

Lectures

Exercises/Home work problems

Ordinary differential equations

Linear dynamics

E1

Non-linear dynamics

E2

Molecular dynamics

H1a/H1b

Stochastic methods

Monte Carlo integration

E3

Metropolis algorithm

H2a/H2b

Brownian dynamics

E4

Partial differential equations

Quantum structure

E5

H3a

Quantum dynamics

H3b

Ising model

Content:

- Ising model
- Ising model: Mean field solution
- Ising model: Metropolis algorithm
- Ising model: Numerical results
- Binary alloy: CuZn
- Binary alloy: Mean field solution

Ising model

The Hamiltonian

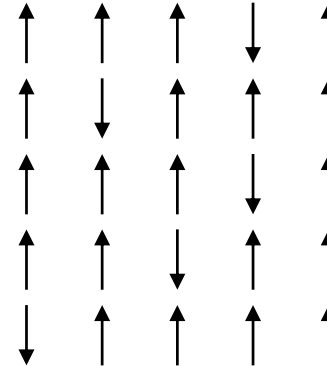
$$\mathcal{H} = -J \sum_{\langle ij \rangle} s_i s_j - h \sum_i s_i$$

$$s_i = \pm 1, \quad i = 1, \dots, N$$

$\langle ij \rangle$ - sum of all nearest neighboring pair of spins

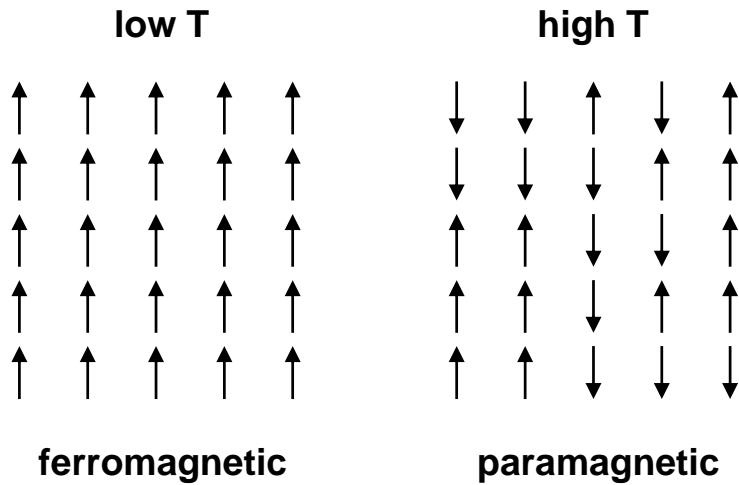
J - coupling constant

h - external field

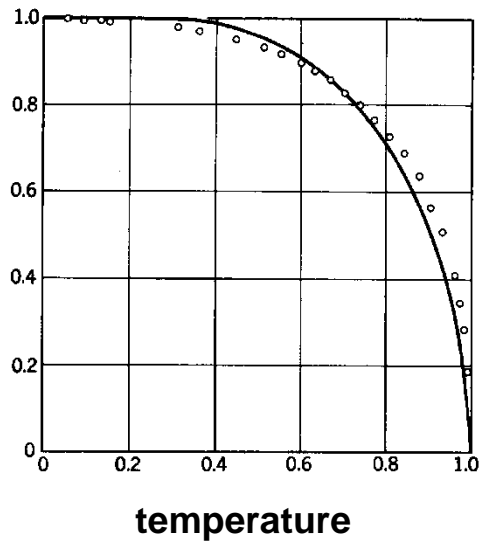


-
- a lattice model
 - one of the simplest and non-trivial model systems of interacting degrees of freedom
 - introduced by Lenz and Ising to model phase transitions in magnetic materials in the 1920s
 - solved exactly in 2D by Onsager 1944
 - has not been solved exactly in 3D (yet)
 - useful in condensed matter physics and field theory
 - a (rather crude) model for magnetism
 - can be used to model binary alloys in materials science
 - can be used to model adsorbed particles in surface science
 - can be extended in many directions

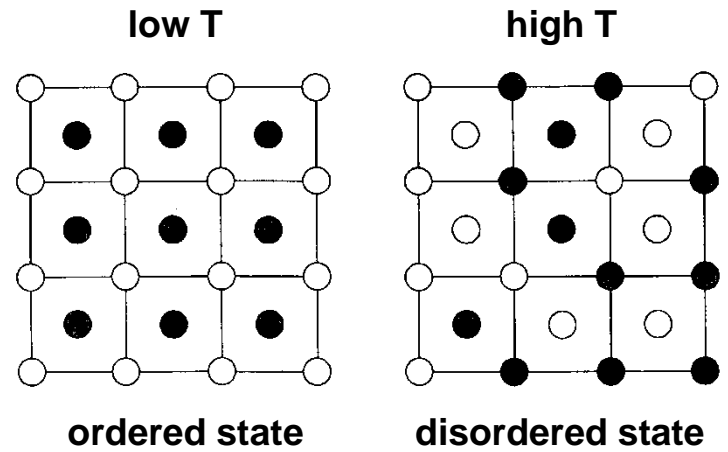
Ferromagnetic substance



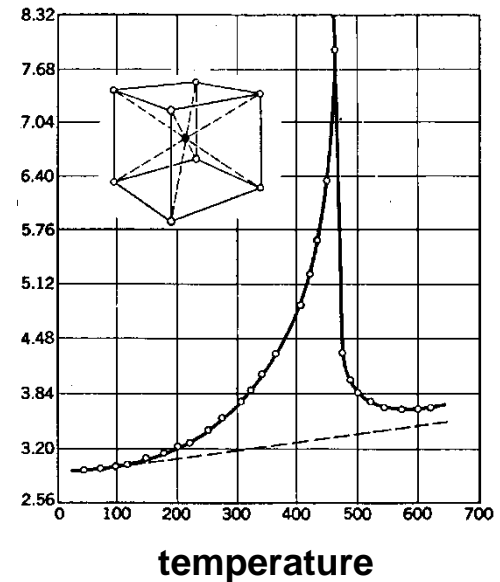
Magnetization



Binary alloy



Heat capacity



Ising model – statistical thermodynamics

Canonical ensemble:

The probability for the system to be in microstate ν :

$$P_\nu = \frac{1}{Z} \exp(-\beta E_\nu)$$

where $\beta = 1/k_B T$.

