

# Quantum Monte Carlo for spin systems

YAN, Zheng 严正 嚴正

Nov.4, 2019

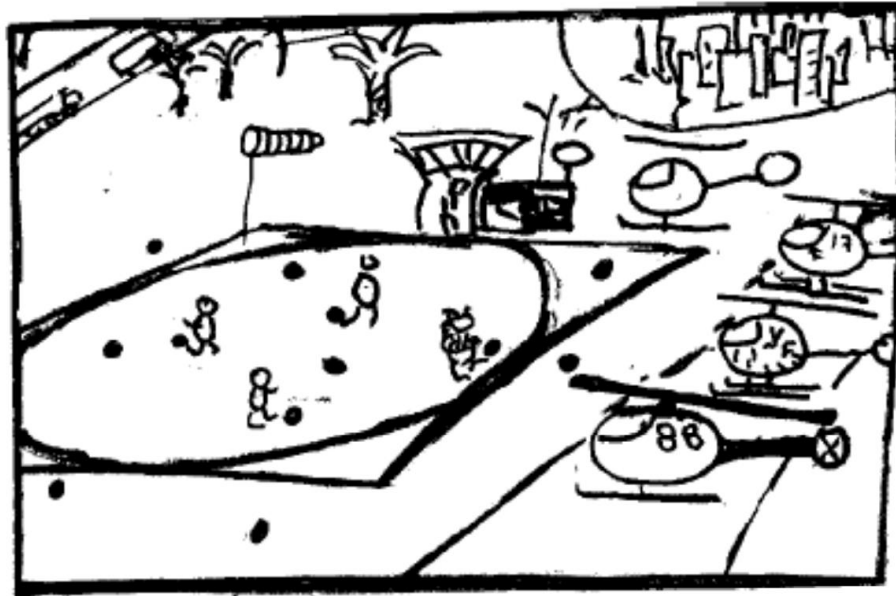
HKU-UCAS Study Group

# Sampling method

- Direct sampling

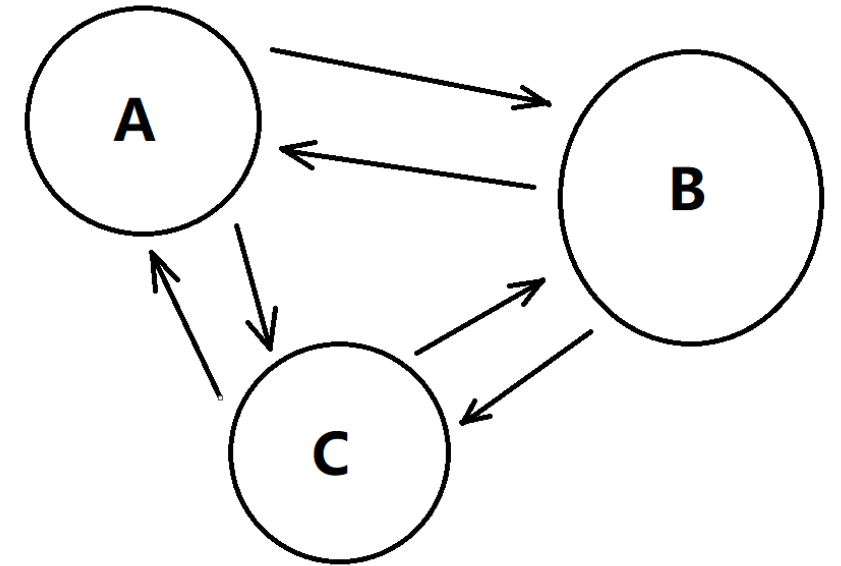
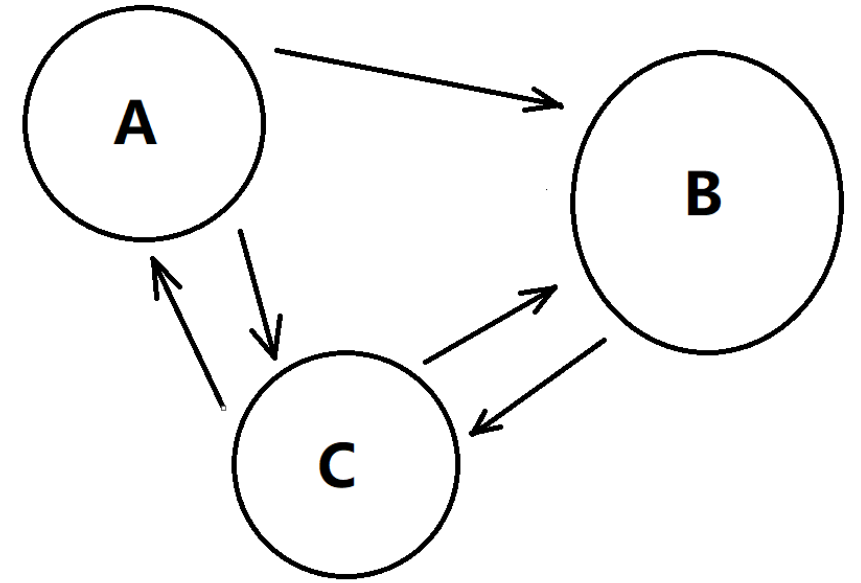


- Markov chain sampling



# Balance and probability

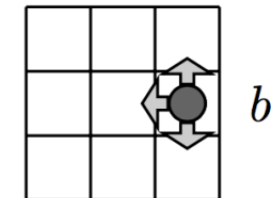
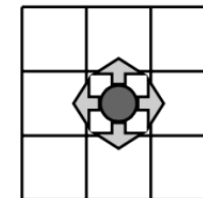
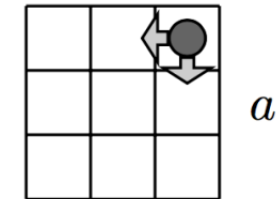
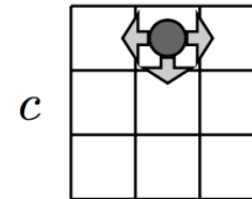
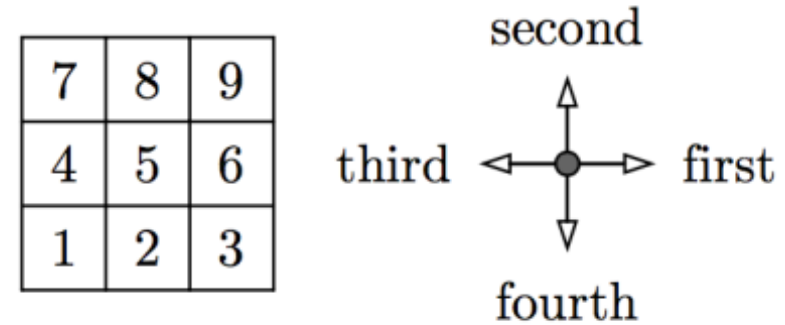
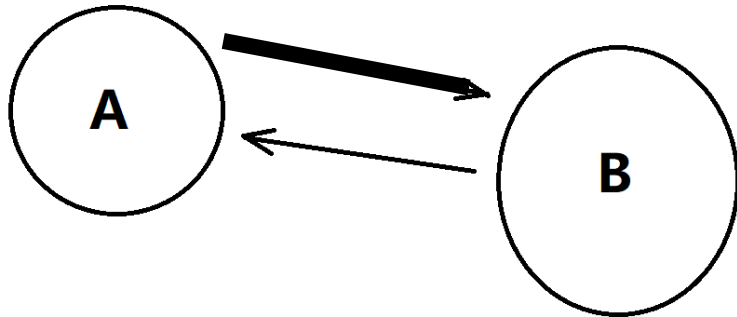
- Balance:  $P(A_{\text{out}}) = P(A_{\text{in}})$
- Detail Balance:  $P(A \text{ to } B) = P(B \text{ to } A)$
- Ergodicity!



