

$$\beta = \frac{1}{2}$$

$$\gamma, \delta \quad \left. \begin{array}{l} \gamma=1 \\ \delta=3 \end{array} \right\}$$

$$f_{MF} = a + bTM^2 + cM^4 + dTM$$

$$f_{MF} = \frac{1}{2} JM^2 - \frac{1}{\beta} \ln [\cosh(\beta[H + JM])] \quad \dots (1)$$

$$\frac{\partial f}{\partial M} = 0 \Rightarrow M = \tanh[\beta(JM + H)]$$

$$T_c = \frac{J/K_B}{}, H=0$$

$$t = \frac{T - \frac{J/K_B}{}}{\frac{J/K_B}{}} = \frac{K_B T - J}{J}$$

start from $M=0$

$$G(P) = \langle s(P) s(0) \rangle \quad \underline{H=0}$$

$$= \frac{\text{Tr} [s(P) s(0) \exp[-\beta H]]}{\text{Tr} [e^{-\beta H}]} \quad \}$$

$$\text{Tr} = \sum_{s(0)=\pm} \sum_{s(2)=\pm} \dots \sum_{s(N)=\pm}$$

$$-\beta H = \beta \sum_{P, P'} \frac{J(P-P')}{2} s(P) s(P') \quad \}$$

$$+ \beta H \sum_P s(P) \quad \cancel{+ P}$$

$$s'(P) = -s(P) \quad \cancel{+ P}$$

$$G = \sum_{S(1)=\pm} \sum_{S(M)=\pm} S(p) \frac{\exp[-\beta E_p]}{Z} + \sum_{S'(1)=\pm} \sum_{S'(M)=\pm} (-S'(p)) (-1) \frac{\exp[-\beta E'_{p'}]}{Z}$$

$$\langle S(p) S(0) \rangle = \frac{\sum_{(i \neq 0) \{ S(i) = \pm 1 \}} S(p) e^{-\beta E_p [S(i)]}}{Z}, \quad S(0) = +1$$

$$= M(p)$$

For the calculation of $G(p)$,
let $M(p)$ depend on p .

$$S(p) = M(p) + \delta S(p)$$

$$\delta S(p) = S(p) - M(p)$$

$$\begin{aligned} & \sum_{p, p'} \frac{\delta(p-p')}{2} S(p) S(p') \\ &= \sum_{p, p'} \frac{\delta(p-p')}{2} [M(p) + \delta S(p)] \\ &\quad [M(p') + \delta S(p')] \\ &\approx \sum_{p, p'} \frac{\delta(p-p')}{2} [M(p) M(p')] \\ &\quad + [\delta S(p) - M(p)] \\ &\quad M(p') \\ &\quad + M(p) [\delta S(p')] \\ &\quad - M(p')] \\ &\approx - \sum_{p, p'} \frac{\delta(p-p')}{2} M(p) M(p') \end{aligned}$$

$$P, P' \sim$$

$$+ \sum_{P, P'} \frac{\delta(P - P')}{2} \left\{ S(P) M(P') + S(P') M(P) \right\}$$

$$\delta(P - P') \equiv \delta(P' - P)$$

(2) $M(P) = \tanh \left[\beta \left(\sum_{P'} \delta(P - P') M(P') \right) \right]$

$$\sum_{ij} \delta_{ij} M_j = \sum_i \left(\sum_j \delta_{ij} M_j \right)$$

$$M_j = M \quad \forall j$$

$$= M \sum_i \sum_j \delta_{ij}$$

$\underbrace{\delta}_{ij}$

$$= M \delta N$$

$$\rightarrow M(P) = \tanh \left[\beta \sum_{P'} \delta(P - P') M(P') \right]$$

$$\text{if } M(P') = M \quad \forall P'$$

$$M = \tanh \left[\beta M \sum_{P'} \delta(P - P') \right]$$

$$= \tanh \left[\beta M \delta \right]$$

$$(M(P) \approx \beta \sum_{P'} \delta(P - P') M(P'))$$

+ correction terms

Fourier transform

... 0 ... n

$$e^{i\theta} = \cos \theta + i \sin \theta$$

$$(*) \quad M(p) = \frac{1}{N} \sum_{k} e^{-ikp} \tilde{M}(k)$$

$$\xrightarrow{p-p'=p} \tilde{J}(p-p') = \frac{1}{N} \sum_{k_1} e^{-ik_p(p-p')} \tilde{J}(k_p) \quad (***)$$