The partition function:

$$Z(\beta, h) = \sum_\nu \exp(-\beta E_\nu)$$

The magnetization

$$M_\nu = \sum_{i=1}^N s_i$$

The energy

$$E_\nu = -J \sum_{\langle ij \rangle} s_i s_j - h \sum_i s_i$$

The mean magnetization

$$M \equiv \langle M_\nu \rangle = \frac{1}{Z} \sum_\nu M_\nu \exp(-\beta E_\nu)$$

The mean energy

$$U \equiv \langle E_\nu \rangle = \frac{1}{Z} \sum_\nu E_\nu \exp(-\beta E_\nu)$$

The isothermal susceptibility

$$\chi_T \equiv \left(\frac{\partial M}{\partial h} \right)_T = \frac{1}{k_B T} \left[\langle M_\nu^2 \rangle - \langle M_\nu \rangle^2 \right]$$

The heat capacity at constant field

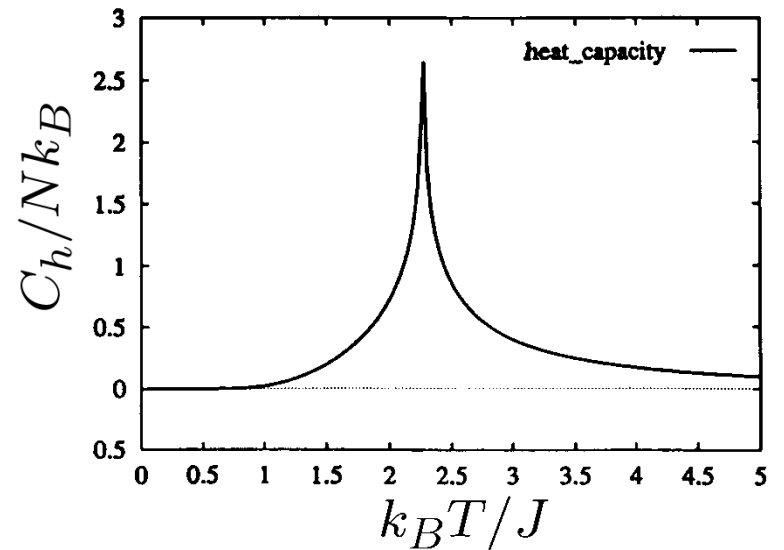
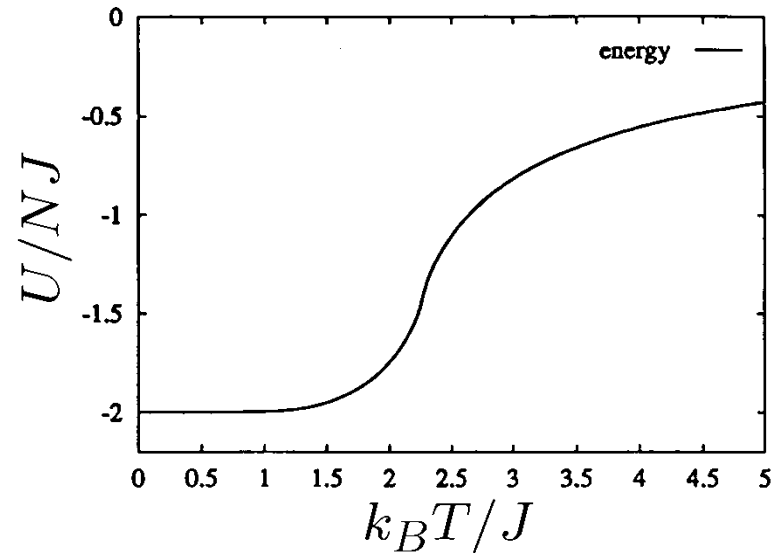
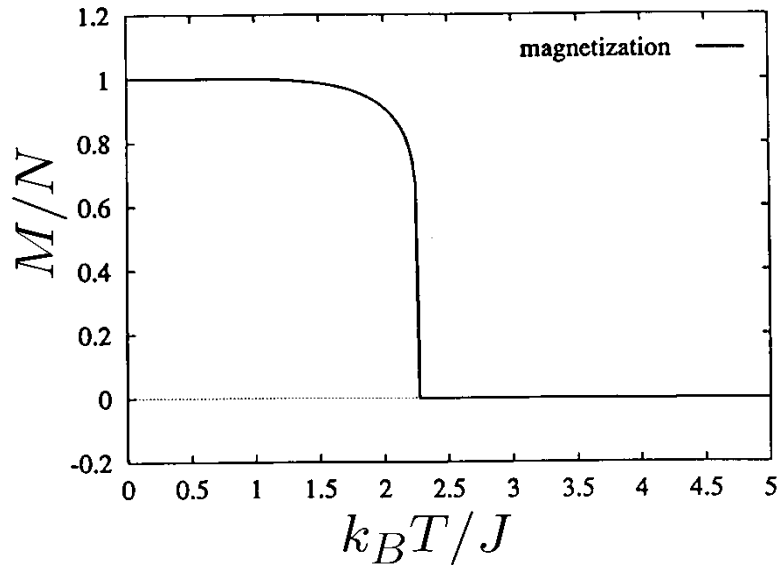
$$C_h \equiv \left(\frac{\partial U}{\partial T} \right)_h = \frac{1}{k_B T^2} \left[\langle E_\nu^2 \rangle - \langle E_\nu \rangle^2 \right]$$

Ising model – the exact solution (2D)

Due to Onsager, 1944

The transition temperature T_c

$$\frac{k_B T_c}{J} = \frac{2}{\ln(\sqrt{2} + 1)} \approx 2.269$$



Ising model – mean field solution (h=0)