# Metropolis Probability

$$P(A \text{ to } B) = \min\left[1, \frac{W_B}{W_A}\right]$$

$$P(A \text{ to } B) = \min\left[1, \frac{W_B P_{\text{select}}(B \text{ to } A)}{W_A P_{\text{select}}(A \text{ to } B)}\right]$$



# Sampling in statistic mechanics: sampling for partition function

$$Z = \sum_{\mathcal{C}} e^{-\beta H[\mathcal{C}]} = \sum_{\mathcal{C}} W(\mathcal{C})$$

Sign problem?!

$$1 = \frac{Z}{Z} = \frac{\sum_{\mathcal{C}} W_{\mathcal{C}}}{Z} = \sum_{\mathcal{C}} P_{\mathcal{C}}$$

$$\langle A \rangle = \frac{\sum_{\mathcal{C}} A_{\mathcal{C}} P_{\mathcal{C}}}{\sum P_{\mathcal{C}}},$$

## ➤ The Ising model: Introduction

Hamiltonian:  $H = -J \sum_{\langle i,j \rangle} \sigma_i \sigma_j$

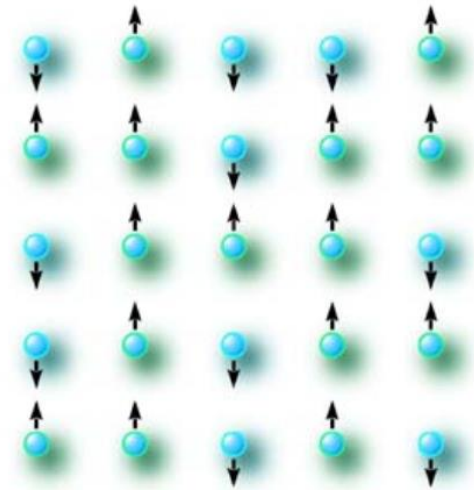
Spin:  $\sigma = \pm 1$

$\langle i,j \rangle$ : Sum over nearest neighbor pairs.

$|J|=1$ : Coupling strength

$J>0$ —ferromagnetic

$J<0$ —antiferromagnetic



# Quantum spin systems

## Stochastic Series Expansion (SSE) picture

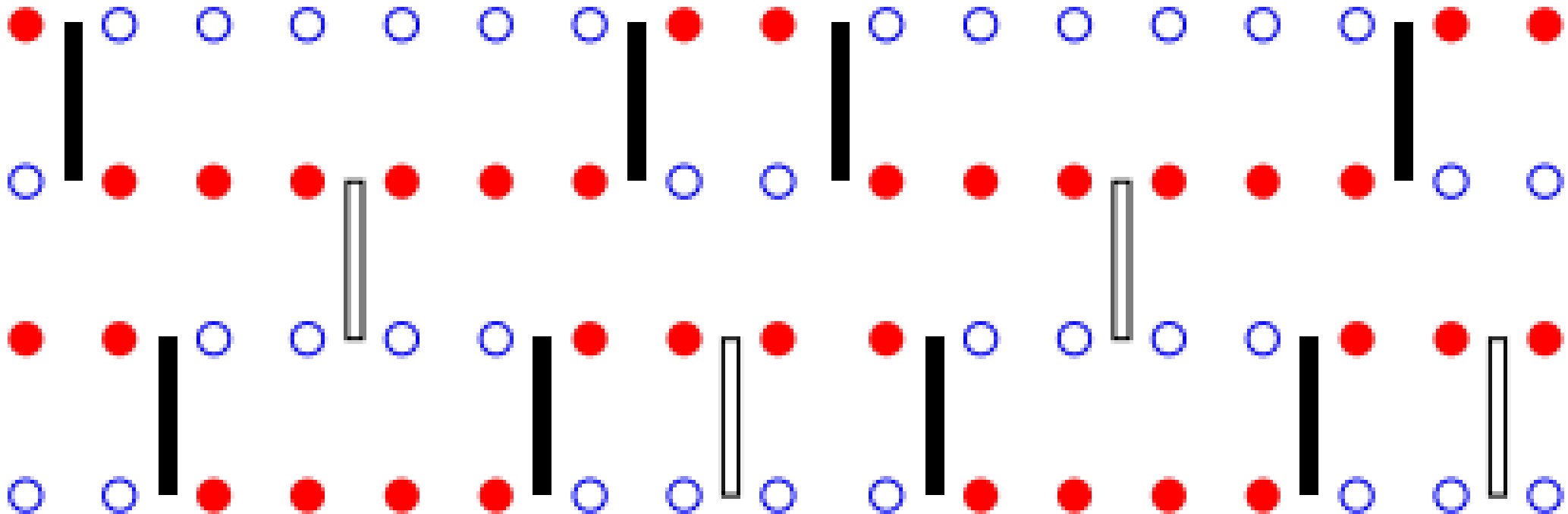
$$Z = \text{Tr}\{e^{-\beta H}\} \quad H = - \sum_{a,b} H_{a,b} \quad \left\{ \begin{array}{l} H = \sum_{b=1}^B \left[ S_{i(b)}^z S_{j(b)}^z + \frac{1}{2} (S_{i(b)}^+ S_{j(b)}^- + S_{i(b)}^- S_{j(b)}^+) \right] \\ H_{1,b} = \frac{1}{4} - S_{i(b)}^z S_{j(b)}^z, \quad H_{2,b} = \frac{1}{2} (S_{i(b)}^+ S_{j(b)}^- + S_{i(b)}^- S_{j(b)}^+) \end{array} \right.$$

$$Z = \sum_{\alpha} \sum_{n=0}^{\infty} \frac{\beta^n}{n!} \langle \alpha | (-H)^n | \alpha \rangle \quad (-H)^n = \sum_{\{H_{ab}\}} \frac{(M-n)!n!}{M!} \prod_{p=1}^M H_{a(p),b(p)}$$

$$Z = \sum_{\alpha} \sum_{\{H_{ab}\}} \frac{\beta^n (M-n)!}{M!} \left\langle \alpha \left| \prod_{i=1}^M H_{a(i),b(i)} \right| \alpha \right\rangle$$

# Stochastic Series Expansion (SSE) picture

$$Z_{\text{SSE}} = \sum_{\{\alpha\}} \sum_{S_M} \frac{\beta^n (M-n)!}{M!} \langle \alpha_0 | H_{b_M} | \alpha_{M-1} \rangle \cdots \langle \alpha_2 | H_{b_2} | \alpha_1 \rangle \langle \alpha_1 | H_{b_1} | \alpha_0 \rangle,$$



