

Simulations in Statistical Physics

Course for MSc physics students

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Parallelization

- ▶ Why?
 - ▶ The speed of one core processor is limited
 - ▶ Larger system sizes
 - ▶ Multi-core processors
 - ▶ On multi-core system inter-processor data change is fast
- ▶ Why not?
 - ▶ Computing power is lost
 - ▶ **Much more code development**
 - ▶ Very often ensemble average is needed
 - ▶ Inter-computer communication is terribly slow

RAM \rightarrow \sim 15GB/s, Ethernet 125MB/s, Infiniband \sim 1GB/s

Simple parallelization

- ▶ Multi-threading:
 - ▶ Code is copied to multiple processors
 - ▶ Memory is shared → no need to copy data between processors
 - ▶ Using semaphores to protect data overwrite
 - ▶ Easy to do but unusable on clusters
- ▶ E.g. BOOST:
 - ▶ Simple parallelization of loops
 - ▶ No history dependence
 - ▶ Example:

```
#include <string>
#include <iostream>
#include <boost/foreach.hpp>
```

```
int main()
{
    std::string hello( "Hello, world!" );

    BOOST_FOREACH( char ch, hello )
    {
        std::cout << ch;
    }

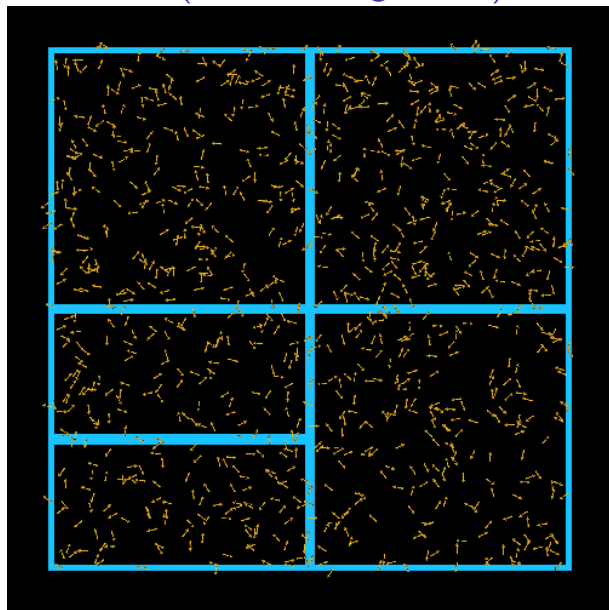
    return 0;
}
```

Message passing interface MPI

- ▶ A given number of copies of the code across processors and machines
- ▶ All processors know their id and the total number of processors
- ▶ Point-to-point communication: synchron and acynchron
- ▶ Gathering data
- ▶ Master-slaves, or real parallel, sharing only parts of the system

```
sprintf(buff, "Hello %d! ", i);  
MPI_Send(buff, BUFSIZE, MPI_CHAR, i, TAG, MPI_COMM_WORLD);  
  
MPI_Recv(buff, BUFSIZE, MPI_CHAR, i, TAG, MPI_COMM_WORLD, &stat);  
  
MPI_Allreduce(&locf, &sumf, 6, MPI_DOUBLE, MPI_SUM, world);
```

Parallelization (Bird flocking model)

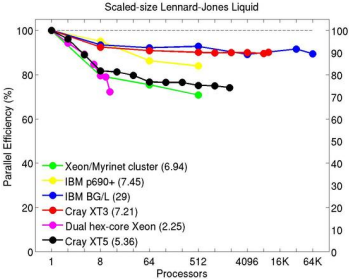
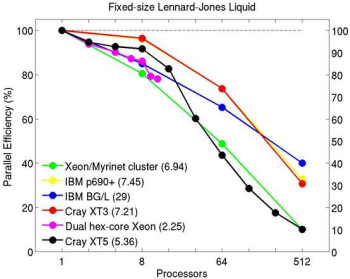