The exact solution

The energy

$$U = -J \sum_{\langle ij \rangle} \langle s_i s_j \rangle$$

The mean field solution

The energy

$$U_{MF} = -J \sum_{\langle ij \rangle} \langle s_i \rangle \langle s_j \rangle = -J2Nm^2$$

The number of microstates

$$W = \frac{N!}{N_{\uparrow}!N_{\downarrow}!}$$

The entropy

$$S = k_B \ln W = \frac{N!}{N_{\uparrow}!N_{\downarrow}!}$$

The free energy

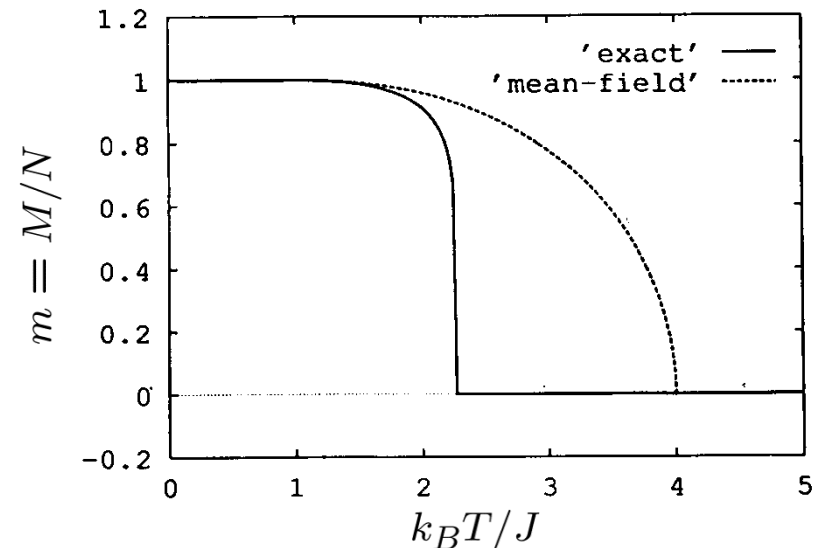
$$F = U - TS$$

Minimum if

$$m = \frac{k_B T}{8J} \ln \frac{1+m}{1-m}$$

which implies that

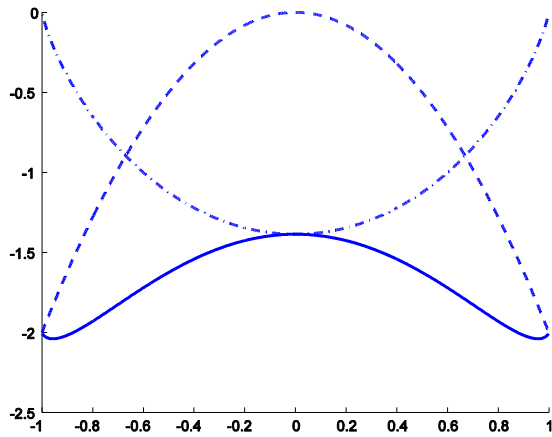
$$k_B T_c / J = 4.0$$



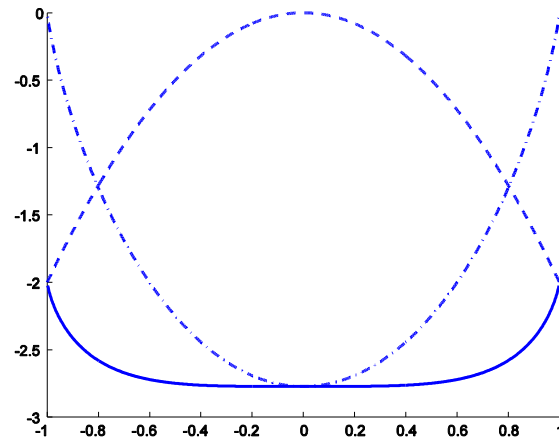
Ising model – mean field solution ($h=0$)

$$\frac{F_{MF}}{NJ} = -2m^2 + \frac{k_B T}{J} \left[\frac{1+m}{2} \ln \frac{1+m}{2} + \frac{1-m}{2} \ln \frac{1-m}{2} \right]$$

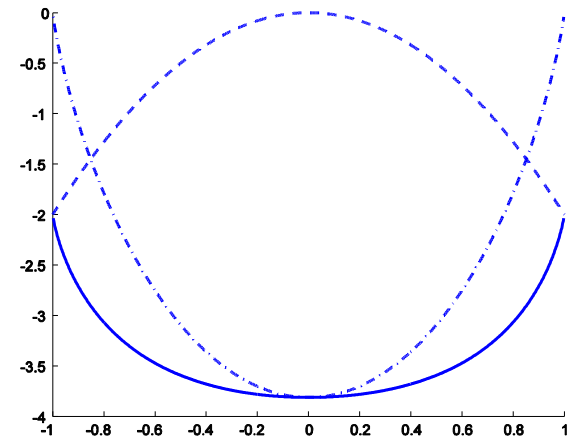
$k_B T/J = 2.5$



$k_B T/J = k_B T_c/J = 4.0$



$k_B T/J = 5.5$



Dashed line: U

Dash-dotted line: $-TS$

Full line: $F = U - TS$

Ising model – Metropolis algorithm

1. Set the desired temperature T and external field h .
 2. Initialize the system, *e.g.* use a random configuration or a configuration from a previous simulation.
 3. Perform the desired number of Monte Carlo sweeps through the lattice.
 4. Exclude the first configurations (let the system equilibrate).
 5. Compute average quantities from subsequent configurations and estimate the error from statistically independent configurations.
- 3a. Make a trial change, *e.g.* by flipping a randomly chosen spin.
 - 3b. Determine the change in energy ΔE
 - 3c. If $\Delta E \leq 0$ accept the new configuration
 - 3d. If $\Delta E > 0$ generate a random number r between 0 and 1, and if
$$\exp(-\Delta E/k_B T) \geq r$$
accept the new configuration, otherwise count the old configuration once more.