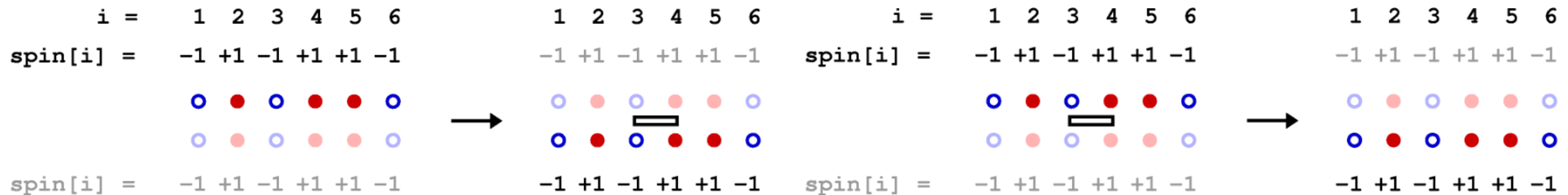
# Diagonal update (e.g. Heisenberg model)

$$H_{1,b} = \frac{1}{4} - S_{i(b)}^z S_{j(b)}^z, \quad H_{2,b} = \frac{1}{2} \left( S_{i(b)}^+ S_{j(b)}^- + S_{i(b)}^- S_{j(b)}^+ \right) \quad Z = \sum_{\alpha} \sum_{\{H_{ab}\}} \frac{\beta^n (M-n)!}{M!} \left\langle \alpha \left| \prod_{i=1}^M H_{a(i),b(i)} \right| \alpha \right\rangle$$

$$P_{\text{accept}}(A \rightarrow B) = \min \left( \frac{W(B)P_{\text{select}}(B \rightarrow A)}{W(A)P_{\text{select}}(A \rightarrow B)}, 1 \right)$$

$$P_{\text{accept}}(n \rightarrow n+1) = \min \left( \frac{B\beta/2}{M-n}, 1 \right)$$

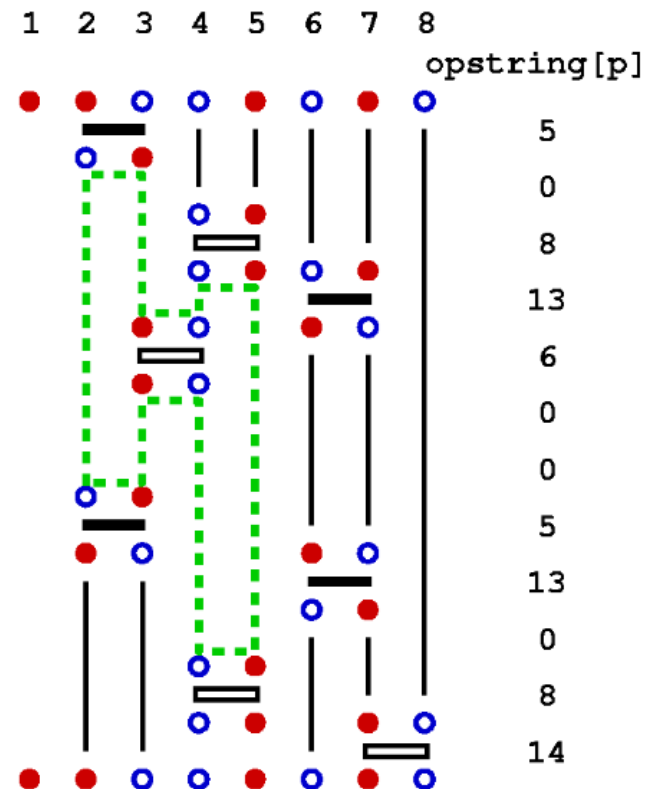
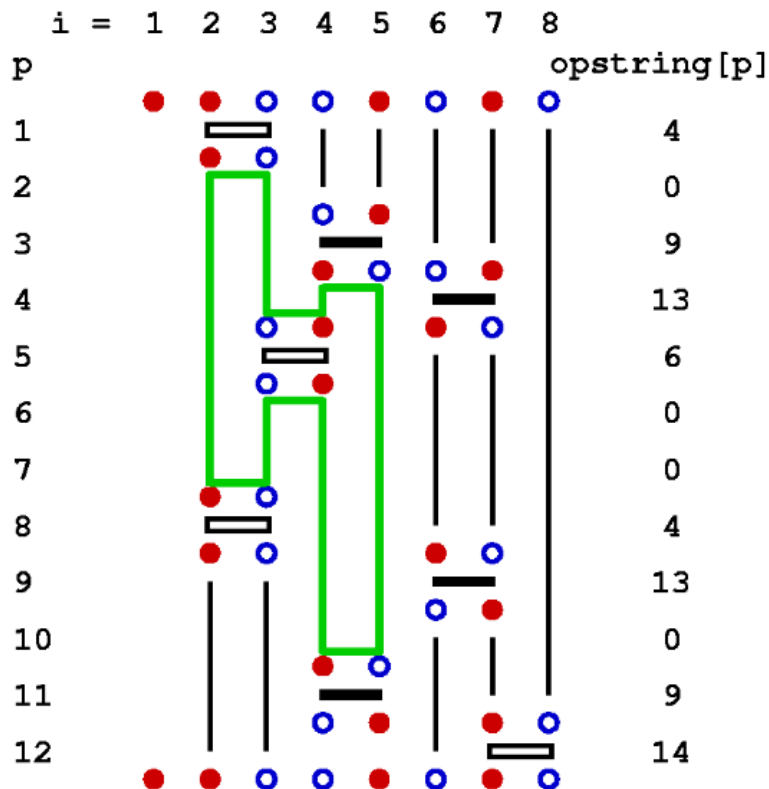
$$P_{\text{accept}}(n \rightarrow n-1) = \min \left( \frac{M-n+1}{B\beta/2}, 1 \right)$$





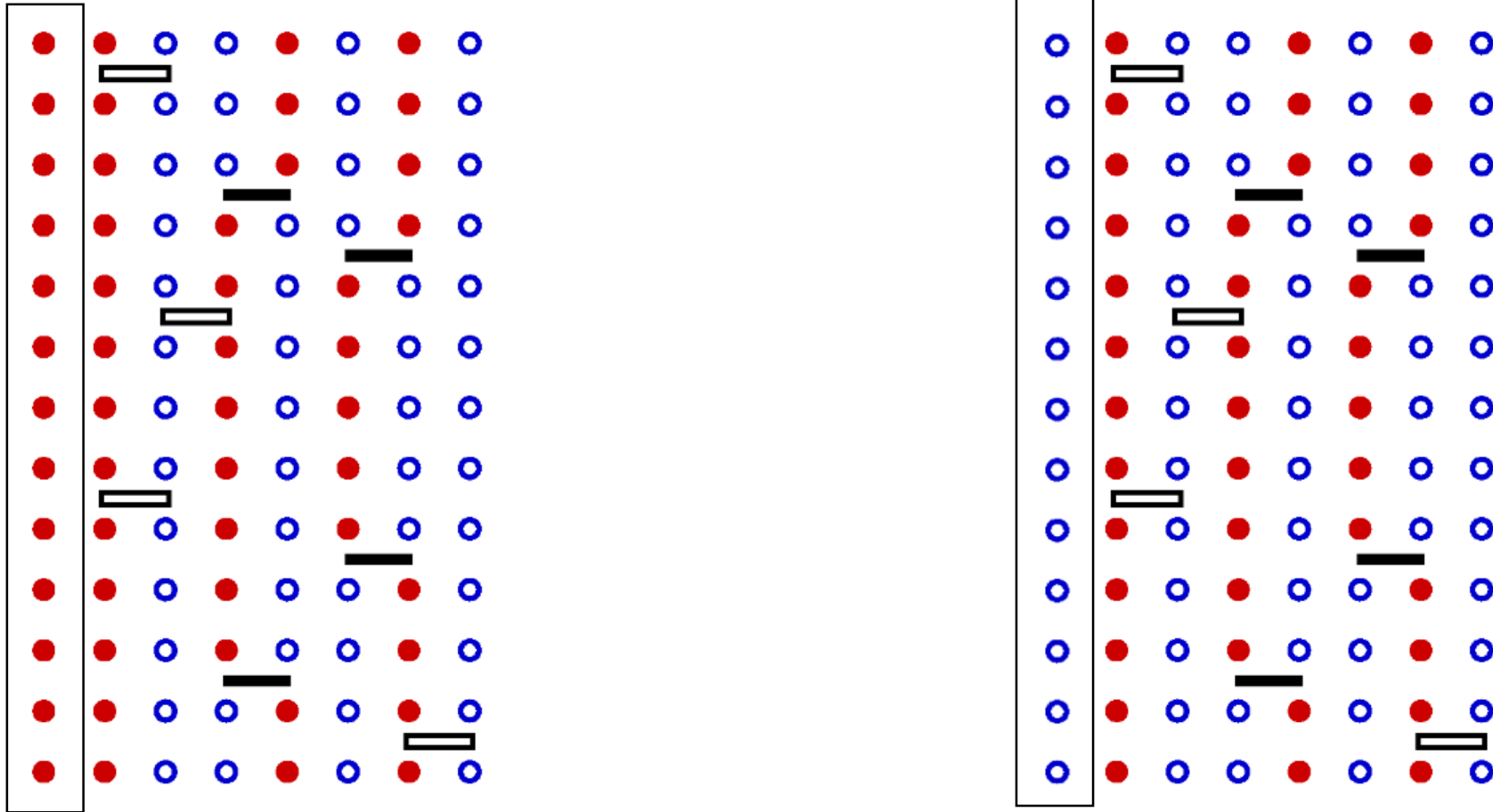
# Off-diagonal update (e.g. Heisenberg model)