Parallelization

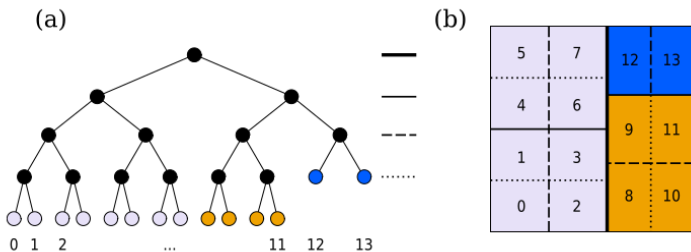
- ▶ Molecular dynamics
 - ▶ Short range interactions: Box must be duplicated, Verlet in parallel
 - ▶ Long range: Parallel fast Fourier transformation
- ▶ Contact dynamics
 - ▶ Short range interactions: Box must be duplicated
 - ▶ Iteration in parallel
- ▶ Event Driven Dynamics
 - ▶ List must be global, no way!
- ▶ Kinetic Monte Carlo
 - ▶ List must be global, no way!

Efficiency of parallelization

- ▶ Large systems are needed
- ▶ Boundary must be minimal



Efficiency of parallelization



- ▶ Calculate time spent in a branch
- ▶ Calculate $\sigma_T = \sqrt{\langle T^2 \rangle - \langle T \rangle^2} / \langle T \rangle$
- ▶ Move line if necessary ($\sigma_T > \sigma_T^*$)
- ▶ Lower in tree (up in Fig), larger the mass of the border
- ▶ Only rarely, data transfer is expensive

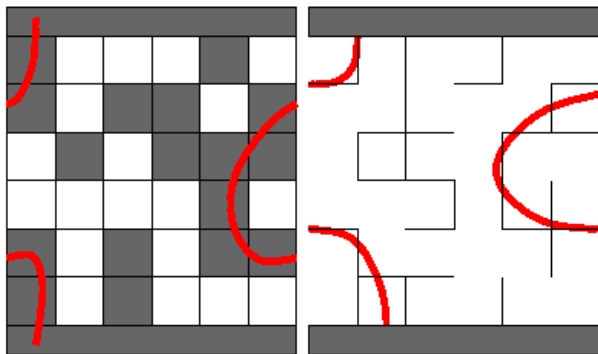
Percolation



Percolation

Behavior of **connected** cluster

- ▶ Site percolation
- ▶ Bond percolation



Percolation theory

Questions (in infinite systems):

1. Is there an infinite cluster in infinite systems?
2. How many infinite clusters are there?
3. Mean cluster size (without the infinite one)?
4. Cluster size distribution

Answers:

1. Above a critical density with probability 1 below it with probability 0
2. Only 1!
3. Decreases as a power law away from the critical density
4. Power law

Percolation theory

Questions (in infinite systems):

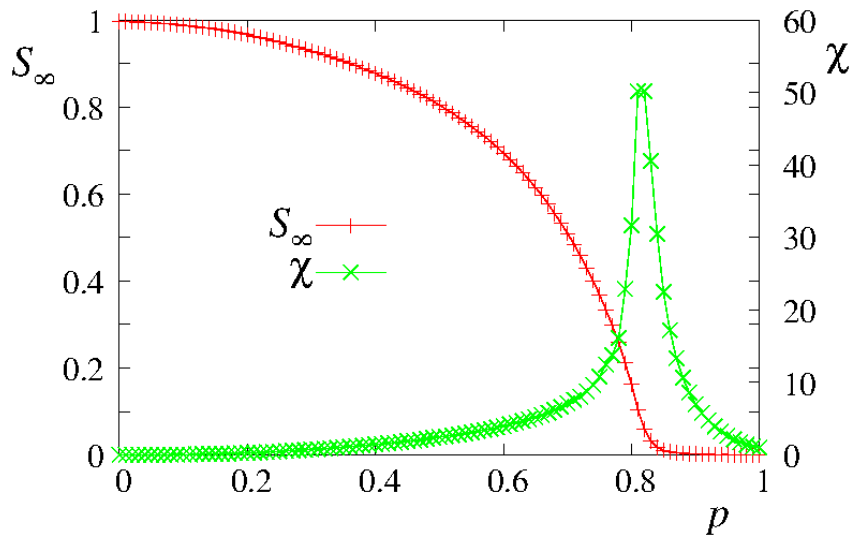
1. Is there an infinite cluster in infinite systems?
2. How many infinite clusters are there?
3. Cluster size distribution (n_s)
4. Mean cluster size (without the infinite one)? ($S = \sum_s s^2 n_s$)