The Metropolis algorithm

The Metropolis algorithm is a particular way of ensuring that the transition rule satisfies detailed balance.

The transition matrix $w_{nm} = w_{n \leftarrow m}$ is split into two parts

$$w_{nm} = \tau_{nm} \alpha_{nm}$$

where τ_{nm} is the probability of making a trial change from state Ω_m to state Ω_n and α_{nm} is the probability of accepting the trial state.

The acceptance probability is assumed to satisfy

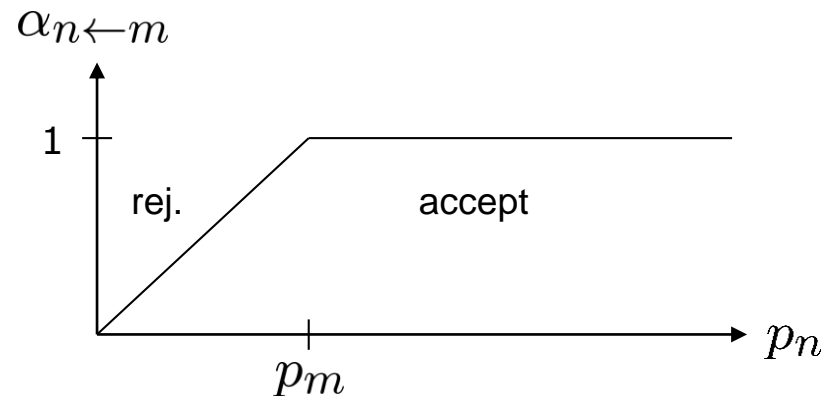
$$\alpha_{nm} = \begin{cases} 1 & \text{if } p_n \geq p_m \\ p_n/p_m & \text{if } p_n < p_m \end{cases}$$

and to ensure detailed balance the trial change then has to be symmetric

$$\tau_{nm} = \tau_{mn}$$

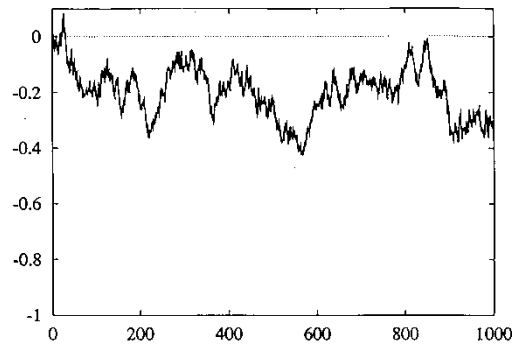
Moves that are not accepted are rejected and remain at the same location for at least one more step

$$w_{mm} = 1 - \sum_{n(\neq m)} w_{nm}$$



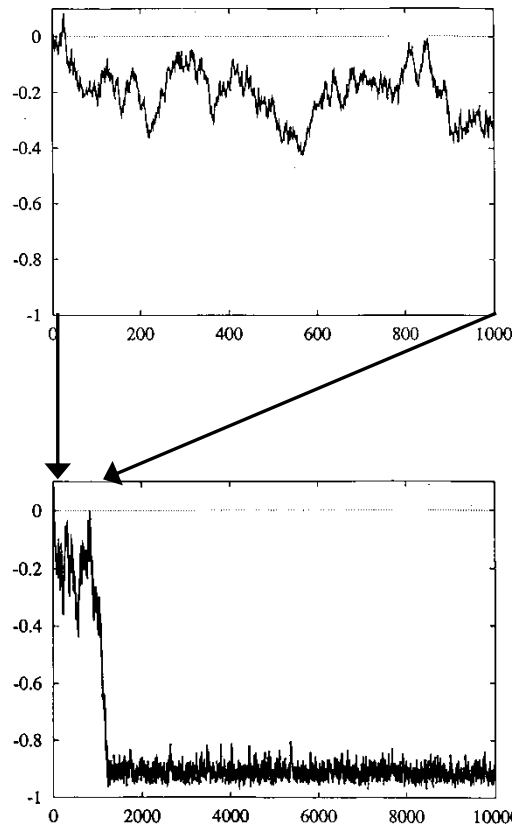
Equilibration

It is important to wait until the system has **equilibrated**.



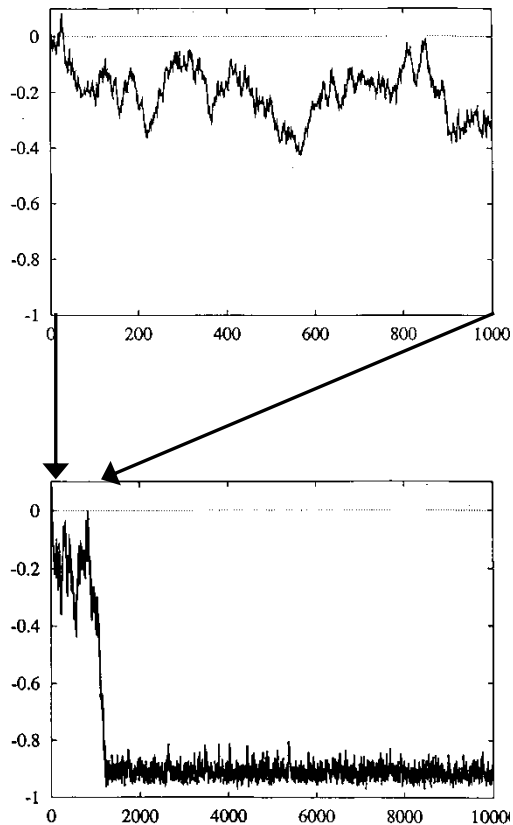
Equilibration

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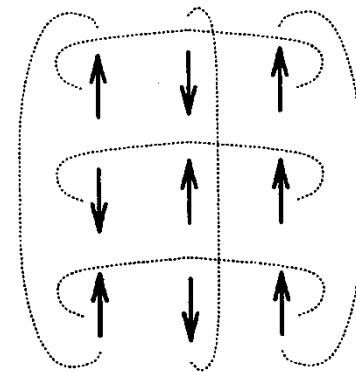
Equilibration