Another choice:  $P=W_2/(W_1+W_2)$



# Free spin update to change total $S^z$

$$Z = \sum_{\{\alpha_i\}} \sum_{n=0}^{\infty} \frac{\beta^n}{n!} \langle \alpha_0 | -H | \alpha_1 \rangle \langle \alpha_1 | -H | \alpha_2 \rangle \cdots \langle \alpha_n | -H | \alpha_n \rangle$$



Code: <http://physics.bu.edu/~sandvik/programs/ssebasic/ssebasic.f90>

# Measurement

- Energy:

$$\begin{aligned}
 Z &= \sum_{n=0}^{\infty} \frac{(-\beta)^n}{n!} \sum_{\{\alpha\}_n} \langle \alpha_0 | H | \alpha_{n-1} \rangle \cdots \langle \alpha_2 | H | \alpha_1 \rangle \langle \alpha_1 | H | \alpha_0 \rangle, \\
 E &= \frac{1}{Z} \sum_{n=0}^{\infty} \frac{(-\beta)^n}{n!} \sum_{\{\alpha\}_{n+1}} \langle \alpha_0 | H | \alpha_n \rangle \cdots \langle \alpha_2 | H | \alpha_1 \rangle \langle \alpha_1 | H | \alpha_0 \rangle, \\
 E &= -\frac{1}{Z} \sum_{n=1}^{\infty} \frac{(-\beta)^n}{n!} \frac{n}{\beta} \sum_{\{\alpha\}_n} \langle \alpha_0 | H | \alpha_n \rangle \cdots \langle \alpha_2 | H | \alpha_1 \rangle \langle \alpha_1 | H | \alpha_0 \rangle.
 \end{aligned}
 \quad E = -\frac{\langle n \rangle}{\beta}.$$

- Specific heat: since  $C = (\langle E^2 \rangle - \langle E \rangle^2) / T^2$  we have  $C = \langle n^2 \rangle - \langle n \rangle^2 - \langle n \rangle$ .

*Diagonal operators.* Expectation values of operators diagonal in the z-component basis are also easy to evaluate, using averages over the propagated states;

$$\langle O_z \rangle = \frac{1}{L} \sum_{p=0}^{L-1} \langle \alpha(p) | O_z | \alpha(p) \rangle = \frac{1}{L} \sum_{p=0}^{L-1} O_z(p).$$

# Imaginary time: from path integral to SSE

$$e^{-\tau \mathcal{H}_0} U(\tau) = e^{-\tau(\mathcal{H}_0 + V)}. \quad V(\tau) = e^{\tau \mathcal{H}_0} V e^{-\tau \mathcal{H}_0}$$

$$\frac{dU(\tau)}{d\tau} = -V(\tau)U(\tau) \quad U(\tau) = 1 - \int_0^\tau d\tau' V(\tau')U(\tau').$$

$$\begin{aligned} Z &= \text{Tr}[e^{-\beta \mathcal{H}_0} U(\beta)] \\ &= \text{Tr}\left[e^{-\beta \mathcal{H}_0} \left(1 - \int_0^\beta d\tau_1 V(\tau_1) + \int_0^\beta d\tau_2 \int_0^{\tau_2} d\tau_1 V(\tau_2)V(\tau_1) - \dots\right)\right] \\ &= \text{Tr}\left[\sum_{n=0}^{\infty} (-1)^n \int_0^\beta d\tau_n \int_0^{\tau_n} d\tau_{n-1} \dots \int_0^{\tau_2} d\tau_1 e^{-(\beta-\tau_n)\mathcal{H}_0} V e^{-(\tau_n-\tau_{n-1})\mathcal{H}_0} \right. \\ &\quad \left. \times V \dots V e^{-\tau_1 \mathcal{H}_0}\right] \\ &= \sum_{n=0}^{\infty} (-1)^n \sum_{\phi_n, \dots, \phi_1} \int_0^\beta d\tau_n \int_0^{\tau_n} d\tau_{n-1} \dots \int_0^{\tau_2} d\tau_1 \langle \phi_1 | e^{-(\beta-\tau_n)\mathcal{H}_0} | \phi_1 \rangle \langle \phi_1 | V | \phi_n \rangle \\ &\quad \times \langle \phi_n | e^{-(\tau_n-\tau_{n-1})\mathcal{H}_0} | \phi_n \rangle \langle \phi_n | V | \phi_{n-1} \rangle \dots \langle \phi_2 | V | \phi_1 \rangle \langle \phi_1 | e^{-\tau_1 \mathcal{H}_0} | \phi_1 \rangle. \end{aligned}$$

$$w(c) d\tau_n \dots d\tau_1$$

$$= \left(\frac{\Gamma}{2}\right)^n e^{-(\beta-\tau_n)E_0(\phi_1)} e^{-(\tau_n-\tau_{n-1})E_0(\phi_n)} \dots e^{-\tau_1 E_0(\phi_1)} d\tau_n \dots d\tau_1,$$