Answers:

1. if $p > p_c$ then yes, otherwise no
2. Only 1!
3. $n_s \sim s^{-\tau}$
4. $S \sim |p - p_c|^{-\gamma}$

Like a second order phase transition in a geometric system!

Percolation model



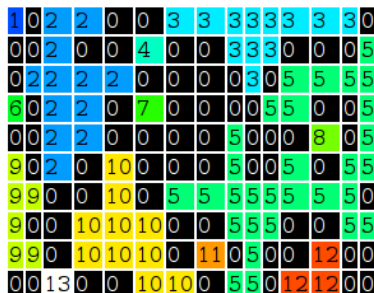
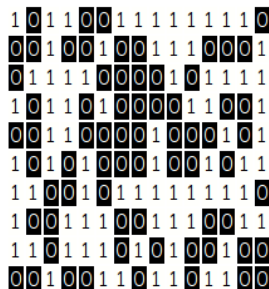
Percolation model

Bond [site] percolation

- ▶ Let us have a lattice (network)
- ▶ Each bond [site] is occupied with probability p
- ▶ (unoccupied with probability $1 - p$)
- ▶ A cluster is a set of sites connected by occupied bonds
[A cluster is a set of occupied sites]

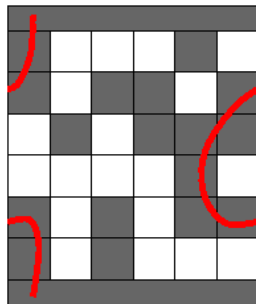
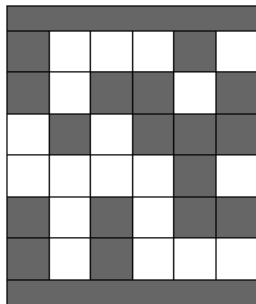
Hoshen-Kopelman Algorithm

- ▶ Numerical task: find clusters
- ▶ Identify clusters
- ▶ Visit all sites
- ▶ Mark them with numbers

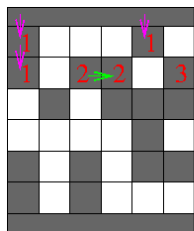


Hoshen-Kopelman Algorithm

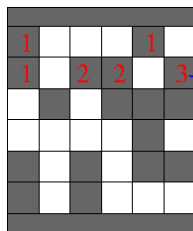
- ▶ Site percolation
- ▶ Helical boundary conditions
- ▶ Go through site in typewriter style
- ▶ Check left and above



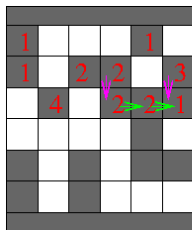
Hoshen-Kopelman Algorithm, Helical BC



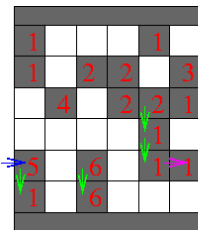
link[1]=1
link[2]=2
link[3]=3



link[1]=1
link[2]=2
link[3]=1



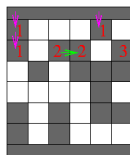
link[1]=1
link[2]=1
link[3]=1
link[4]=4



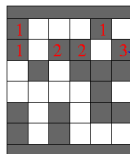
link[1]=1
link[2]=1
link[3]=1
link[4]=4
link[5]=1
link[6]=6