$$\sum_{p'} e^{i(k_1 - k_2)p'} = \delta_{k_1, k_2}$$

$\begin{cases} 1 & i=k \\ 0 & i \neq k \end{cases}$

$$\text{RHS} \div B \sum_{p'} \tilde{J}(p-p') M(p')$$

$$= B \sum_{k_1} \sum_{k} \frac{1}{N^2} \sum_{p'} e^{-ik_p(p-p')} \tilde{J}(k_p)$$

$$= B \sum_{k_1} \sum_{k} \frac{1}{N^2} \tilde{J}(k_p) \tilde{M}(k) \left[e^{-ik_p p'} \sum_{p'} e^{+i(k_1 - k)p'} \right]$$

$$= \frac{B}{N} \sum_{k} \tilde{J}(k) \tilde{M}(k) e^{-ikp}$$

$$\text{LHS} = M(p) = \frac{1}{N} \sum_{k} e^{-ikp} \tilde{M}(k)$$

$$\frac{1}{N} \sum_{k} e^{-ikp} \tilde{M}(k) = \frac{B}{N} \sum_{k} e^{-ikp} \tilde{J}(k) \tilde{M}(k)$$

$$\boxed{\tilde{M}(k) = \beta \tilde{e}^*(k) \tilde{M}(k)} + \text{const.} \quad (2)$$

$$G(p) \xrightarrow{\downarrow} M(p) \xrightarrow{\text{F.T.}} \tilde{M}(k) \quad \text{MFT}$$

$$G(k)$$

$$\frac{1}{N} \sum_p e^{+ikp}$$

Use $\tilde{e}(p-p') = \tilde{e}(p'-p)$ to connect $\tilde{e}(p)$ and $\tilde{e}(k)$.

$$\tilde{e}(k) = [\tilde{e}(+k)]^*$$

If \tilde{e} is real, $\tilde{e} = \text{real function of } k$

expand $\tilde{e}(k)$ for small k , about $k \approx 0$.

$$\lim_{p \rightarrow \infty} G(p) \underbrace{e^{ikp}}$$

$$\tilde{e}(k) \approx \tilde{e}(1 - R^2 k^2) + O(k^4)$$

$$\tilde{M}(k) = \beta \tilde{e}(1 - R^2 k^2) + \tilde{M}(k) + \text{const.}$$

$$\tilde{M}(k) \approx \text{const.}$$

$$\frac{1}{[1 - \beta \tilde{e}(1 - R^2 k^2)]}$$

$$e^{-R^2 k^2 / \zeta_c}$$

$$G(r) \equiv M(r) \sim \frac{1}{r} \text{ ft.w.}$$

$$\int dr \frac{e^{ikr}}{r} e^{-r/\xi_c} \cdot \left(ik - \frac{1}{\xi_c} \right) r \frac{1}{k^2 + \frac{1}{\xi_c^2}}$$

const.

