

# The Ising Model

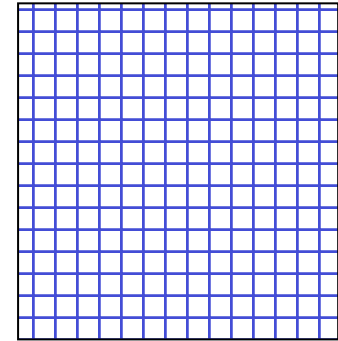
Today we will switch topics and discuss one of the most studied models in statistical physics the **Ising Model**

- Some applications:
  - Magnetism (the original application)
  - Liquid-gas transition
  - Binary alloys (can be generalized to multiple components)
- Onsager solved the 2D square lattice (1D is easy!)
- Used to develop *renormalization group theory* of phase transitions in 1970' s.
- Critical slowing down and “cluster methods”.

*Figures from Landau and Binder (LB), MC Simulations in Statistical Physics, 2000.*

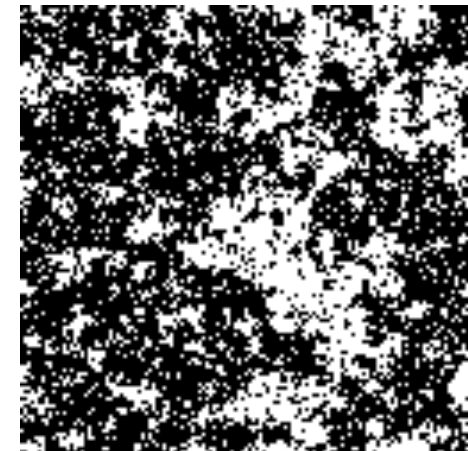
# The Model

- Consider a lattice with  $L^2$  sites and their connectivity (e.g. a square lattice).
- Each lattice site has a single spin variable:  $s_i = \pm 1$ .
- With magnetic field  $\mathbf{h}$ , the energy is:



$$H = -\sum_{(i,j)} J_{ij} s_i s_j - \sum_{i=1}^N h_i s_i \quad \text{and} \quad Z = \sum e^{-\beta H}$$

- $J$  is the nearest neighbors  $(i,j)$  coupling:
  - $J > 0$  ferromagnetic.
  - $J < 0$  antiferromagnetic.
- Picture of spins at the critical temperature  $T_c$ .  
(Note that connected (percolated) clusters.)



# Mapping liquid-gas to Ising

- For **liquid-gas** transition let  $n(r)$  be the density at lattice site  $r$  and have two values  $n(r)=(0,1)$ .

$$E = \sum_{(i,j)} v_{ij} n_i n_j + \mu \sum_i n_i$$

- Let's map this into the Ising model spin variables:

$$s = 2n - 1 \quad \text{or} \quad n = \frac{1}{2}(s + 1)$$

$$H = \frac{v}{4} \sum_{(i,j)} s_i s_j + \frac{(v + \mu)}{2} \sum_i s_i + c$$

$$J = -v / 4$$

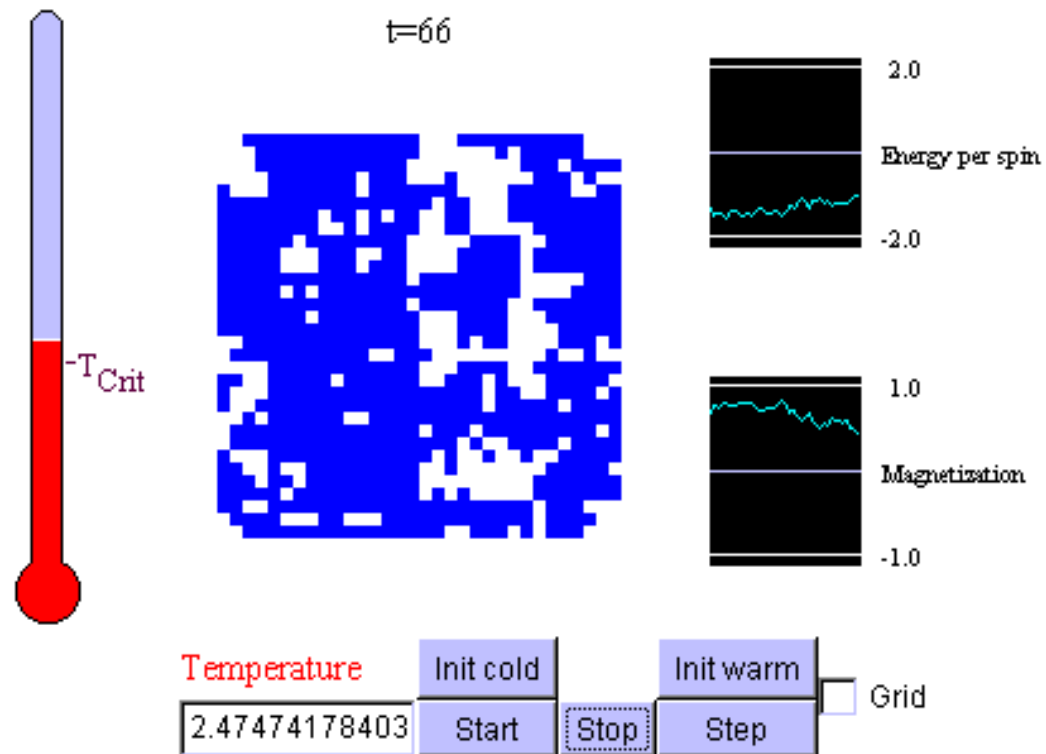
$$h = -(v + \mu) / 2$$

$$M = \frac{1}{N} \sum_i s_i \quad \langle n \rangle = \frac{1}{N} \sum_i n_i = \frac{1}{2}(M + 1)$$

# JAVA Ising applet

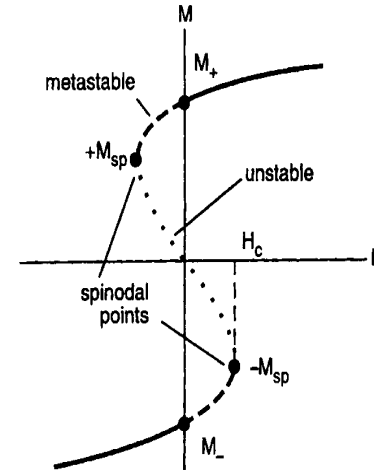
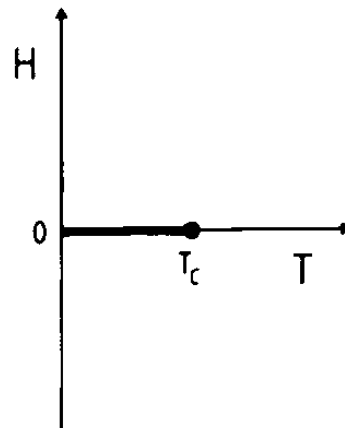
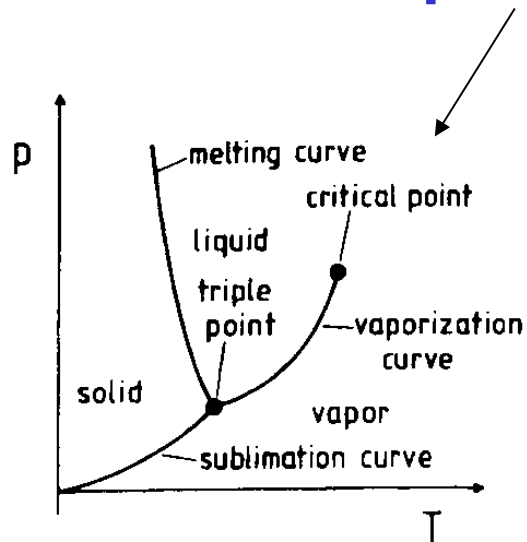
<http://physics.weber.edu/schroeder/software/demos/IsingModel.html>

Dynamically runs using heat bath algorithm.