Hoshen-Kopelman Algorithm

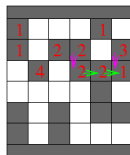
```
largest_label = 0;
for (y = 0; y < n_rows; y++) {
  for (x = 0; x < n_columns; x++) {
    if (occupied[x][y]) {
      left = occupied[x-1][y];
      above = occupied[x][y-1];
      if (left == 0) && (above == 0) {
        largest_label ++;
        label[x][y] = largest_label;
      } else if (left != 0) && (above == 0) {
        label[x,y] = find(left);
      } else if (left == 0) && (above != 0) {
        label[x,y] = find(above);
      } else {
        label[x,y] = union(left, above);
      }
    }
  }
}
/* Helical boundary conditions */
if (occupied[n_columns-1][y]) && (occupied[0][y]) {
  union(occupied[n_columns-1][y], occupied[0][y])
}
}
```



link[1]=1
link[2]=2
link[3]=3



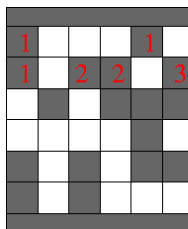
link[1]=1
link[2]=2
link[3]=1



link[1]=1
link[2]=1
link[3]=1
link[4]=4

Hoshen-Kopelman Algorithm

```
largest_label = 0;
for x in 0 to n_columns {
  for y in 0 to n_rows {
    if occupied[x,y] then
      left = occupied[x-1,y];
      above = occupied[x,y-1];
      if (left == 0) and (above == 0) then
        largest_label = largest_label + 1;
        label[x,y] = largest_label;
      else {
        if (left != 0) {
          if (right != 0)
            UNION(left,above);
          label[x,y] = FIND(above);
        } else
          label[x,y] = FIND(right);
      }
    }
  }
}
```



link[1]=1
link[2]=2
link[3]=1

```
int link[N];

int find(int x) {
  while (link[x] != x)
    x = link[x];
  return x;
}
```

```
int union(int x, int y) {
  int fx = find(x);
  int fy = find(y);
  if (fx < fy) {
    link[fy] = fx;
    return (fx);
  } else {
    link[fx] = fy;
    return (fy);
  }
}
```

Hoshen-Kopelman Algorithm

- ▶ Go through lattice as typewriter
- ▶ Check neighbors
- ▶ Resolve conflicts by linking clusters together
- ▶ Original trick: use `link[]` array for cluster size measure
 - ▶ `link[]` positive: number of sites in the cluster
 - ▶ `link[]` negative: cluster is linked to on other cluster
 - ▶ Not necessary faster than a separate array for size

Percolation on networks (graphs)

- ▶ Network is defined by nodes and links
- ▶ Two arrays:
 - ▶ `node[]` list of nodes
 - ▶ `link[i][]` list of links of node i
 - ▶ `link[i][j]` is a link between i and j
- ▶ Cluster: nodes connected with links
- ▶ Links can be directed `link[i][j]` is a link from $i \rightarrow j$

Stack (Verem – Hole/Pitfall)

- ▶ Last in first out (LIFO)
- ▶ Code:

```
int Stack_size = Hopefully_large_enough_number;  
int stack[Stack_size];  
int sp=0;  
  
void push(int item) {  
    stack[sp++] = item;  
    if (sp == Stack_size) enlarge_array(stack);  
}  
  
int pop() {  
    return(stack[--sp]);  
}
```

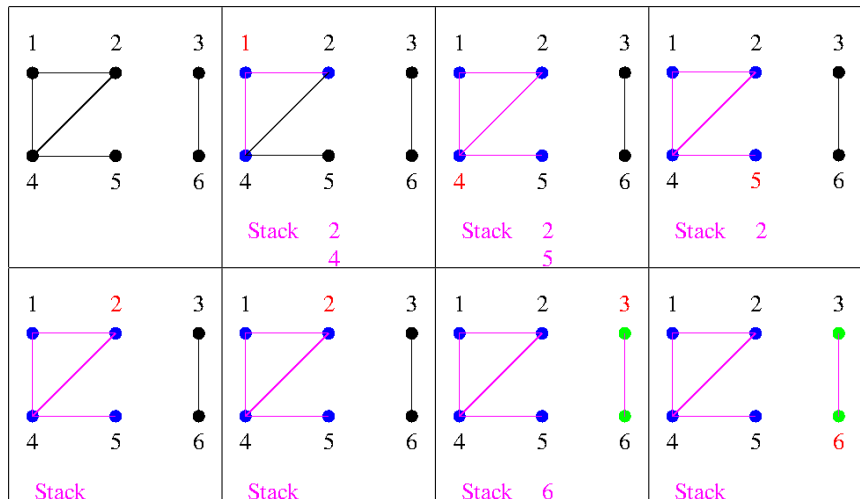
- ▶ Error handling?
- ▶ Size of the stack?