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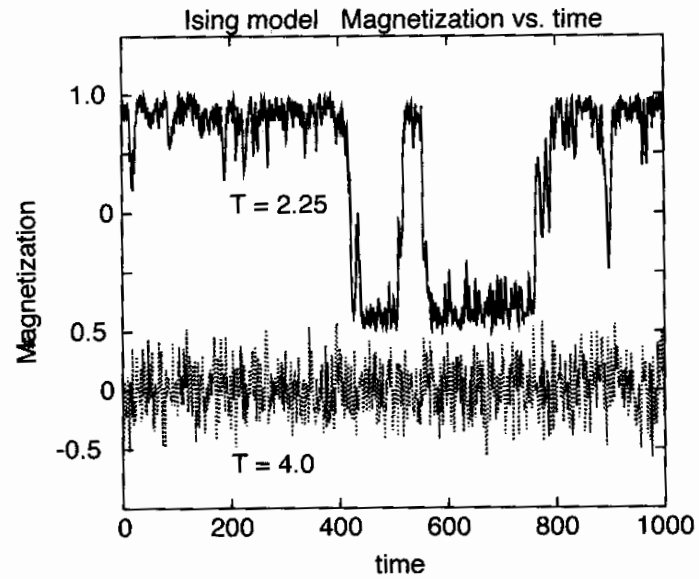
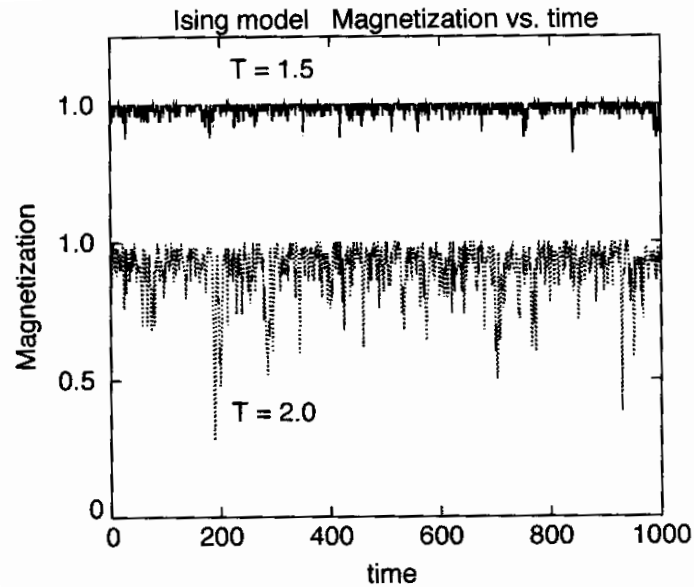


Boundary conditions

To mimic a large system **periodic boundary conditions** are commonly used.

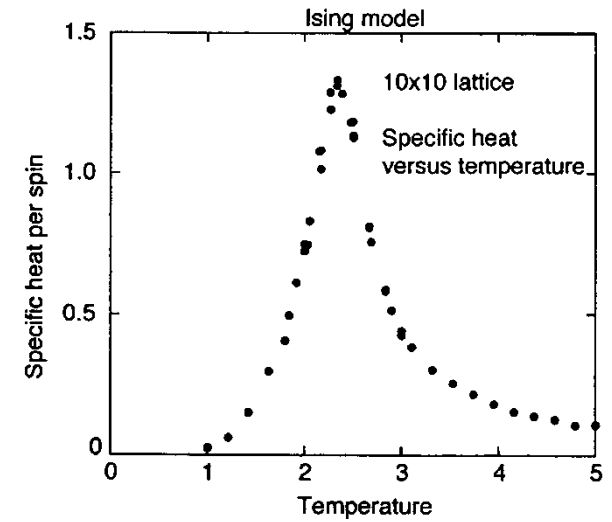
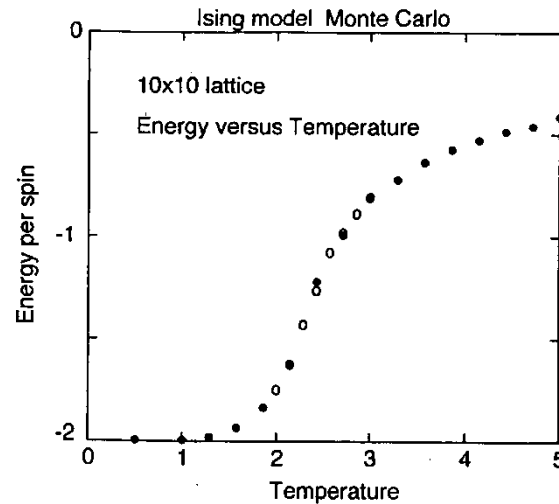
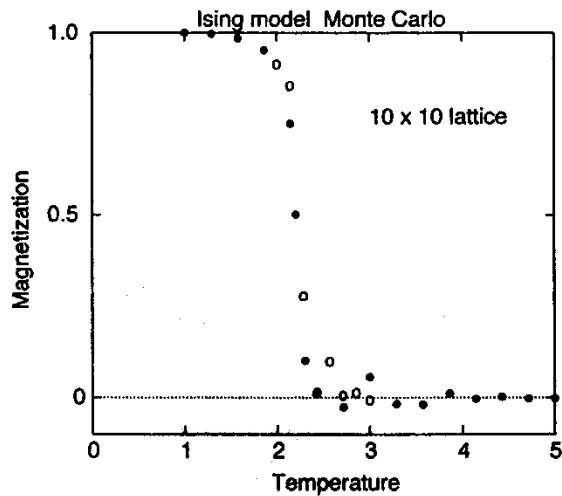