# Phase Diagram

- **High-T phase:** spins are random (uncorrelated).
- **$T > T_c$  phase near  $T_c$ :** spins are random but correlated: magnetic short-range (local) order.
- **Low-T ( $T \sim 0$ ) phase:** spins are aligned (fully correlated).
- A **first-order transition** (where there is a discontinuous jump in **M**) occurs as **H passes through zero for  $T < T_c$** .
- Similar to **LJ phase diagram**. Magnetic field=pressure.



# Critical point

- Concepts and understanding are universal.  
Apply to all phase transitions of similar type.
- Order parameter is *average* magnetization:  $\langle s(r) \rangle = m(r)$
- Look at correlation function:  $\chi(r-r') = \langle s(r)s(r') \rangle - \langle s(r) \rangle \langle s(r') \rangle$ .
- Magnetic susceptibility is:  $dm(r)/dh(r')|_{h \rightarrow 0} = \beta \chi(r-r')$
- In ordered phase, spin is correlated over long distance.
- At critical point, fluctuations of all scales.

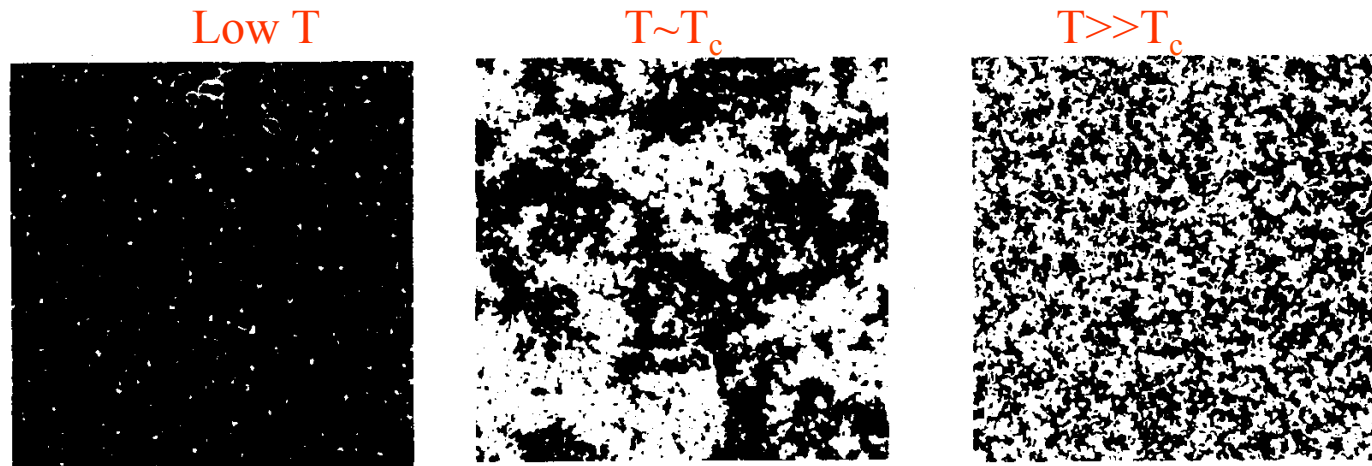


Fig. 4.1 Typical spin configurations for the two-dimensional Ising square lattice: (left)  $T \ll T_c$ ; (center)  $T \sim T_c$ ; (right)  $T \gg T_c$

# Magnetization probability

- How does magnetization vary across transition?
- And with the system size?
- In ordered phase, broken symmetry and barrier to flipping.