Algorithm percolation on networks (graphs)

1. Go through each node
2. Put node in the stack
3. Get a node from the stack
4. Go through each unmarked link of the node
5. Put other end of links in the stack if it is not marked
6. Mark nodes
7. If the stack not empty Go to 3.
8. If the stack empty Go to 1.

Algorithm percolation on networks (graphs)



Algorithm percolation on networks (graphs)

```
int node[N];
int nlnk[N];
int link[N][N];
int stack[N];
int sp;
```

```
void percol() {
  int a,b,i;
  int cluster;
```

```
  sp = 0;
  cluster = 1;
```

```
  for (a = 0; a < N; a++) node[a]=0;
```

```
  for (a = 0; a < N; a++) {
```

```
    if (node[a] == 0) {
      stack[sp++] = a;
      node[a] = cluster++;
```

```
    }
    while (sp > 0) {
```

```
      i = stack[--sp];
```

```
      for (b = 0; b < nlnk[i]; b++) {
```

```
        if (node[b] == 0) {
```

```
          stack[sp++] = b;
```

```
          node[b] = node[a];
```

```
        }
      }
    }
  }
}
```

1. Go through each node

2. Put node in the stack

3. Get a node from the stack

4. Go through each unmarked link of the node

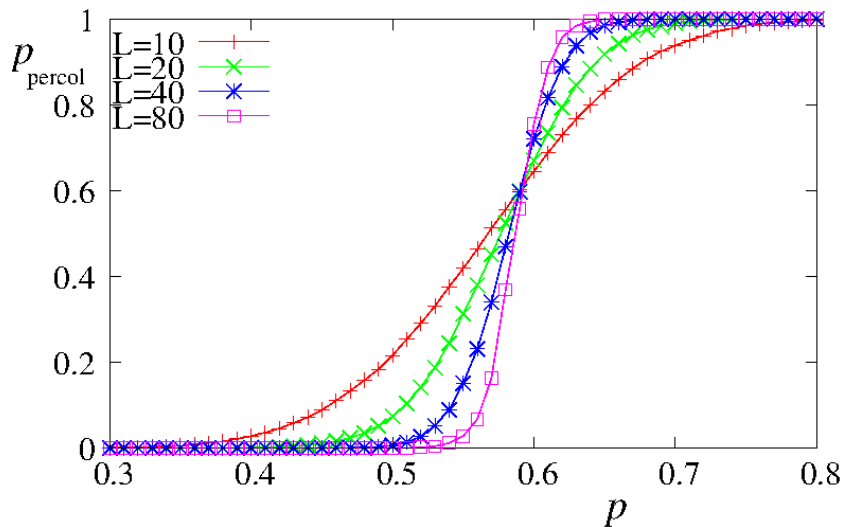
5. Put other end of links in the stack if it is not

6. Mark nodes

7. if the stack not empty Go to 3.

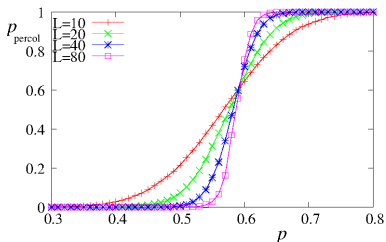
8. if the stack empty Go to 1.