Another useful continuous time path integral representation can be obtained by splitting the Hamiltonian into  $\mathcal{H}_0 = 0$  and  $V = \mathcal{H}$ . The weight of configuration  $c$  is then written as

$$w(c) d\tau_m \dots d\tau_1$$

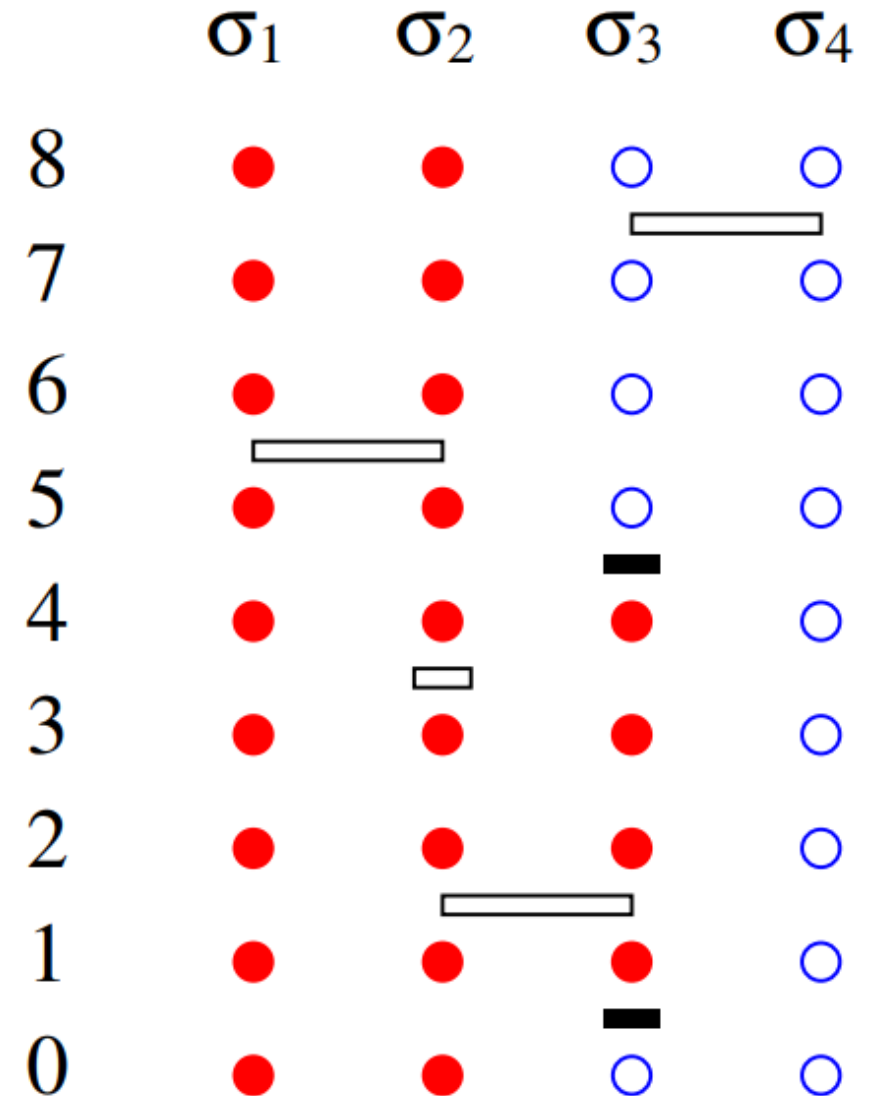
$$= (-1)^m \langle \phi_1 | \mathcal{H} | \phi_m \rangle \langle \phi_m | \mathcal{H} | \phi_{m-1} \rangle \dots \langle \phi_2 | \mathcal{H} | \phi_1 \rangle d\tau_m \dots d\tau_1.$$



$$Z = \sum_{m=0}^{\infty} \frac{(-\beta)^m}{m!} \sum_{\phi_m, \dots, \phi_1} \langle \phi_1 | \mathcal{H} | \phi_m \rangle \langle \phi_m | \mathcal{H} | \phi_{m-1} \rangle \dots \langle \phi_2 | \mathcal{H} | \phi_1 \rangle.$$

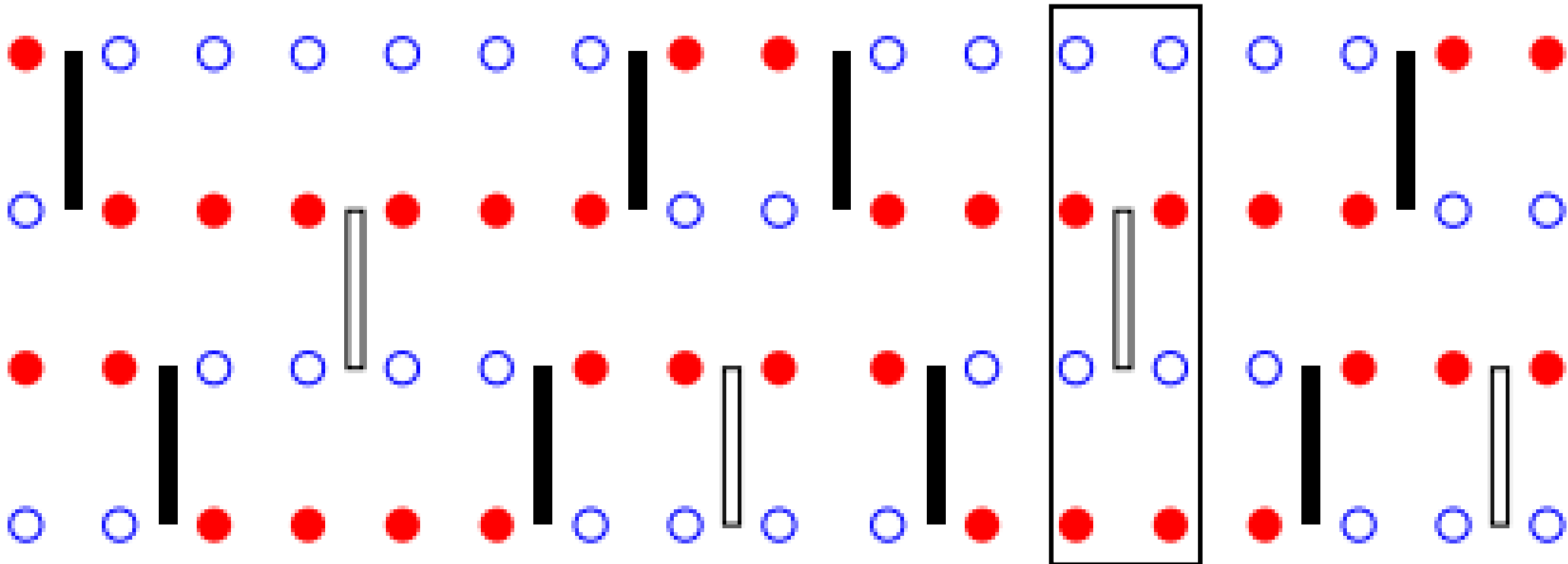
Sign the imaginary time  
on every piece!

Note: get random numbers  
from 0 to beta with  
equal probability!



# Stochastic Series Expansion (SSE) picture

$$Z_{\text{SSE}} = \sum_{\{\alpha\}} \sum_{S_M} \frac{\beta^n (M-n)!}{M!} \langle \alpha_0 | H_{b_M} | \alpha_{M-1} \rangle \cdots \boxed{\langle \alpha_2 | H_{b_2} | \alpha_1 \rangle} \langle \alpha_1 | H_{b_1} | \alpha_0 \rangle,$$



# Diagonal update for TFIM

$$H = -J \sum_{\langle i,j \rangle} \sigma_i^z \sigma_j^z - h \sum_i \sigma_i^x,$$

$$\begin{aligned} H_{0,0} &= I, \\ H_{-1,a} &= h(\sigma_a^+ + \sigma_a^-), \\ H_{0,a} &= h, \\ H_{1,a} &= J(\sigma_i^z \sigma_j^z + 1). \end{aligned}$$

$$\begin{aligned} \langle \bullet \mid H_{-1,a} \mid \circ \rangle &= \langle \circ \mid H_{-1,a} \mid \bullet \rangle = h, \\ \langle \bullet \mid H_{0,a} \mid \bullet \rangle &= \langle \circ \mid H_{0,a} \mid \circ \rangle = h. \end{aligned}$$

$$\langle \bullet \bullet \mid H_{1,a} \mid \bullet \bullet \rangle = \langle \circ \circ \mid H_{1,a} \mid \circ \circ \rangle = 2J.$$

Choose a kind of operators to be inset.

