

5. Quantum Monte Carlo



Quantum Monte Carlo

- Not as easy as classical Monte Carlo

$$Z = \sum_c e^{-E_c / k_B T}$$

- Calculating the eigenvalues E_c is equivalent to solving the problem
- Need to find a mapping of the quantum partition function to a classical problem

$$Z = \text{Tr} e^{-\beta H} \equiv \sum_c p_c$$

- “Negative sign” problem if some $p_c < 0$

Quantum Monte Carlo

- Feynman (1953) lays foundation for quantum Monte Carlo
- Map quantum system to classical world lines



THE PHYSICAL REVIEW

A journal of experimental and theoretical physics established by E. L. Nichols in 1893

SECOND SERIES, VOL. 91, No. 6

SEPTEMBER 15, 1953

Atomic Theory of the λ Transition in Helium

R. P. FEYNMAN

California Institute of Technology, Pasadena, California

(Received May 15, 1953)

Use Metropolis algorithm to update world lines

Diagrammatic QMC

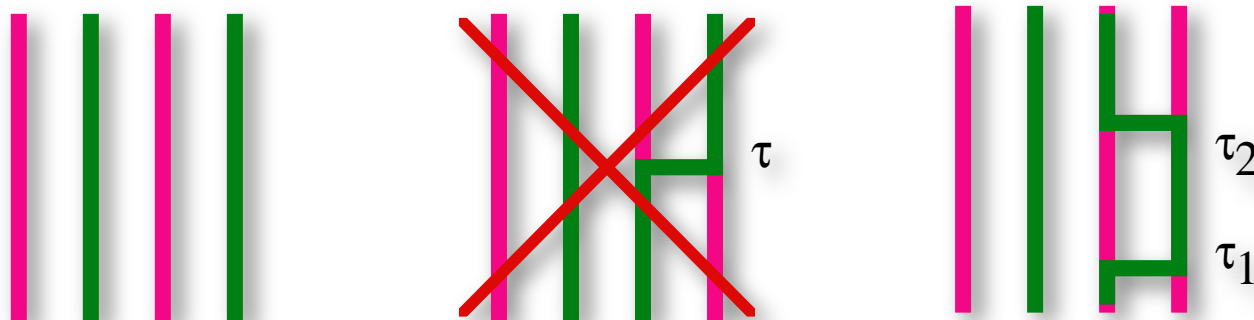
- Split the Hamiltonian into diagonal term H_0 and perturbation V
- Then perform time-dependent perturbation theory

$$H = H_0 + V, \quad H_0 = \sum_{\langle i,j \rangle} J_{ij}^z S_i^z S_j^z - \sum_i h S_i^z, \quad V = \sum_{\langle i,j \rangle} J_{ij}^{xy} (S_i^x S_j^x + S_i^y S_j^y)$$

$$Z = \text{Tr}(e^{-\beta H}) = \text{Tr}(e^{-\beta H_0} \mathcal{T} e^{-\int_0^\beta d\tau V(\tau)})$$

$$Z = \text{Tr}(e^{-\beta H_0} (1 - \int_0^\beta d\tau V(\tau) + \int_0^\beta d\tau_1 \int_{\tau_1}^\beta d\tau_2 V(\tau_1) V(\tau_2) + \dots))$$

- Each term is represented by a diagram (world line configuration)