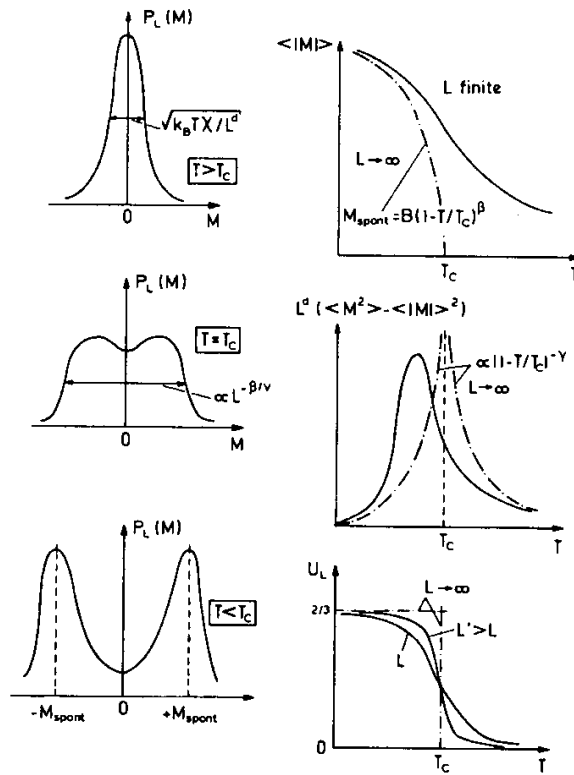


Figure 3. Schematic variation of the probability distribution  $P_L(m)$  to magnetization  $m$  in a finite system of linear dimension  $L$  from  $T > T_c$  to  $T < T_c$  (left part) and the associated temperature variation of the average order parameter  $\langle |m| \rangle$ , “susceptibility”  $k_B T \chi' = L^d (\langle m^2 \rangle - \langle |m| \rangle^2)$  and reduced order cumulant  $U_L = 1 - \langle m^4 \rangle / [3 \langle m^2 \rangle^2]$  (right part).

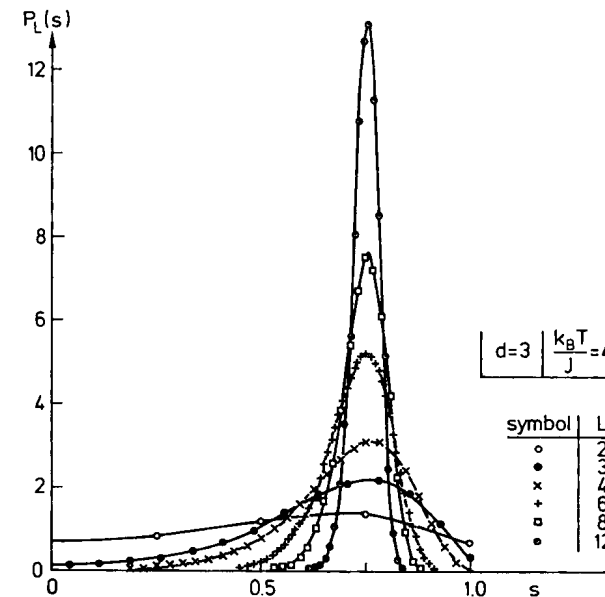


Figure 2. Probability distribution  $P_L(s)$  of the magnetization  $s$  per spin of  $L \times L \times L$  subsystems of a simple cubic Ising lattice with  $N = 24^3$  spins and periodic boundary conditions for zero magnetic field and temperature  $k_B T/J = 4.0$  (note that the critical temperature occurs at about  $k_B T_c/J \approx 4.51$ [26]).

- If we quench too fast we will end in a two phase region.
- The larger the system the sharper the phase transition.