$$\frac{1 - \frac{T_c}{T} (1 - R^2 k^2)}{k^2 + \frac{1}{\xi_c^2}}$$

≈ const. R^{-2}

$$\xi_c \approx R t^{-\frac{1}{2}}$$

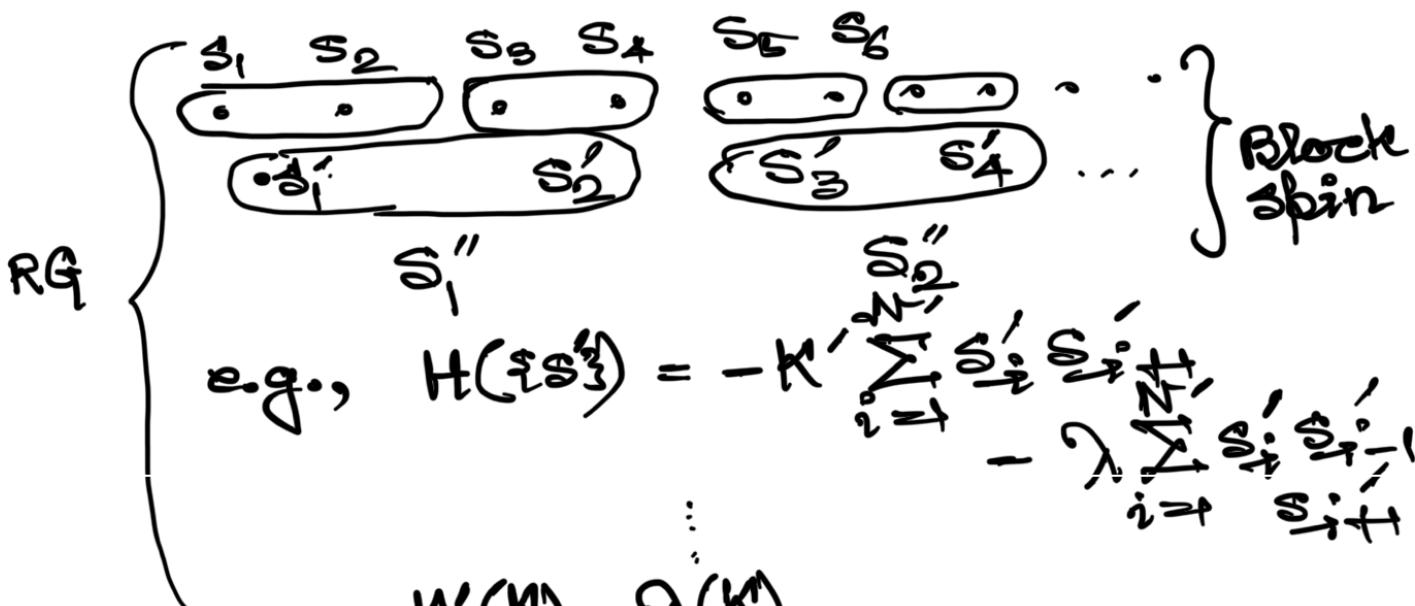
≈ $t^{\frac{1}{2}}$

$\beta = \frac{1}{2}$

Renormalization Group

$$H = -K \sum_{i=1}^N S_i S_{i+1}, \quad K = \beta J, \quad \beta = \frac{1}{k_B T}$$

Block Spin



$\kappa(n), \lambda(n)$

critical phenomenon \leftrightarrow long-distance physics

$H(\beta, \eta)$ का

$$S \mapsto S' \\ H(S) \xrightarrow{R.G.} H'(S')$$

$$\chi = \text{Tr } e^{-\beta H(S)}$$

Reduced Hamiltonian.



'Majority spin - method :

Projection operator :

$$T(S'_1; S_1, S_2, S_3) = 1 \quad \text{if}$$

$$S'_1 \left[\sum_{i=1}^3 S_i \right] > 0$$

= 0 otherwise

$$e^{-H'(S')} \equiv \text{Tr}_{\substack{S \\ \text{blocks}}} T(S'_1; S_1, S_2, S_3) e^{-H(S)}$$