Ising model – Numerical results

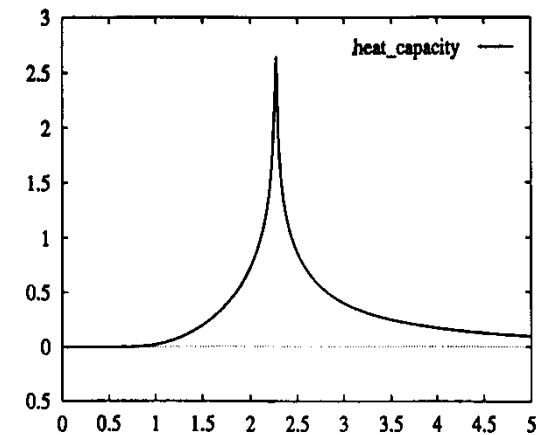
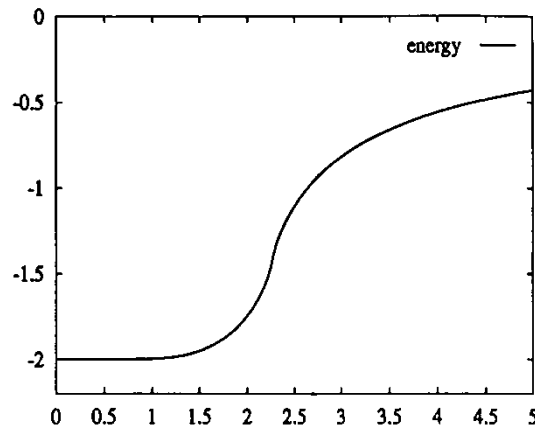
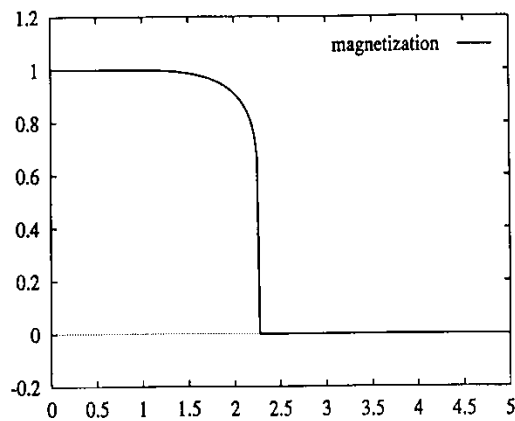


The temperature T in units of J/k_B .