### Phase Diagram: T vs. M

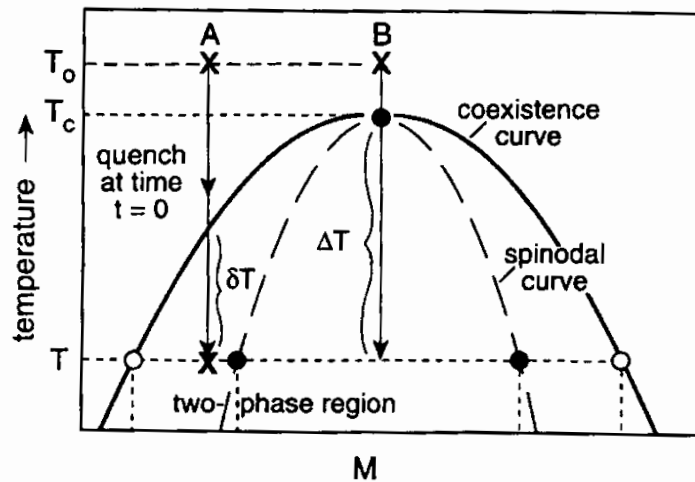
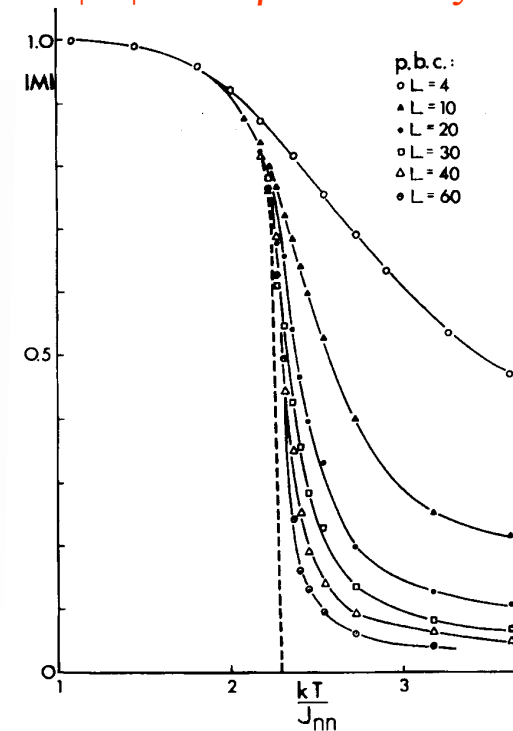


Fig. 2.11 Schematic phase coexistence diagram showing the 'spinodal' line. Paths (A) and (B) represent quenches into the nucleation regime and the spinodal decomposition regime, respectively.

### $|M|$ vs. $1/\beta J$ for varying L



### Magnetization Scaling depends on T:

$$M \sim (T_c - T)^\beta \quad \text{for } T < T_c$$

$$\beta = 0.125 \text{ for } D=2.$$

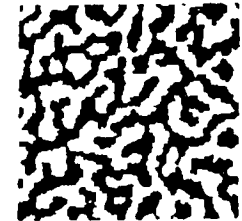
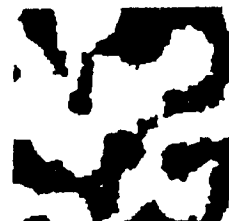
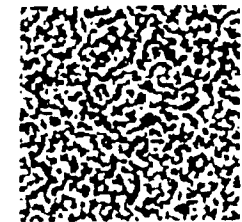
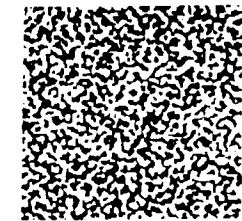
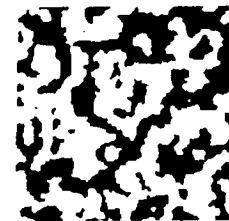
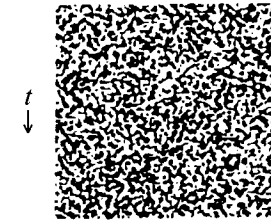
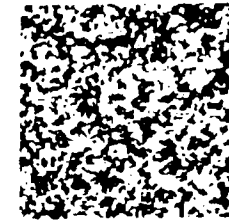
$$\beta = 0.325 \text{ for } D=3.$$



# Spinoidal decomposition

*Suppose spin flips only locally.*

- Model for phase separation such as a binary “alloy” (oil and vinegar).
- Dynamics depends on whether the spin is conserved
  - Spin flip (left)
  - Spin exchange (right). conserves particle number
- Transition appears through a coarsening of the separation.
- Becomes slower and slower as the transition proceeds.
  - **Critical Slowing down.**