Result

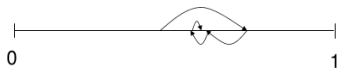


Determine p_c

- ▶ From order parameter:

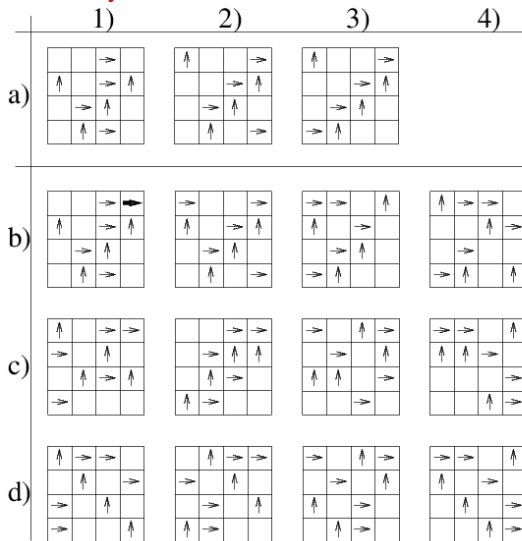


- ▶ Increase and decrease p by $p/2$ to converge to p_c
- ▶ Use the monotonicity of the percolation
- ▶ Same random number sequence can be generated!



Monotonicity

Not always true!



9. ábra: Az a/1 helyen található konfigurációból kiindulva blokkolt határciklushoz jutunk (a/3). A b/1 helyen az a/1 konfigurációhoz hozzávettük még a vastagon kihúzott nyilat, így a b/1-ben a sűrűség nagyobb lett, mint az a/1-ben. Innét indítva a modellt Szabadon mozgó fázishoz jut (d/4).

Ising-model

- ▶ Spins
 - ▶ Interact with external field h_i
 - ▶ Interact with neighbors with coeff. J_{ij}
- ▶ The Hamiltonian:

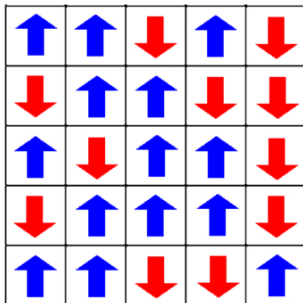
$$H(\sigma) = - \sum_{\langle i j \rangle} J_{ij} \sigma_i \sigma_j - \mu \sum_i h_i \sigma_i$$

- ▶ Order parameter magnetization

$$M = \sum_i \sigma_i$$

2D Ising-model

- ▶ 2 dimensions
- ▶ Homogeneous interaction: $J_{ij} = J$
- ▶ No external field (for the time being) $h = 0$



Importance sampling

- ▶ Given a Hamiltonian $H(\mathbf{q}, \mathbf{p})$
- ▶ We ask for the time average of a dynamics quantity at temperature T

$$\bar{A} = \int A(\mathbf{q}, \mathbf{p}) P^{eq}(\mathbf{q}, \mathbf{p}, T) d\mathbf{q} d\mathbf{p}$$

- ▶ In the canonical ensemble

$$P^{eq}(\mathbf{q}, \mathbf{p}, T) = \frac{1}{Z} e^{-\beta H(\mathbf{q}, \mathbf{p})}$$

- ▶ If A depends only on the energy (often the case):

$$\bar{A} = \int A(E) \omega(E) P^{eq}(E, T) dE$$

Importance sampling is needed!

Importance sampling

- ▶ $\omega(E)P^{eq}(E, T)$ has a very sharp peak (for large N)
- ▶ System spends most of its time *in equilibrium*
- ▶ Importance sampling:

Generate configurations with the equilibrium probability

- ▶ if configurations are chosen accordingly, then for K measurements:

$$\bar{A} \simeq \frac{1}{K} \sum_{i=1}^K A_i$$

How to generate equilibrium configurations?