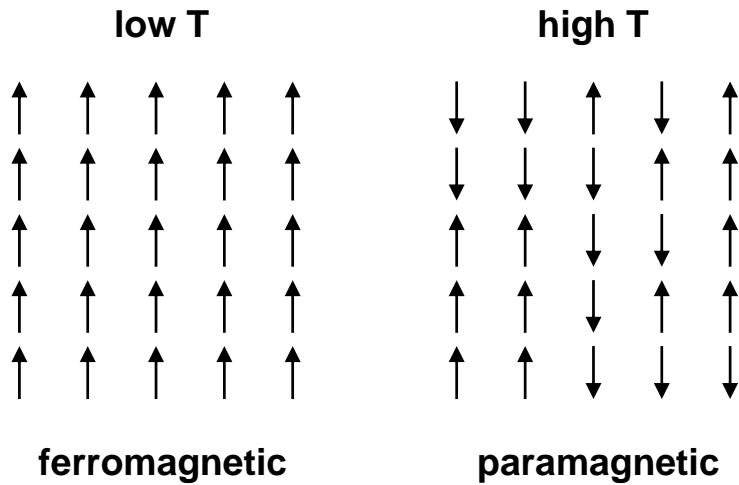
Ising model – Numerical results



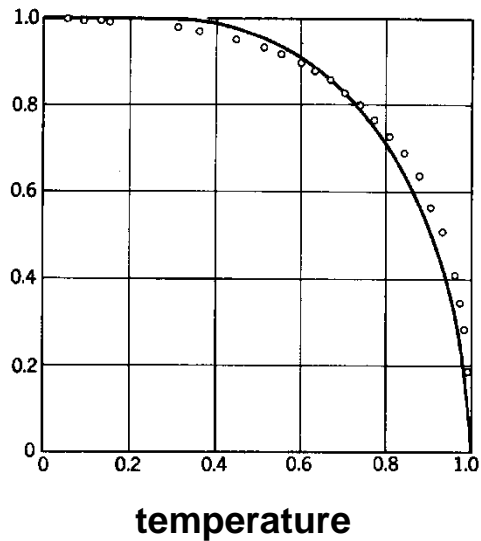
Ising model – Exact results



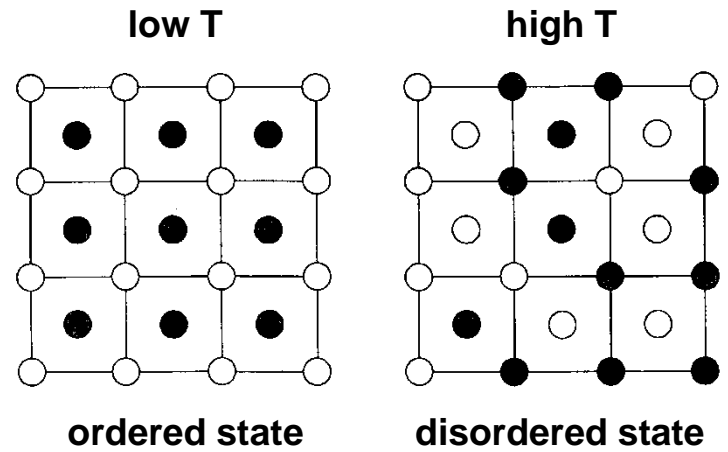
Ferromagnetic substance



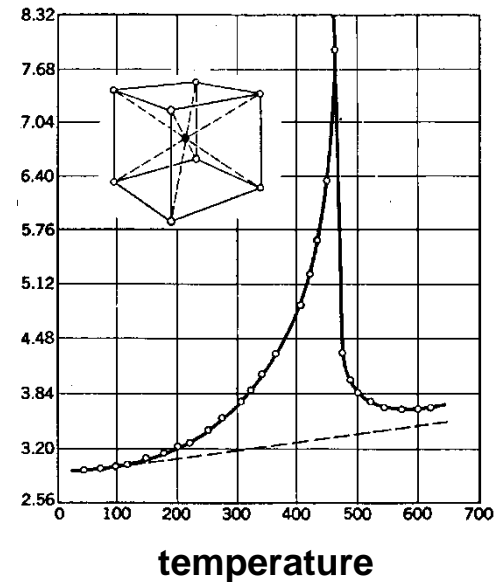
Magnetization



Binary alloy

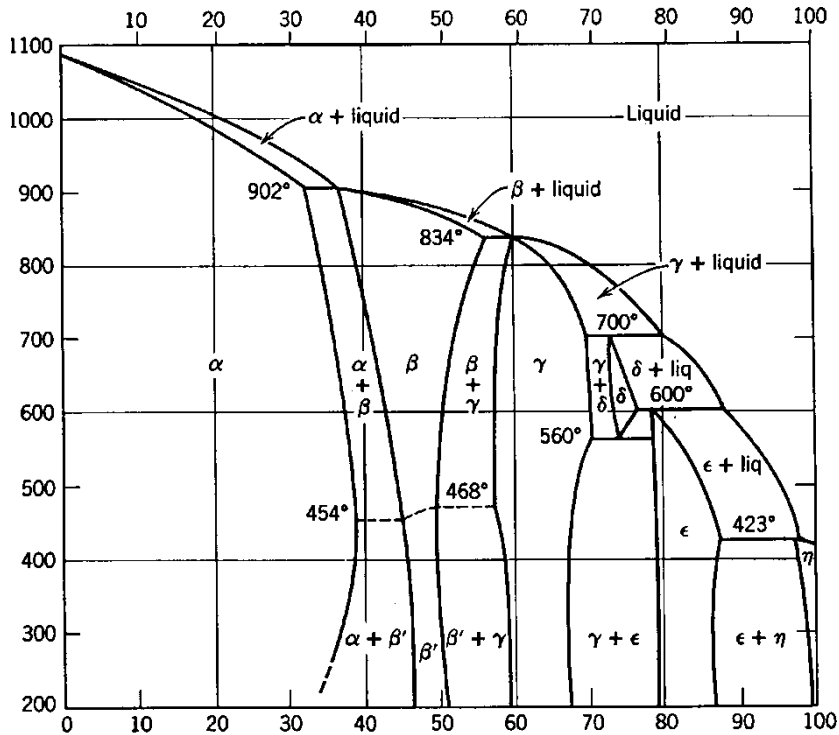


Heat capacity

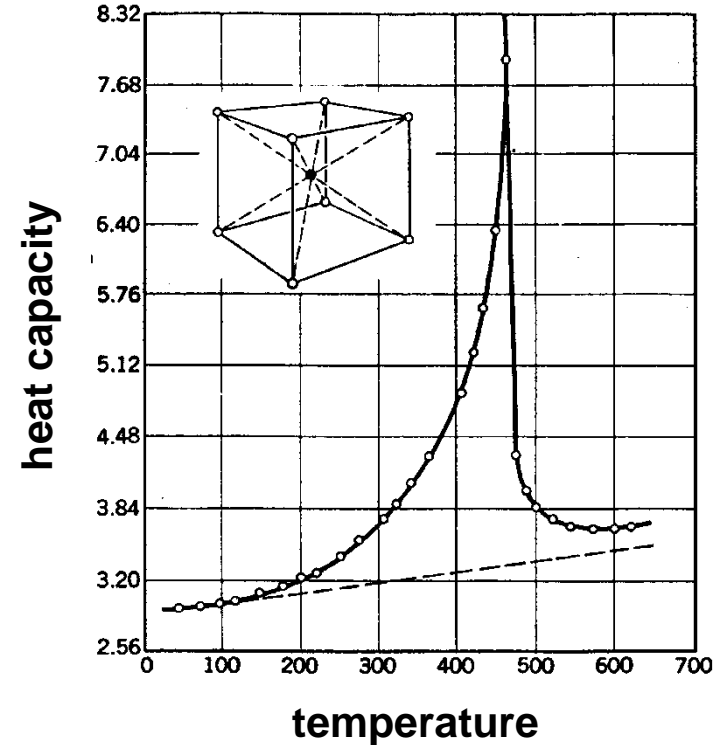


The Cu-Zn system – a binary alloy

Equilibrium phase diagram



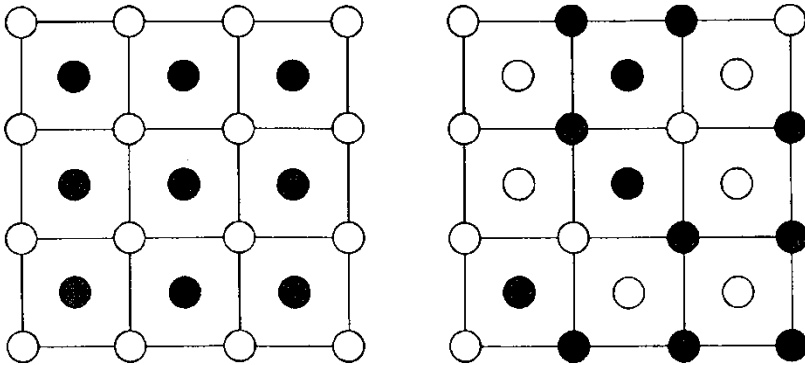
CuZn (β -brass)



Binary alloy – simple model

AB alloy with bcc structure

ex. Cu-Zn (β -brass)



- N atoms A, N atoms B

- bcc-lattice = two interpenetrating sc-lattices (a and b)

Long-range order parameter P

Number of A atoms on the a sublattice
 $= \frac{1}{2}(1 + P)N$

$P = \pm 1$ perfect order

$P = 0$ no order

Short-range order parameter r

q = number of nearest-neighbor bonds
that are AB bonds

$$r = \frac{1}{4}(q - 4)$$

$r = 1$ complete order

$r = 0$ complete disorder

Binary alloy – mean field solution

The energy

$$E = N_{AA}E_{AA} + N_{BB}E_{BB} + N_{AB}E_{AB}$$

where N_{ij} is the number of nearest-neighbor ij bonds, and E_{ij} is the energy of an ij bond.