$$\boxed{\sum_{S'} T(S'_1; S_1) = 1}$$

$$\sum_{S'} e^{-H'(S')}$$

$$-H(S)$$

$$\text{Tr} = \sum_{S} \sum_{S_i=\pm 1}^{S_i=1 \dots N}$$

$$\text{Tr}' = \sum_{S'_i=\pm 1}$$

$$Z = \prod_{i=1}^N e^{-\beta H(s_i)}$$

$$= \prod_{i=1}^N e^{-\beta H(s'_i)}$$

$$\bar{Z}(s'_1) = \bar{Z}(s_1)$$

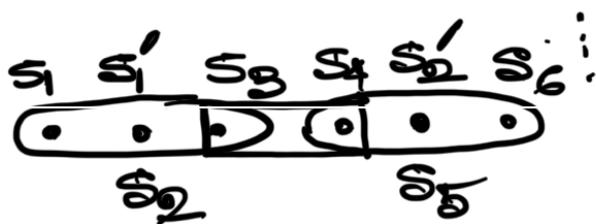
1-d Ising Model

$$H = -J \sum_{i,j} s_i s_j$$

Fluctuation Rule :-

$$T(s'_1, s_2, s_3, s_4) = \delta_{s'_1, s_2}$$

$$T(s'_1, s_2, s_3, s'_4) = \delta_{s'_1, s'_4}$$



$$\ln Z, \sum_{s_3, s_4} [e^{K s'_1 s_3} e^{K s_3 s_4} e^{K s_4 s'_2}] \dots$$

$$s_3 = \pm 1$$

$$s_4 = \pm 1$$

$$e^{K s_3 s_4} = \cosh K + \sinh K \quad s_3 s_4$$

$$x = \tanh K = \cosh K (1 + (\tanh K) s_3 s_4)$$

$$\sum_{s_3 = \pm 1} \sum_{s_4 = \pm 1} (\cosh K)^3 \frac{(1 + x s'_1 s_3)}{(1 + x s_3 s_4)}$$

$$\begin{aligned}
 &= \sum_{\substack{s_3 = \pm 1 \\ s_4 = \pm 1}} (\cosh K)^3 \left[1 + \bar{x} s_3 s_4 + \bar{x} s_1' s_3 \right. \\
 &\quad \left. + \bar{x}^2 s_1' s_4 \right] \frac{(1 + \bar{x} s_1 s_2')}{(1 + \bar{x} s_1 s_2')} \\
 &= (\cosh K)^3 \sum_{\substack{s_3 = \pm 1 \\ s_4 = \pm 1}} \left\{ 1 + \bar{x} s_4 s_2' \right. \\
 &\quad \left. + \cancel{\bar{x} s_3 s_4 + \bar{x}^2 s_3 s_2'} \right. \\
 &\quad \left. + \cancel{\bar{x} s_1' s_3 + \bar{x}^2 s_1' s_2'} \right. \\
 &\quad \left. + \cancel{\bar{x}^2 s_1' s_4 + \bar{x}^3 s_1' s_2'} \right\} \\
 &= \dots 2^2 (\cosh K)^3 (1 + \bar{x}^3 s_1' s_2') \dots \\
 e^{K' s_1' s_2'} &= [(\cosh K') (1 + (\tanh K') s_1' s_2')]
 \end{aligned}$$

$$s_1' = \dots + (-K') s_1' s_2' + \dots$$

$$(\cosh K)^3 = \cosh K'$$

$$\tanh K' = [\tanh K]^3$$

$$K' = \tanh^{-1} [\tanh^3(K)]$$

$$\bar{x}' = \bar{x}^3$$

$$\bar{x}' = \tanh K'$$

$$\bar{x} = \tanh K$$

$$\bar{x}_1 = \bar{x}_0^3$$

$$\bar{x}_2 = \bar{x}_1^3 = \bar{x}_0^9$$

$$\bar{x}_3 = \bar{x}_2^3 = (\bar{x}_0^9)^3$$

$$\frac{N/3}{\dots} \dots = \bar{x}_0^{27}$$

$$H = \sum_{i=1}^N -K \vec{s}_i \cdot \vec{s}_{i+1} + \text{constant}$$

$$Z = T_p [e^{-\beta E(\vec{s})}]$$

$$\underline{\underline{H.W.}} \div H'(\vec{s}) = N' g(K') - \sum_{i=1}^{N'} K' \vec{s}_i \cdot \vec{s}_{i+1}$$