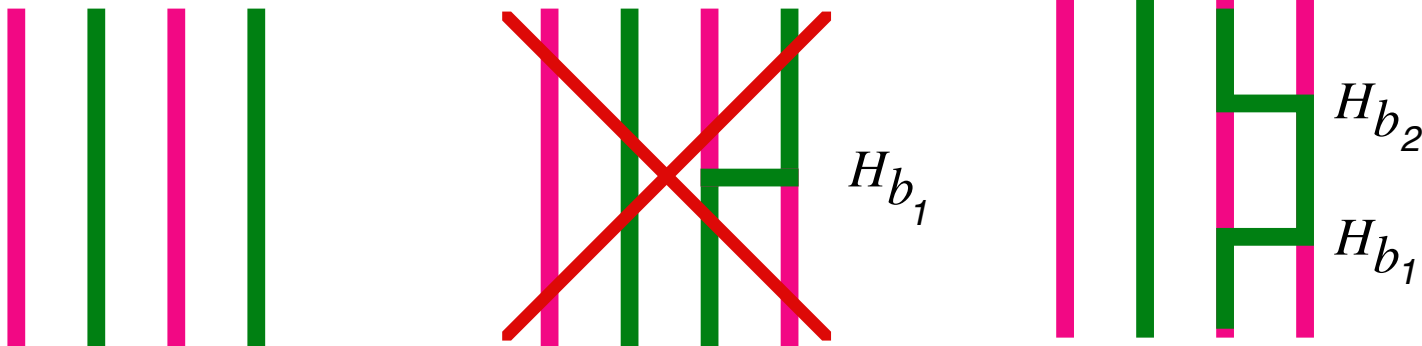
Stochastic Series Expansion

- based on high temperature expansion, developed by Sandvik

$$Z = \text{Tr}(e^{-\beta H}) = \sum_{n=0}^{\infty} (-\beta)^n \text{Tr}(H^n)$$

$$= \sum_{n=0}^{\infty} \frac{\beta^n}{n!} \sum_{|\alpha\rangle} \sum_{(b_1, \dots, b_n)} \langle \alpha | \prod_{i=1}^n (-H_{b_i}) | \alpha \rangle$$

with $H = \sum_i H_i$



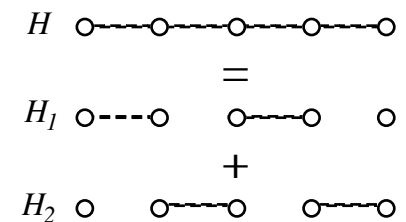
- Similar world line representation but without times assigned

The Suzuki-Trotter Decomposition

- Generic mapping of a quantum spin system to Ising model
 - basis of most discrete time QMC algorithms
 - not limited to special models
- Split Hamiltonian into two easily diagonalized pieces

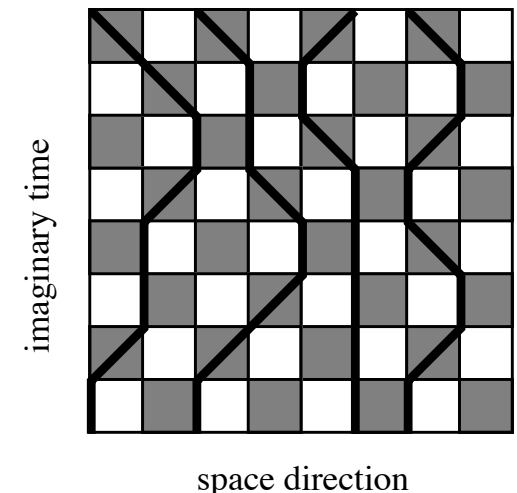
$$H = H_1 + H_2$$

$$e^{-\epsilon H} = e^{-\epsilon(H_1 + H_2)} = e^{-\epsilon H_1} e^{-\epsilon H_2} + O(\epsilon^2)$$



- Obtain the checkerboard decomposition

$$\begin{aligned} Z &= \text{Tr}[\exp(-\beta H)] = \text{Tr}[e^{-\beta(H_1 + H_2)}] \\ &= \text{Tr}[e^{-(\beta/M)H_1} e^{-(\beta/M)H_2}]^M + O(\beta^3 / M^2) \end{aligned}$$