Mean field approximation

Assume no correlations, i.e.

$$N_{AA} = 8 \left[\frac{1}{2}(1+P)N \right] \left[\frac{1}{2}(1-P) \right] = 2(1-P^2)N$$

$$N_{BB} = 8 \left[\frac{1}{2}(1+P)N \right] \left[\frac{1}{2}(1-P) \right] = 2(1-P^2)N$$

$$N_{AB} = 8N \left[\frac{1}{2}(1+P) \right]^2 + 8N \left[\frac{1}{2}(1-P) \right]^2 = 4(1+P^2)N$$

This implies that

$$E_{MF} = E_0 - 2NP^2\Delta E$$

where

$$E_0 = 2N(E_{AA} + E_{BB} + 2E_{AB})$$

$$\Delta E = E_{AA} + E_{BB} - 2E_{AB}$$

Binary alloy – mean field solution

The number of configurations

$$W = \left[\frac{N!}{\left[\frac{1}{2}(1+P)N\right]! \left[\frac{1}{2}(1-P)N\right]!} \right]^2$$

The entropy

$$\begin{aligned} S &= k_B \ln W \\ &= 2Nk_B \ln 2 - Nk_B [(1+P) \ln(1+P) + (1-P) \ln(1-P)] \end{aligned}$$

The free energy

$$\begin{aligned} F &= U - TS \\ &= E_0 - 2NP^2 \Delta E \\ &\quad - 2Nk_B T \ln 2 + Nk_B T [(1+P) \ln(1+P) + (1-P) \ln(1-P)] \end{aligned}$$

The equilibrium structure is obtained by finding the minimum of F with respect to the order parameter P . It leads to a phase transition at

$$T_c = 2\Delta E/k_B$$

References:

Computational treatment of the Ising model:

- Giordano and Nakanishi, *Computational Physics*
- Koonin and Meredith, *Computational Physics*

Binary alloy:

- Kittel, *Introduction to Solid State Physics*

Advanced treatment of the Ising model:

- Huang, *Statistical Mechanics*

Variational Monte Carlo

Content:

- Variational Quantum Monte Carlo
- Multi-dimensional case
- Example: Helium

Variational Quantum Monte Carlo

Aim: To obtain a good estimate of the ground state energy E_0 for a quantum system.

Based on the variational theorem

$$E[\psi_T] = \frac{\langle \psi_T | \mathcal{H} | \psi_T \rangle}{\langle \psi_T | \psi_T \rangle} \geq E_0$$

where

ψ_T is a trial wave function

Choose a physically reasonable form for the trial wave function that depends on one or more parameters α_i . Determine $E[\psi_T]$ and vary α_i until a minimum is obtained.

Multi-dimensional case

Write in terms of a normalized weight function $\rho(\mathbf{X})$, or probability distribution

$$E[\psi_T] = \frac{\int d\mathbf{X} \psi_T^*(\mathbf{X}) \mathcal{H} \psi_T(\mathbf{X})}{\int d\mathbf{X} \psi_T^*(\mathbf{X}) \psi_T(\mathbf{X})} = \int d\mathbf{X} E_L(\mathbf{X}) \rho(\mathbf{X})$$

where

$$E_L(\mathbf{X}) = \frac{\mathcal{H} \psi_T(\mathbf{X})}{\psi_T(\mathbf{X})}$$

is the called the local energy and

$$\rho(\mathbf{X}) = \frac{|\psi_T(\mathbf{X})|^2}{\int d\mathbf{X} |\psi_T(\mathbf{X})|^2}$$

is the weight function. Use the Metropolis algorithm to evaluate the multi-dimensional integral.

Example: Helium

Hamiltonian in atomic units ($\hbar = m_e = e = 4\pi\epsilon_0 = 1$)

$$\mathcal{H} = -\frac{1}{2}(\nabla_1^2 + \nabla_2^2) - \frac{2}{r_1} - \frac{2}{r_2} + \frac{1}{r_{12}}$$

Trial wavefunction

$$\psi_T(\mathbf{r}_1, \mathbf{r}_2) = \phi(r_1)\phi(r_2)f(r_{12})$$

with

$$\begin{aligned}\phi(r) &= \exp[-\alpha_1 r] \\ f(r) &= \exp\left[\frac{\alpha_2 r}{1 + \alpha_3 r}\right]\end{aligned}$$

Example: Helium

The cusp conditions (boundary conditions)

$$\lim_{r \rightarrow 0} \frac{1}{\phi} \frac{\partial \phi(r)}{\partial r} = -2$$

$$\lim_{r \rightarrow 0} \frac{1}{f} \frac{\partial f(r)}{\partial r} = \frac{1}{2}$$

imply that

$$\psi_T(\mathbf{r}_1, \mathbf{r}_2) = \exp[-2r_1] \exp[-2r_2] \exp\left[\frac{r_{12}}{2(1 + \alpha r_{12})}\right]$$

and that the local energy can be written as

$$E_L(\mathbf{r}_1, \mathbf{r}_2) = -4 + \frac{(\hat{\mathbf{r}}_1 - \hat{\mathbf{r}}_2) \cdot \hat{\mathbf{r}}_{12}}{(1 + \alpha r_{12})^2} + \frac{\alpha}{(1 + \alpha r_{12})} + \frac{\alpha}{(1 + \alpha r_{12})^2} + \frac{\alpha}{(1 + \alpha r_{12})^3} - \frac{1}{4(1 + \alpha r_{12})^4}$$

References:

Variational Quantum Monte Carlo:

- Koonin and Meredith, *Computational Physics*
- Thijssen, *Computational Physics*
- P.J. Reynolds, J. Tobochnik, and H. Gould, Computers in Physics, nov/dec 1990.

Green's function and Diffusion Monte Carlo:

- Koonin and Meredith, *Computational Physics*
- Thijssen, *Computational Physics*