$$g(K') = \frac{-1}{3} \ln \left[\frac{\cosh^3 K'}{\cosh K'} \right]$$

$$x = \tanh(K) - \frac{2}{3} \ln 2$$

$$x' = x^3$$

$$\text{high } T ; \quad K = \frac{e}{k_B T}$$

$$x = \tanh \left(\frac{e}{k_B T} \right)$$

$$x \rightarrow 0^+$$

$$\text{low } T ; \quad x \rightarrow 1^-$$

$$\text{zero Temperature} \div T = 0$$

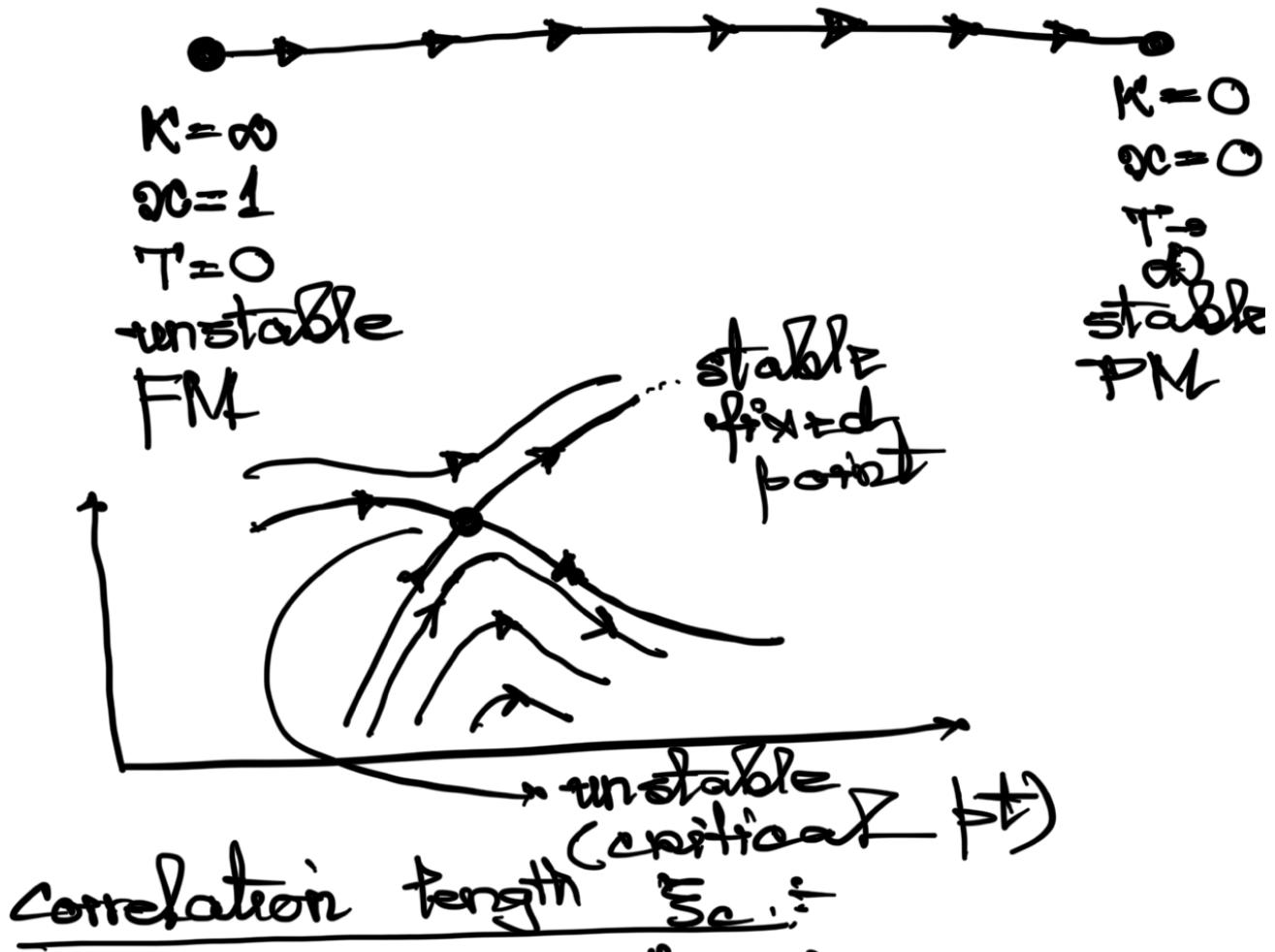
$x = 1$
 fixed point for $K \div T = 0$
 considered state'

$$0 < x < 1 \text{ if } x_{\text{initial}} < 1$$

$$x_{\text{final}} = 0$$

for any $K < \infty$, eventually $K \rightarrow 0$
 Only for $T = 0$, ferromagnet
 $T > 0$, paramagnet