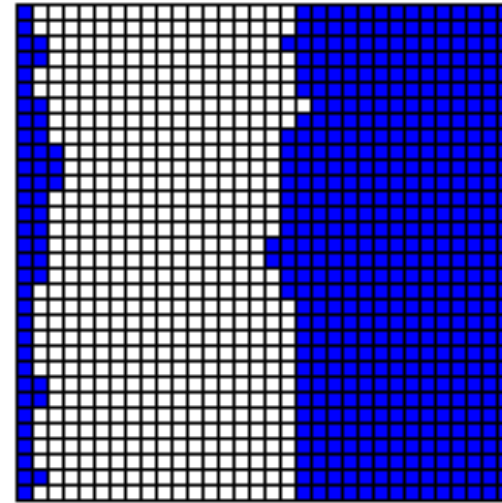
Flip

exchange

$T=0.6T_c$

# Surfaces/Boundary Conditions

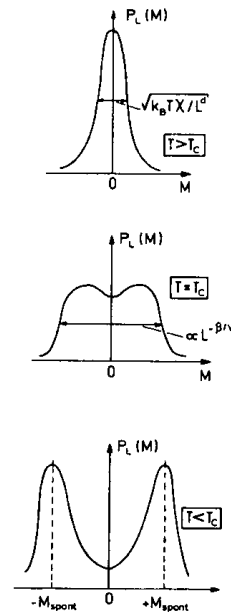
- By quenching quickly we may catch a “trapped” surface.
- Topological excitation.
- You can see steps, etc.
- Can use *twisted boundary conditions* to study a liquid-gas surface without worrying about it disappearing.
- Just put  $-J$  along one plane (side). Antiferromagnetic interaction along one plane.



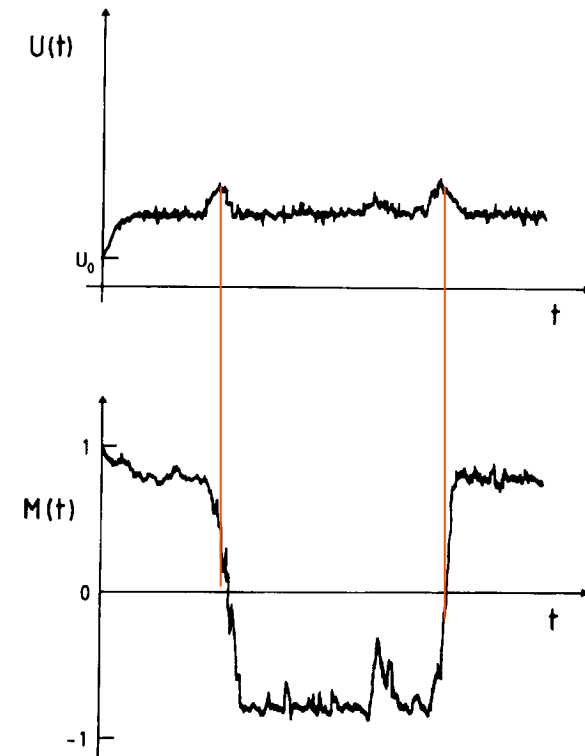
$$H = - \sum_{(i,j)} J_{ij} s_i s_j$$
$$J_{ij} = \begin{cases} J & i \neq 0 \\ -J & i = 0 \end{cases}$$

# Critical slowing down

- Near the transition dynamics gets very slow if you use any local update method.
- The larger the system the less likely it is that the system can flip over.



Monte Carlo of a zero-field Ising Lattice  
 $U$  vs. time and  $M$  vs. time.



## Metropolis importance sampling Monte Carlo scheme

- (1) Choose an initial state
- (2) Choose a site  $i$
- (3) Calculate the energy change  $\Delta E$  which results if the spin at site  $i$  is overturned
- (4) Generate a random number  $r$  such that  $0 < r < 1$
- (5) If  $r < \exp(-\Delta E/k_B T)$ , flip the spin
- (6) Go the next site and go to (3)

# Local versus cluster algorithms

- Simplest Metropolis:
  - Lots of tricks to make it run faster.
  - Tabulate  $\exp(-E/kT)$
  - Do several flips each cycle by packing bits into a word
  - But critical slowing down near  $T_c$ .
  - At low  $T$  accepted flips are rare--can speed up by sampling **acceptance time**.
  - At high  $T$  all flips are accepted--ergodic problem.