Metropolis algorithm

(Metropoli-Rosenbluth-Rosenbluth- Teller-Teller=MR²T² algorithm)

- ▶ Sequence of configurations using a Markov chain
- ▶ Configuration is generated from the previous one
- ▶ Transition probability: equilibrium probability
- ▶ Detailed balance:

$$P(x)P(x \rightarrow x') = P(x')P(x' \rightarrow x)$$

- ▶ Rewritten:

$$\frac{P(x \rightarrow x')}{P(x' \rightarrow x)} = \frac{P(x')}{P(x)} = e^{-\beta\Delta E}$$

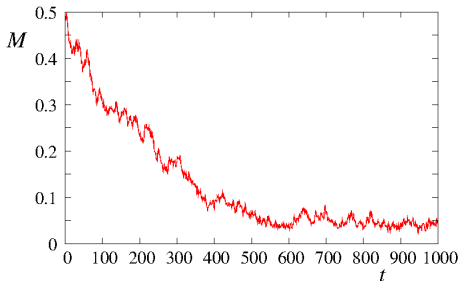
- ▶ Only the ration of transition probabilities are fixed

Characteristic time

- ▶ Equilibrium: system is stationary.
- ▶ We can measure after relaxation time
- ▶ New measurement after correlation time

$$\phi_{EE}(t) = \frac{\langle E(t')E(t'+t) \rangle - \langle E \rangle^2}{\langle E^2 \rangle - \langle E \rangle^2}, \quad \tau = \int_0^{\infty} \phi_{EE}(t) dt$$

- ▶ Sample with intervals $\Delta t > \tau$



Metropolis algorithm

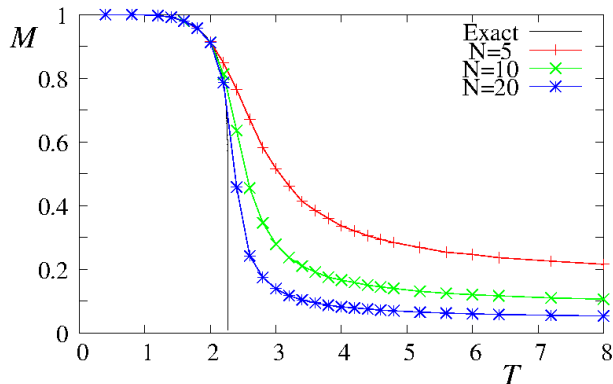
Recipes:

- ▶ Choose an elementary step $x \rightarrow x'$
- ▶ Calculate ΔE
- ▶ Calculate $P(x \rightarrow x')$
- ▶ Generate random number $r \in [0, 1]$
- ▶ If $r < P(x \rightarrow x')$ then new state is x' ; otherwise it remains x
- ▶ Increase time
- ▶ Measure what you want
- ▶ Restart

Finite size effects

Magnetization 2d lattice Ising model

- ▶ Determine critical temperature
- ▶ Determine critical exponents
- ▶ System size dependence???



Finite size scaling

- ▶ Correlation length

$$\xi \propto |T - T_c|^{-\nu}$$

- ▶ If L is finite ξ cannot be larger than L

$$L \propto |T(L) - T_c|^{-\nu}$$

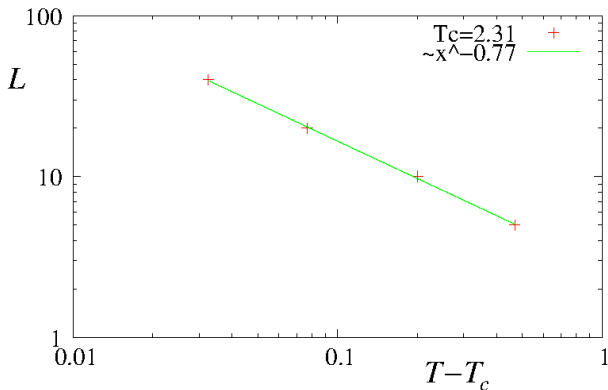
- ▶ The position and the width of the transition

$$|T(L) - T_c| \propto L^{-1/\nu}$$

$$\sigma(L) \propto L^{-1/\nu}$$

Three parameter fit: Ising model

- ▶ Theory: $\nu = 1$, $T_c \simeq 2.27$



Finite size scaling: Ising model

► Theory: $\nu = 1$, $T_c \simeq 2.27$