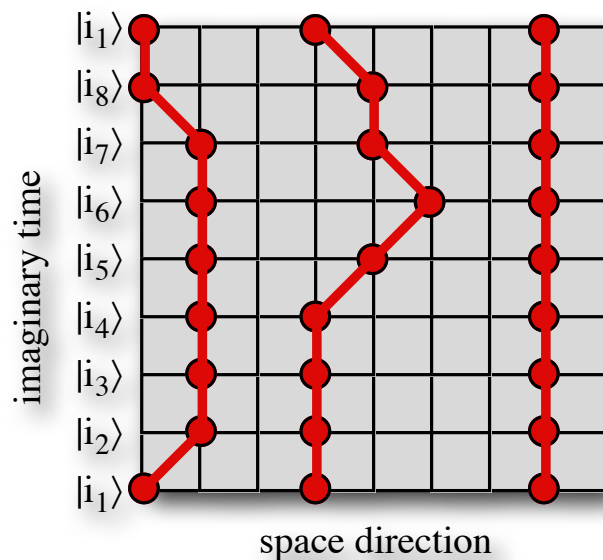
Path integral QMC

- Use Trotter-Suzuki or a simple low-order formula

$$Z = \text{Tr} e^{-\beta H} = \text{Tr} e^{-M\Delta\tau H} = \text{Tr} (e^{-\Delta\tau H})^M = \text{Tr} (1 - \Delta\tau H)^M + O(\beta\Delta\tau)$$

$$= \sum_{\{(i_1 \dots i_M)\}} \langle i_1 | 1 - \Delta\tau H | i_2 \rangle \langle i_2 | 1 - \Delta\tau H | i_3 \rangle \dots \langle i_M | 1 - \Delta\tau H | i_1 \rangle$$

- gives a mapping to a $(d+1)$ -dimensional classical model



place particles (spins)

for Hamiltonians conserving
particle number (magnetization)
we get world lines

- partition function of quantum system is sum over classical world lines

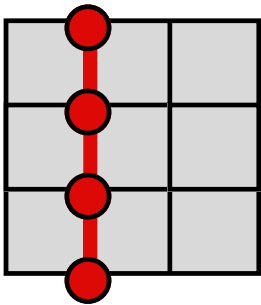
Calculating configuration weights

$$Z = \text{Tr} e^{-\beta H} = \text{Tr} e^{-M\Delta\tau H} = \text{Tr} (e^{-\Delta\tau H})^M = \text{Tr} (1 - \Delta\tau H)^M + O(\beta\Delta\tau)$$

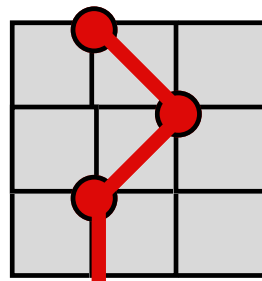
$$= \sum_{\{(i_1 \dots i_M)\}} \langle i_1 | 1 - \Delta\tau H | i_2 \rangle \langle i_2 | 1 - \Delta\tau H | i_3 \rangle \dots \langle i_M | 1 - \Delta\tau H | i_1 \rangle$$

- Examples: particles with nearest neighbor repulsion

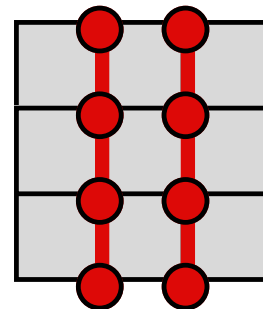
$$H = -t \sum_{\langle i,j \rangle} (a_i^\dagger a_j + a_j^\dagger a_i) + V \sum_{\langle i,j \rangle} n_i n_j$$



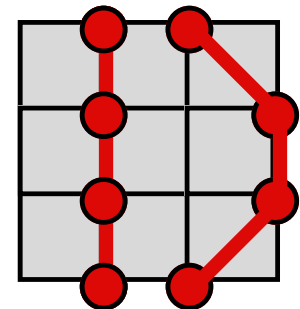
1



$(\Delta\tau t)^2$



$(1 - \Delta\tau V)^3$



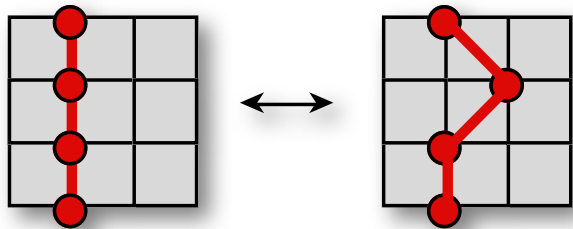
$(\Delta\tau t)^2$

Monte Carlo updates

- just move the world lines locally
 - probabilities given by matrix element of Hamiltonian
 - example: tight binding model

$$H = -t \sum_{\langle i,j \rangle} (c_i^\dagger c_{i+1} + c_{i+1}^\dagger c_i)$$

introduce or remove two kinks:

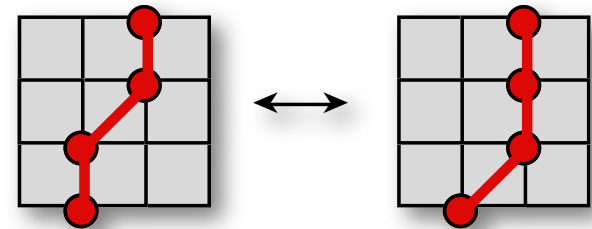


$$P = 1 \quad P = (\Delta\tau t)^2$$

$$P_{\rightarrow} = \min[1, (\Delta\tau t)^2]$$

$$P_{\leftarrow} = \min[1, 1/(\Delta\tau t)^2]$$

shift a kink:



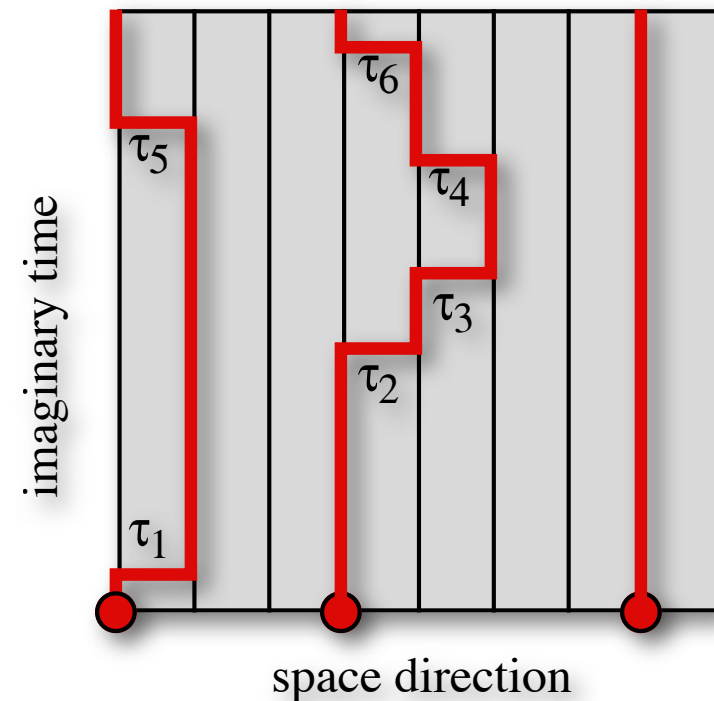
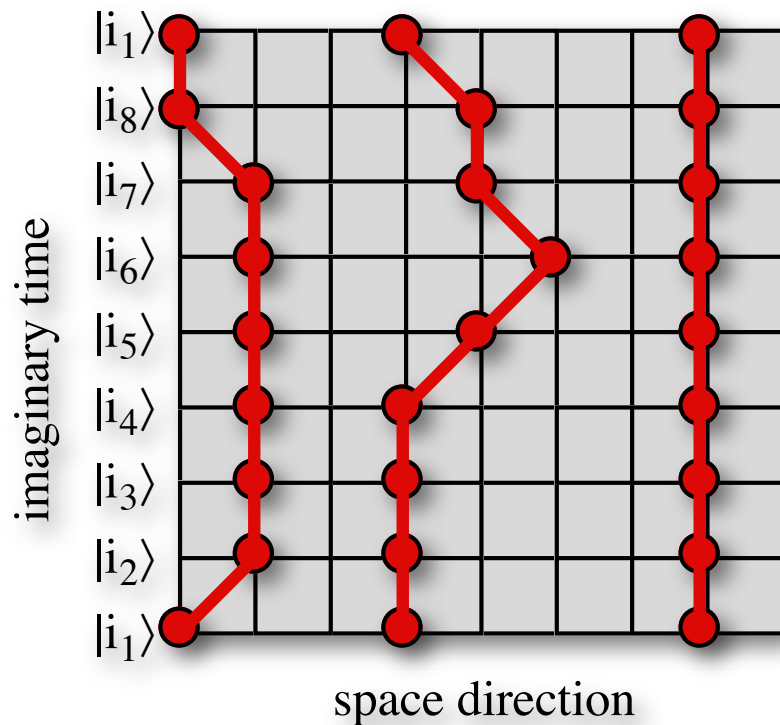
$$P = \Delta\tau t$$

$$P_{\rightarrow} = P_{\leftarrow} = 1$$

The continuous time limit

- the limit $\Delta\tau \rightarrow 0$ can be taken in the algorithm

[Prokof'ev *et al.*, Pis'ma v Zh.Eks. Teor. Fiz. **64**, 853 (1996)]