$\boxed{xc' = xc^3}$ $\xrightarrow{\sim}$ stable: paramag
 unstable: FM



[ξ]

$$a' = ba$$

$$\boxed{b = 3}$$

$$\xi^\# = \xi/a$$

(dimensionless)

$$\xi(s') = \xi(s)$$

$$\xi^\#(s') a' = \xi^\#(s) a$$

or

$$\xi^\#(s') = \frac{1}{b} \xi^\#(s)$$

$$\boxed{xc' = xc^b}$$

$$\boxed{\xi^\#(xc') = \frac{1}{b} \xi^\#(xc)}$$

$$\xi^*(\varphi) = \frac{\text{const.}}{\partial \ln \varphi}$$

$$\begin{aligned}\xi^*(\varphi') &= \frac{\text{const.}}{\partial \ln \varphi'} \\ &= \frac{1}{b} \cdot \frac{\text{const.}}{\partial \ln \varphi} \\ &= \frac{1}{b} \xi^*(\varphi)\end{aligned}$$

$$\bar{\xi}^* = \frac{\text{const.}}{\partial \ln [\tanh(K)]}$$

FM critical point
 $\xi^* \rightarrow \infty$
 $K \rightarrow 0$
 $\varphi = +1$

Monte Carlo

Each spin $\pm \frac{1}{2}$

$k^N \div e^N \partial \ln K$

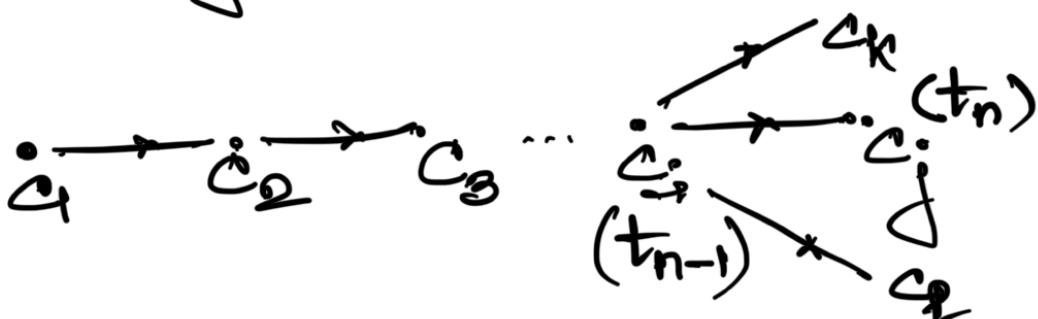
cannot sample efficiently all config.

$$P(c) = e^{-\beta E_c} / Z$$

- Do not sample configurations at high E_c / low T / high β .

$$\begin{aligned}
 M = \langle s_i \rangle &= \frac{1}{N} \sum_i \langle s_i \rangle \\
 &= \frac{1}{N} \sum_i \left[\sum_c e^{-\beta E_c} s_i / Z \right] \\
 &= \frac{1}{N} \sum_i \frac{e^{-\beta E_c}}{Z} \left[\sum_c \frac{\langle s_i \rangle_c}{N} \right] \\
 &= \cancel{\langle s_i \rangle} \quad \langle \theta \rangle = \sum_c p_c \theta_c \\
 &= \sum_c \left(\frac{e^{-\beta E_c}}{Z} \right) \theta_c \\
 \frac{N_c}{N_{\text{tot}}} &= e^{-\beta E_c} / Z \quad (\text{Monte Carlo})
 \end{aligned}$$