## Metropolis importance sampling Monte Carlo scheme

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- (6) Go the next site and go to (3)

# Glauber and Kawasaki dynamics

- **Heat bath or Glauber:**

- Pick a spin and flip with probability
- Will have lower flipping rate but no high T problem.

$$p^i = \frac{\pi_i}{\pi_i + \pi_j} = \frac{1}{1 + e^{-\beta\Delta E}}$$

- **N-fold way:**

- Look at all the sites, choose the site “i” according to:
- The normalization determines how time advances.
- Discuss this later with kinetic MC

$$T^i = \frac{\pi_i}{\sum_j \pi_j}$$

- **Kawasaki dynamics**

- Exchange spins and accept or reject
- Spin is constant as in spinoidal decomposition.

- ALL THESE ARE LOCAL hence suffer from slowdown.

# Swendsen-Wang cluster algorithm

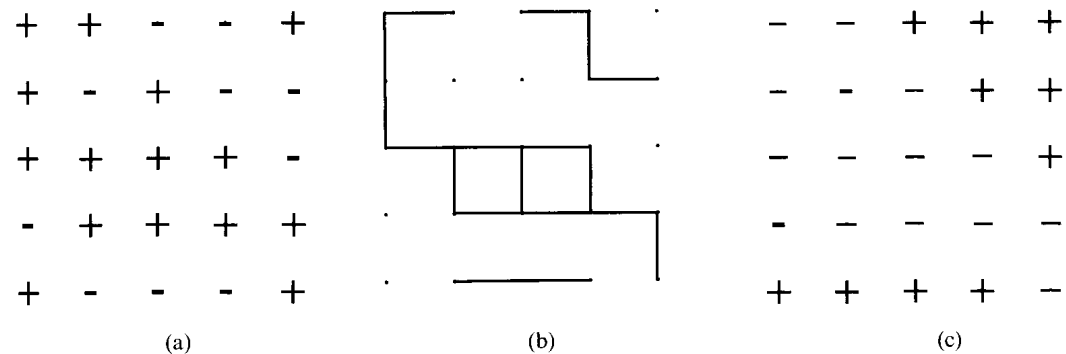


Fig. 5.1 Schematic view of the Swendsen-Wang algorithm for an Ising model: (a) original spin configuration; (b) clusters formed; (c) 'decorated' clusters.

## Wolff cluster flipping method for the Ising model

- (1) Randomly choose a site
- (2) Draw bonds to all nearest neighbors with probability  $p = 1 - e^{-K\delta_{\sigma_i\sigma_j}}$
- (3) If bonds have been drawn to any nearest neighbor site  $j$ , draw bonds to all nearest neighbors  $k$  of site  $j$  with probability  $p = 1 - e^{-K\delta_{\sigma_j\sigma_k}}$
- (4) Repeat step (3) until no more new bonds are created
- (5) Flip all spins in the cluster
- (6) Go to (1)

## Swendsen-Wang algorithm for a $q$ -state Potts model

- (1) Choose a spin
- (2) Calculate  $p = 1 - e^{-K\delta_{\sigma_i\sigma_j}}$  for each nearest neighbor
- (3) If  $p < 1$ , generate a random number  $0 < rng < 1$ ;  
If  $rng < p$  place a bond between sites  $i$  and  $j$
- (4) Choose the next spin and go to (2) until all bonds have been considered
- (5) Apply the Hoshen-Kopelman algorithm to identify all clusters
- (6) Choose a cluster
- (7) Generate a random integer  $1 \leq R_i \leq q$
- (8) Assign  $\sigma_i = R_i$  to all spins in the cluster
- (9) Choose another cluster and go to (7)
- (10) When all clusters have been considered, go to (1)

No critical slowing down at the critical point.

Non-local algorithm. **Prove detailed balance!** See FS 399-408