- discrete time: store configuration at all time steps
- continuous time: store times at which configuration changes

Calculating configuration weights

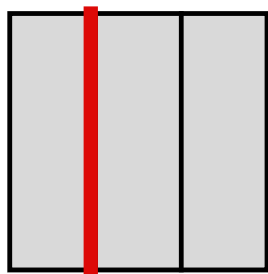
- Continuous time algorithms just sample time-dependent perturbation expansion

$$Z = \text{Tr} \left(e^{-\beta H_0} \mathcal{T} e^{-\int_0^\beta d\tau \mathcal{V}(\tau)} \right)$$

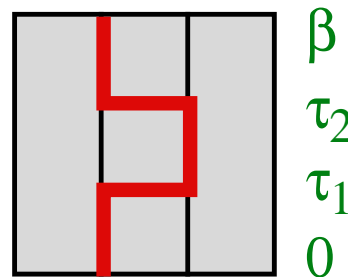
- Examples: particles with nearest neighbor repulsion

$$H_0 = V \sum_{\langle i,j \rangle} n_i n_j$$

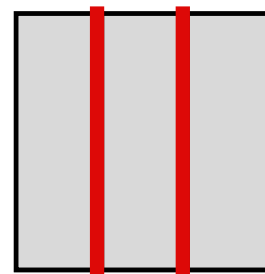
$$\mathcal{V} = -t \sum_{\langle i,j \rangle} (a_i^\dagger a_j + a_j^\dagger a_i)$$



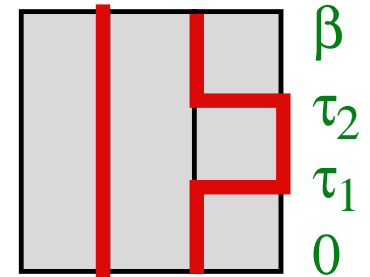
1



$t^2 d\tau_1 d\tau_2$



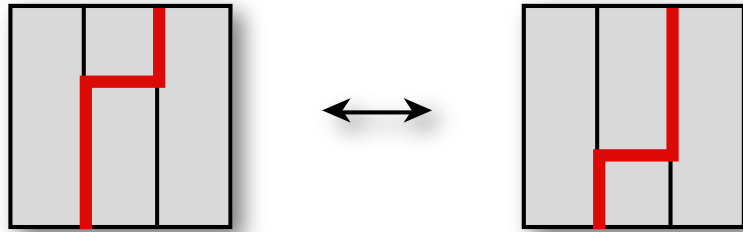
$e^{-\beta V}$



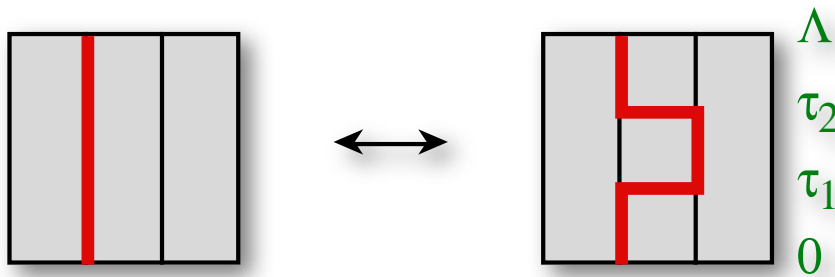
$e^{-\tau_1 V} e^{-(\beta - \tau_2) V} t^2 d\tau_1 d\tau_2$

Updates in continuous time

- Shift a kink to any new position:



- Insert a pair of kinks:



$$P = 1$$

~~$$P = (\Delta\tau t)^2 \rightarrow 0$$~~

~~$$P_{\rightarrow} = \min[1, (\Delta\tau t)^2] \rightarrow 0$$~~

*vanishing
acceptance rate*

solution:

*integrate over all possible
insertions in an interval*

$$P = \int_0^{\Lambda} \int_{\tau_1}^{\Lambda} t^2 d\tau_2 d\tau_1 = \frac{\Lambda^2 t^2}{2} \neq 0$$

$$P_{\rightarrow} = \min[1, \Lambda^2 t^2 / 2] \neq 0$$

Advantages of continuous time

- No need to extrapolate in time step
 - a single simulation is sufficient
 - no additional errors from extrapolation
- Less memory and CPU time required
 - Instead of a time step $\Delta\tau \ll t$ we only have to store changes in the configuration happening at mean distances $\approx t$
 - Speedup of $t / \Delta\tau \approx 10$
- Conceptual advantage
 - we directly sample a diagrammatic perturbation expansion

6. Cluster updates and the loop algorithm



Problems with local updates

- Local updates cannot change global topological properties
 - number of world lines (particles, magnetization) conserved
 - winding conserved
 - braiding conserved
 - cannot sample grand-canonical ensemble
- Critical slowing down at second order phase transitions
 - solved by cluster updates (today)
- Tunneling problem at first order phase transitions
 - solved by extended sampling techniques (S. Trebst's lecture)

Cluster algorithms: the formal explanation

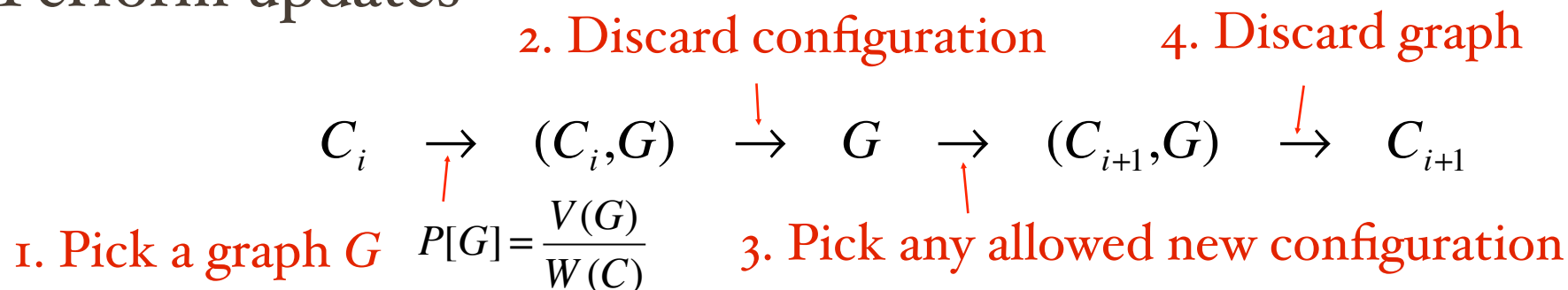
- Extend the phase space to **configurations + graphs** (C, G)

$$Z = \sum_C W(C) = \sum_C \sum_G W(C, G) \text{ with } W(C) = \sum_G W(C, G)$$

- Choose graph weights independent of configuration

$$W(C, G) = \Delta(C, G) V(G) \text{ where } \Delta(C, G) = \begin{cases} 1 & \text{graph } G \text{ allowed for } C \\ 0 & \text{otherwise} \end{cases}$$

- Perform updates



- Detailed balance is satisfied

$$\frac{P[(C_i, G) \rightarrow (C_{i+1}, G)]}{P[(C_{i+1}, G) \rightarrow (C_i, G)]} = \frac{1/N_C}{1/N_C} = 1 = \frac{\Delta(C_{i+1}, G) V(G)}{\Delta(C_i, G) V(G)} = \frac{P[(C_{i+1}, G)]}{P[(C_i, G)]}$$

Cluster algorithms: Ising model

- We need to find $\Delta(C,G)$ and $V(G)$ to fulfill $W(C) = \sum_G W(C,G) = \sum_G \Delta(C,G)V(G)$

$\Delta(C,G)$	o-o	o o	$W(C)$
$\uparrow\uparrow, \downarrow\downarrow$	I	I	$e^{+\beta J}$
$\uparrow\downarrow, \downarrow\uparrow$	O	I	$e^{-\beta J}$
$W(G)$	$e^{+\beta J} - e^{-\beta J}$	$e^{-\beta J}$	