- Importance Sampling
- specific kind of movement through $\{C_i\}$



$$W_{i \rightarrow j} = W(C_i \rightarrow C_j)$$

$$= P(X_{t_n} = C_j \wedge X_{t_{n-1}} = C_i)$$

$$P(X_{t_n} = C_j) = P(C_j, t)$$

$$\frac{dP(C_j, t)}{dt} = - \sum_i W_{j \rightarrow i} P(C_j, t)$$

$$+ \sum_i W_{i \rightarrow j} P(C_i, t)$$

: Equilibrium $\frac{dP(C_j, t)}{dt} = 0$

$$P_{eq}(C_j) = e^{-\beta E_{C_j}} / Z$$

$$\frac{dP_{eq}}{dt} = 0 \Rightarrow - \sum_i W_{j \rightarrow i} P_{eq}(C_j)$$

$$+ \sum_i W_{i \rightarrow j} P_{eq}(C_i) = 0$$

Detailed Balance \therefore (*)
 (sufficient, not necessary)

$$W_{i \rightarrow j} P_{eq}(C_i) = W_{j \rightarrow i} P_{eq}(C_j)$$

$$\frac{W_{i \rightarrow j}}{W_{j \rightarrow i}} = \frac{P_{eq}(C_j)}{P_{eq}(C_i)}$$

$$= e^{-\beta E_{C_j}} / e^{-\beta E_{C_i}}$$

$$\Delta E = E_{C_j} - E_{C_i} = (E_{n-} - E_{n+}) \dots$$

$$\Delta = \exp(-\frac{E_{Cj} - E_{Ci}}{kT}) \quad \dots (1)$$

$$W_{i \rightarrow j} = T_0^{-1} \exp(-\beta(E_{Cj} - E_{Ci})) \text{ or}$$

$$= T_0^{-1} \quad \text{if } E_{Cj} < E_{Ci} \\ \text{or } \Delta E > 0$$

Metropolis algorithm:

$$W_{i \rightarrow j} = 1 \quad \text{if } E_{Cj} > E_{Ci}$$

$$= \exp[-\beta(E_{Cj} - E_{Ci})] \quad \text{if } E_{Cj} > E_{Ci}$$

$$\boxed{T_0 = T_0^{-1} = 1}$$

Metropolis algorithm :

- Start with some configuration I_0 .
- Choose some site i
- If O_i is flipped, then calculate $\Delta E = E_F - E_I$
- If $E_F < E_I$, go to C_{new}
- If $E_F > E_I$, go to C_{new} with $P = \exp[-\beta \Delta E]$

[• Generate a random r
 $0 < r < 1, r \sim U(0, 1)$]

- If $r < \exp[-\beta \Delta E]$, go to

• Move
 ↗ new
 over, stay on C_{st}
 do next site.

exponential in $N \rightarrow$ slower in N
 exact \rightarrow convergent
 N_{tot} but statistical

$$M_f = \sum_{c=1}^{N_3} \langle \sigma_i^z \rangle, c / N_{\text{tot}} = \langle \sigma_i^z \rangle$$

$$\begin{aligned} M &= \sum_{i=1}^{N_3} \langle \sigma_i^z \rangle / N_3 \\ &= \frac{1}{N_{\text{tot}}} \sum_{c=1}^{N_{\text{tot}}} \left[\frac{1}{N_3} \sum_{i=1}^{N_3} \langle \sigma_i^z, c \rangle \right] \end{aligned}$$

H.W \div simulate by Metropolis algorithm,
 a $d=2$ Ising Model
 at temperature T .

$$C_V = \frac{\partial \langle E \rangle}{\partial T} \propto \langle E^2 \rangle - \langle E \rangle^2$$

$$\chi = \frac{\partial M}{\partial T} \propto \langle \sigma^2 \rangle - \frac{\langle \sigma \rangle^2}{N^2}$$

$$H_{2-d} = -J \sum_{\langle ij \rangle} \sigma_i \sigma_j - B \sum_i \sigma_i$$

\Rightarrow Input \div dimensions of lattice
 N_x, N_y
 $T(\beta), J, B$

$$\begin{aligned} k_B &= 1 \\ T &\text{ (energy)} \end{aligned}$$

$$N_S = N_{SC} N_y$$

$\downarrow \downarrow \downarrow$
 K, H

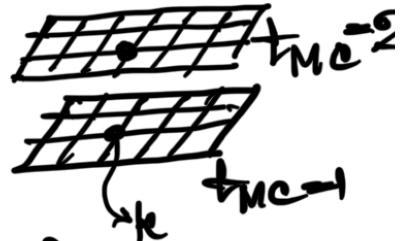
$\downarrow \tau, \wedge$
 $B/\tau, H$

N_{warms}
(warm-up sweep #)
• No not measure. N_{warms}

$$\frac{1}{N_S} \sum_{k=1}^{N_S} \frac{1}{N_{\text{warms}}} \sum_{\substack{\text{warm} \\ -\text{MC}}}^k S(t_{MC}) S(t_{MC} + \tau)$$

