rangle} \right) \tag{5}$$

Fixed Energy Monte Carlo IV





Metropolis-Hastings Monte Carlo



- The simplest and most convenient choice for the actual simulation is a transition probability involving only a single spin; all other spins remain fixed.
- It should depend only on the momentary state of the nearest neighbours.
- After all spins have been given the possibility of a flip a new state is created.
 Symbolically, the single-spin-flip transition probability is written as

$$W_i(s_i):(s_1,...,s_i,...,s_N)\longrightarrow (s_1,...,-s_i,...,s_N)$$

where W_i is the probability per unit time that the ith spin changes from s_i to $-s_i$.

■ With such a choice the model is called the single-spin-flip Ising model (Glauber).

Let P(s) be the probability of the state s. In thermal equilibrium at the fixed temperature T and field K, the probability that the i-th spin takes on the value s_i is proportional to the Boltzmann factor

$$P_{eq}(s_i) = \frac{1}{Z} exp\left(\frac{-\mathcal{H}(s_i)}{k_B T}\right)$$

The fixed spin variables are suppressed.

■ We require that the detailed balance condition be fulfilled:

$$W_i(s_i)P_{eq}(s_i) = W_i(-s_i)P_{eq}(-s_i)$$

or

$$\frac{W_i(s_i)}{W_i(s_i)} = \frac{P_{eq}(-s_i)}{P_{eq}(s_i)}$$

It follows that



$$\frac{W_i(s_i)}{W_i(s_i)} = \frac{\exp(-s_i/E_i)}{\exp(s_i/E_i)}$$

where

$$E_i = J \sum_{\langle i,j \rangle} s_j$$

- The derived conditions do not uniquely specify the transition probability W.
- The *Metropolis function*

$$W_i(s_i) = \min\left\{\tau^{-1}, \tau^{-1} \exp(-\Delta \mathcal{H}/k_B T)\right\}$$

and the Glauber function

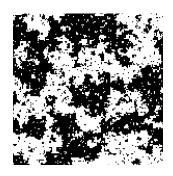
$$W_i(s_i) = \frac{(1-s_i \tanh E_i/k_BT)}{2\tau}$$

where τ is an arbitrary factor determining the time scale.

Algorithmically the Metropolis MC method looks as follows:



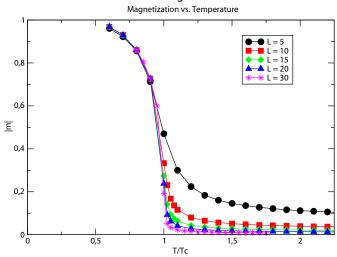
- 1: Specify an initial configuration.
- 2: Choose a lattice site i.
- 3: Compute W_i .
- 4: Generate a random number $R \in [0,1]$.
- 5: if $W_i(s_i) > R$ then
- 6: $s_i \rightarrow -s_i$
- 7: else
- 8: Otherwise, proceed with Step 2 until MCSMAX attempts have been made.
- 9: end if



```
#include <iostream.h>
#include <math.h>
# define L 10
int main(int argc, char *argv[])
    int mcs,i,j,k,ip,jp,kp,in,jn,kn;
   int old spin, new spin, spin sum;
   int old_energy,new_energy;
   int mcsmax;
    int spin(L)(L)(L);
   int seed:
   double r;
   double beta;
   double energy diff;
   double mag;
mcsmax = 100:
beta = 0.12:
                   // beta = J/kT KC = 0.2216544 Talapov and Blöte (1996)
seed = 4711;
srand(seed);
    for (i=0;i<L;i++) {
       for (1=0:1<L:1++) {
           for (k=0:k<L:k++) {
            spin[i][j][k] = -1;
mag = - L*L*L;
   // Loop over sweeps
   for (mcs=0:mcs<mcsmax:mcs++) {
       // Loop over all sites
       for (i=0;i<L;i++) {
           for (j=0;j<L;j++) {
                for (k=0:k<L:k++) {
                   // periodic boundary conditions
                ip = (i+1) % L;
                 ip = (i+1) % L;
                kp = (k+1) % L;
                in = (i+L-1) % L;
                 jn = (j+L-1) & L;
                kn = (k+L-1) % L:
                old_spin = spin[i][j][k];
                new_spin = - old_spin;
                   // Sum of neighboring spins
                spin sum = spin[i][jp][k] + spin[ip][j][k] +
                spin[i][jn][k] + spin[in][j][k] +
                spin[i][j][kn] + spin[i][j][kp];
                old energy = - old spin * spin sum:
                new energy = - new spin * spin sum;
                energy diff = beta * (new energy - old energy);
```



3D Ising Model

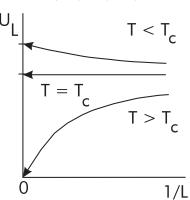


Cumulant I



$$U_L = 1 - \frac{\langle s^4 \rangle_L}{3\langle s^2 \rangle_I^2} \tag{6}$$

$$U(T,L) = b(\epsilon L^{1/\nu}) \tag{7}$$



Global Algorithms



See lecture on cluster and multi-grid algorithms

Literature I



- [1] E. Ising: Z. Phys. 31,253 (1925)
- [2] M. Creutz: Phys. Rev. Lett. 50, 1411 (1983)