Probability of inset.

$$P(h) = \frac{hN}{hN + (2J)N_b},$$

$$P = \min \left( \frac{\beta(hN + (2J)N_b)}{M - n}, 1 \right),$$

$$P(J) = \frac{(2J)N_b}{hN + (2J)N_b}.$$

## Another choice?

Probability of remove

$$P = \min \left( \frac{M - n + 1}{\beta[hN + (2J)N_b]}, 1 \right),$$

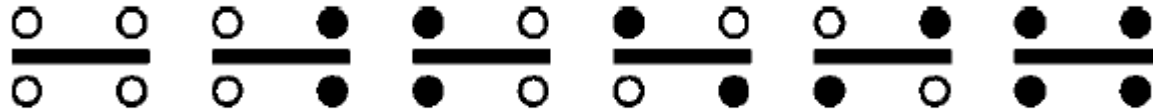
What's the probability here? —0.5





# Directed loop update

$$H = J \sum_{\langle i,j \rangle} [S_i^x S_j^x + S_i^y S_j^y + \Delta S_i^z S_j^z] - h \sum_i S_i^z \quad \Longrightarrow \quad \begin{aligned} H_{1,b} &= C - \Delta S_{i(b)}^z S_{j(b)}^z + h_b [S_{i(b)}^z + S_{j(b)}^z] \\ H_{2,b} &= \frac{1}{2} [S_{i(b)}^+ S_{j(b)}^- + S_{i(b)}^- S_{j(b)}^+] \end{aligned}$$

Only 6 configurations: 

$$\begin{aligned} \langle \downarrow \downarrow | H_b | \downarrow \downarrow \rangle &= \epsilon, \quad C = C_0 + \epsilon, \quad C_0 = \Delta/4 + h_b, \\ \langle \downarrow \uparrow | H_b | \downarrow \uparrow \rangle &= \langle \uparrow \downarrow | H_b | \uparrow \downarrow \rangle = \Delta/2 + h_b + \epsilon, \\ \langle \uparrow \downarrow | H_b | \downarrow \uparrow \rangle &= \langle \downarrow \uparrow | H_b | \uparrow \downarrow \rangle = 1/2, \\ \langle \uparrow \uparrow | H_b | \uparrow \uparrow \rangle &= \epsilon + 2h_b. \end{aligned}$$

Probability:

$$\begin{aligned} P([0,0]_p \rightarrow [1,b]_p) &= \frac{N_b \beta \langle \alpha(p) | H_{1,b} | \alpha(p) \rangle}{M - n}, \\ P([1,b]_p \rightarrow [0,0]_p) &= \frac{M - n + 1}{N_b \beta \langle \alpha(p) | H_{1,b} | \alpha(p) \rangle}. \end{aligned}$$

Question: if flip-symmetry doesn't work, how to do it??

# Directed loop update

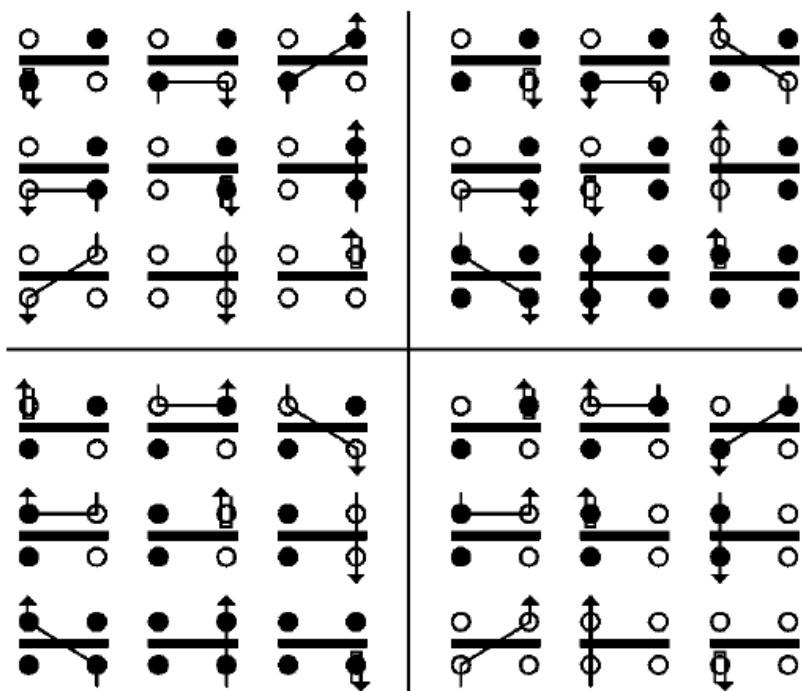
$$\langle \downarrow\downarrow | H_b | \downarrow\downarrow \rangle = \epsilon,$$

$$\langle \uparrow\downarrow | H_b | \downarrow\uparrow \rangle = \langle \uparrow\downarrow | H_b | \uparrow\downarrow \rangle = \Delta/2 + h_b + \epsilon,$$

$$\langle \uparrow\downarrow | H_b | \downarrow\uparrow \rangle = \langle \downarrow\uparrow | H_b | \uparrow\downarrow \rangle = 1/2,$$

$$\langle \uparrow\uparrow | H_b | \uparrow\uparrow \rangle = \epsilon + 2h_b.$$

Which types are acceptant??