$\sim f(\tau)$

$$\frac{1}{N_S} \sum_{k=1}^{N_S} \frac{1}{N_{\text{warms}}} \left[S(1) S^k(4) + S(2) S^k(5) + \dots + S(997) S^k(1000) + S^k(998) S^k(1) + \dots + S^k(1000) \right] \sim F(k)$$



$\boxed{N_{\text{sweeps}}} \gg N_{\text{warms}}$.

(M)

$(S_i)_c$,

$$S_C = \frac{1}{N_S} \sum_{i=1}^{N_S} (S_i)_C$$

$i = 1, \dots, N_S$

$(S_i)_C \rightarrow \text{Store it!}$

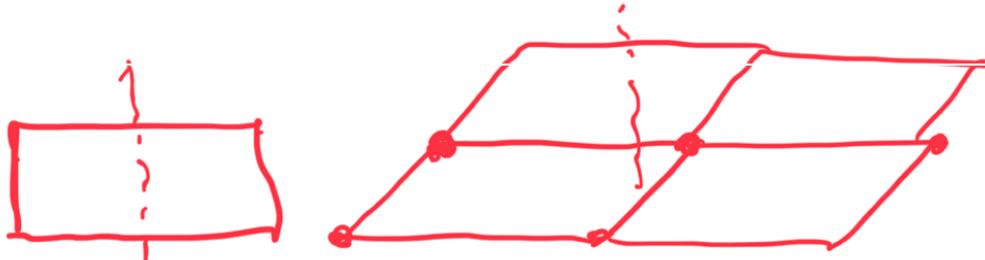
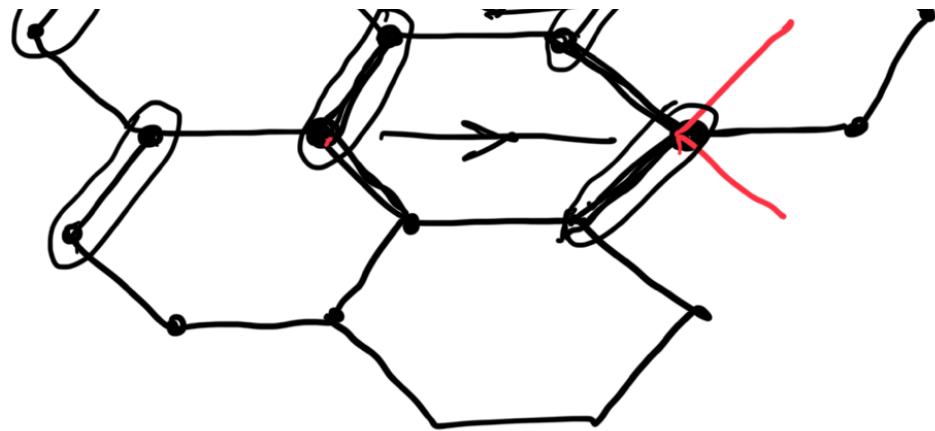
$$\frac{1}{N_{\text{sweeps}}} \sum_{C=1}^{N_{\text{sweeps}}} S_C = \langle M \rangle$$

Reheating



Bravais Lattices





$$-\left(\frac{F - \langle E \rangle}{T}\right) = \beta$$

$$\text{or } \langle S^2 \rangle - \langle S \rangle^2$$

Cr,

$$\begin{matrix} T \rightarrow T_c \leftarrow T \\ H \rightarrow O^+, O^- \end{matrix}$$

$\alpha, \beta, \gamma, \delta, \nu, \eta$

$$J_{ij} \sim U\left[-\frac{J}{2}, \frac{J}{2}\right]$$