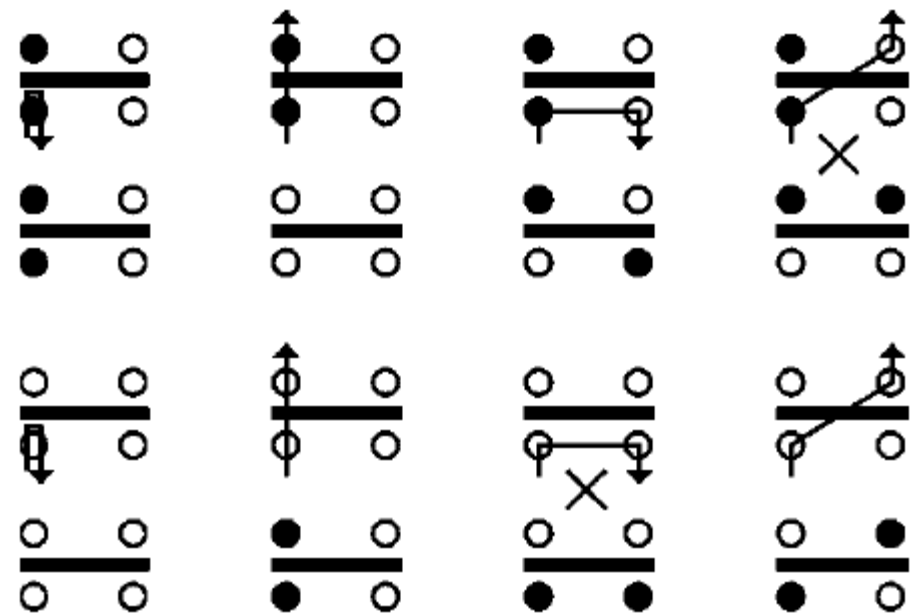


$$a = W_1 W_2 / (W_1 + W_2 + W_3),$$

$$b = W_1 W_3 / (W_1 + W_2 + W_3),$$

$$c = W_2 W_3 / (W_1 + W_2 + W_3),$$

$$b_i = W_i^2 / (W_1 + W_2 + W_3).$$



Because of symmetry reasons,  
there are only two different types of sets

$$W_1 = b_1 + a + b,$$

$$W_1 = b'_1 + a' + b',$$

$$W_2 = a + b_2 + c,$$

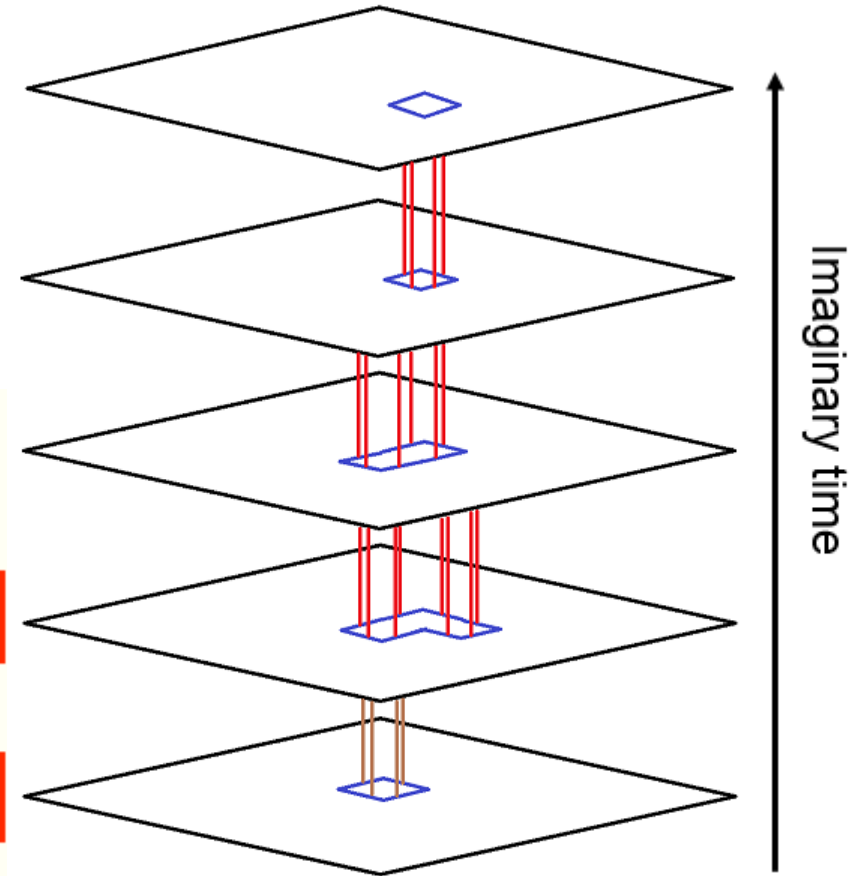
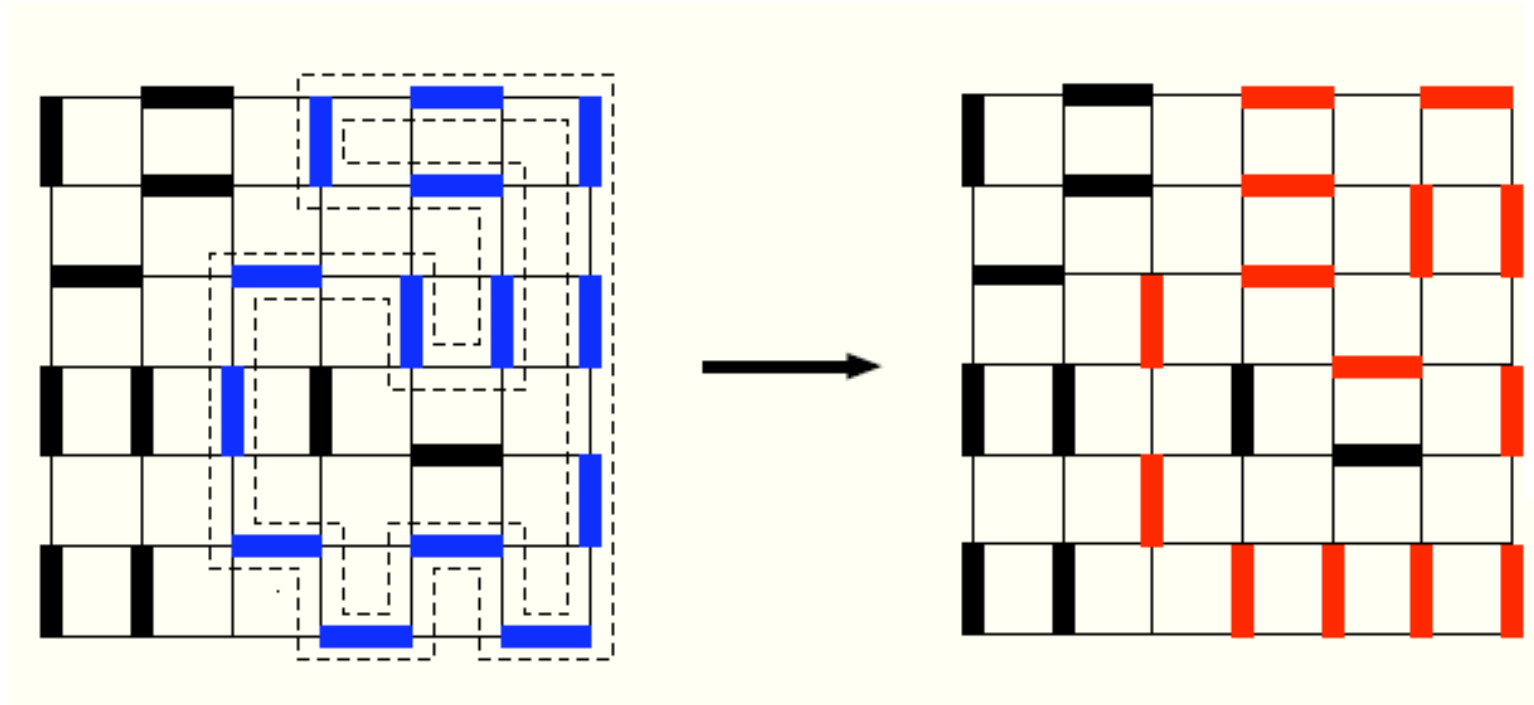
$$W_2 = a' + b'_2 + c',$$

$$W_3 = b + c + b_3,$$

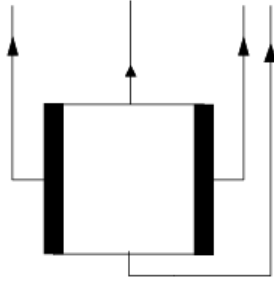
$$W_4 = b' + c' + b'_3.$$

# Sweeping cluster algorithm

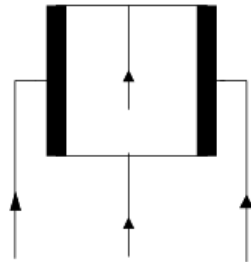
$$H = -J \sum_{\text{plaq}} \left( |\uparrow\uparrow\rangle\langle\downarrow\downarrow| + \text{H.c.} \right) + V \sum_{\text{plaq}} \left( |\uparrow\uparrow\rangle\langle\uparrow\uparrow| + |\downarrow\downarrow\rangle\langle\downarrow\downarrow| \right)$$



# Along imaginary time-Sweeping cluster

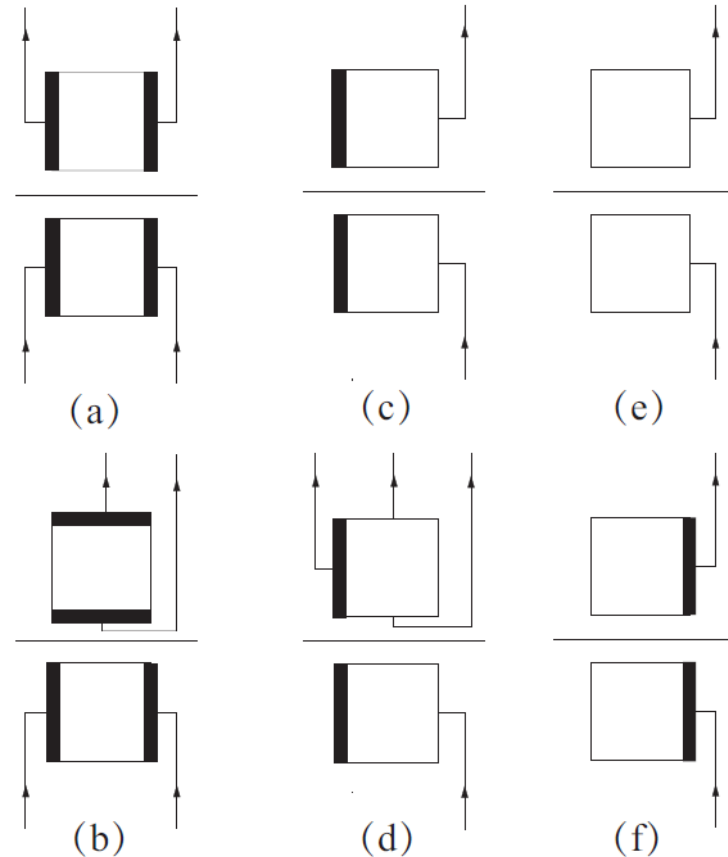


Select a flippable block to start the update, and generate/destroy a dimer where the update line passes.

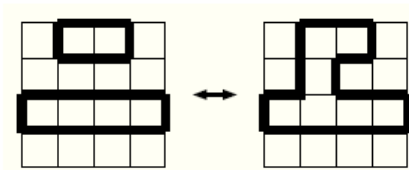


Cluster completion until only 4 update lines remain and terminate in a flippable plaquette.

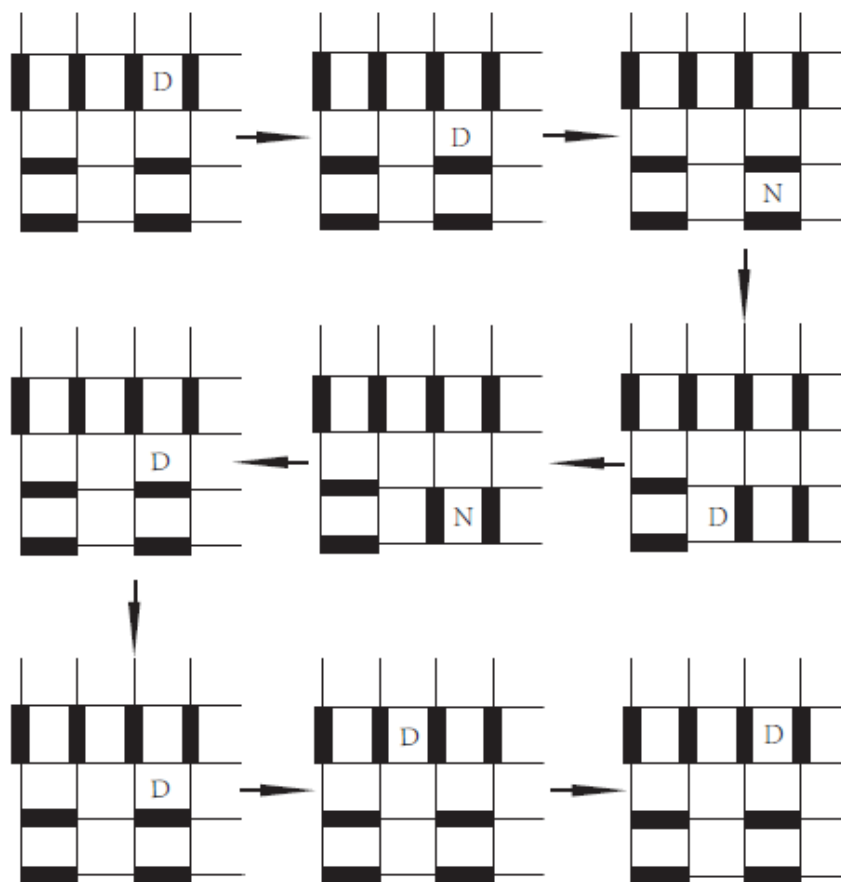
$$H = -J \sum_{\text{plaq}} \left( |11\rangle \langle \bar{1}\bar{1}| + \text{H.c.} \right) + V \sum_{\text{plaq}} \left( |11\rangle \langle 11| + |\bar{1}\bar{1}\rangle \langle \bar{1}\bar{1}| \right)$$



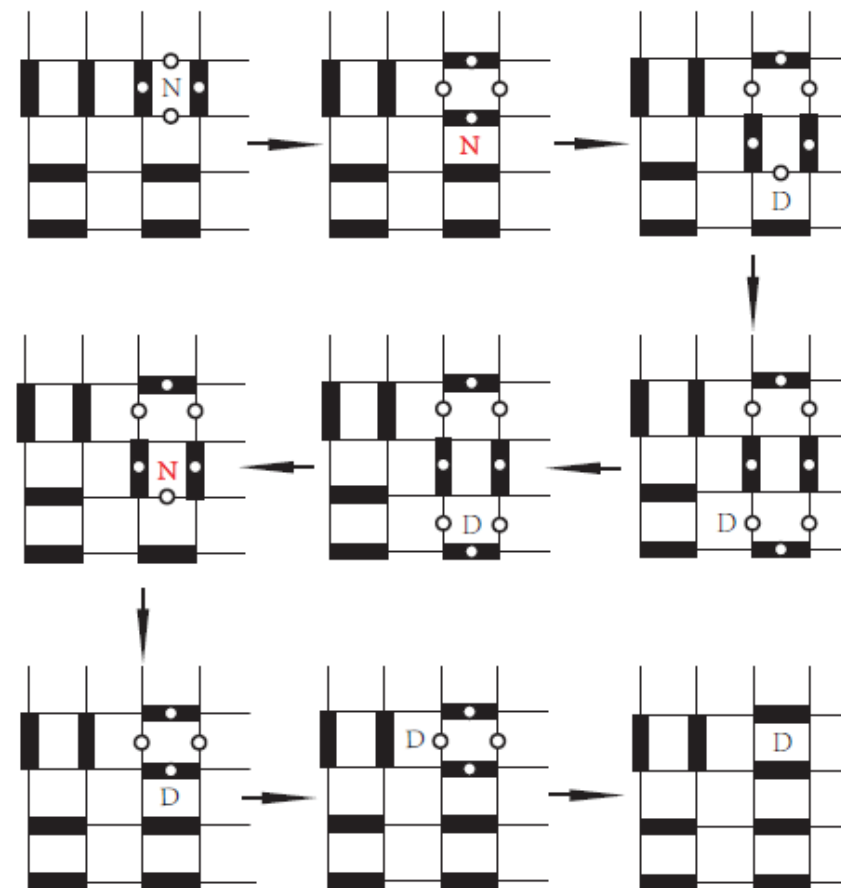
# Details show



There are cases of split/synthesis of loops!



(a)



(b)

Thanks for your attentions!