- This means for: $C_i \rightarrow (C_i, G) \rightarrow G$
 - Parallel spins: pick connected graph **o-o** with $P(\text{ **o-o** }) = \frac{e^{+\beta J} + e^{-\beta J}}{e^{+\beta J}} = 1 - e^{-2\beta J}$
 - Antiparallel spins: always pick open graph **o o**
- And for: $G \rightarrow (C_{i+1}, G) \rightarrow C_{i+1}$
 - Configuration must be allowed $\Rightarrow$ connected spins must be parallel
 $\Rightarrow$ connected spins flipped as one cluster

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Cluster Algorithm for Vertex Models

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(Received 17 November 1992)

We present a new type of cluster algorithm that strongly reduces critical slowing down in simulations of vertex models. Since the clusters are closed paths of bonds, we call it the *loop algorithm*. The basic steps in constructing a cluster are the breakup and the freezing of vertices. We concentrate on the case of the F model, which is a subset of the six-vertex model exhibiting a Kosterlitz-Thouless transition. The loop algorithm is also applicable to simulations of other vertex models and of one- and two-dimensional quantum spin systems.

PACS numbers: 02.70.-c, 05.50.+q, 68.35.Rh, 75.10.Jm

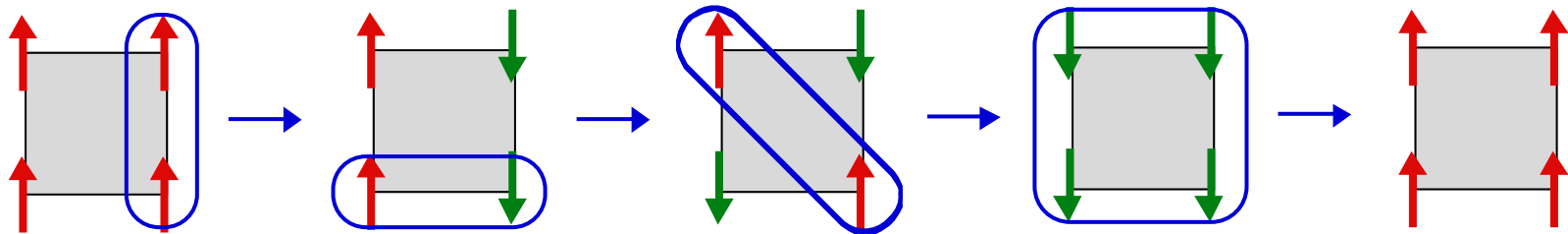
The loop algorithm (Evertz *et al*)

- Swendsen-Wang cluster algorithm for the Ising model
 - two choices on each bond: *connected (flip both spins)* or *disconnected*

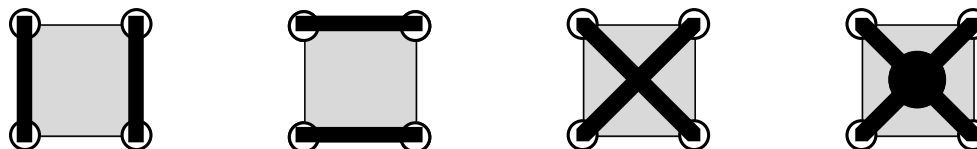


- all connected spins are flipped together
- Loop algorithm is a generalization to quantum systems

- world lines may not be broken
- always 2 or 4 spins must be flipped together



- four different connection types



Hamiltonian of spin-1/2 models

- Consider a 2-site quantum spin-1/2 model

$$\begin{aligned} H_{XXZ} &= J_{xz}(S_1^x S_2^x + S_1^y S_2^y) + J_z S_1^z S_2^z - h(S_1^z + S_2^z) \\ &= \frac{J_{xz}}{2}(S_1^+ S_2^- + S_1^- S_2^+) + J_z S_1^z S_2^z - h(S_1^z + S_2^z) \end{aligned}$$

- Heisenberg model if $J_{xy} = J_z = J$

$$H = J\vec{S}_1\vec{S}_2 - h(S_1^z + S_2^z)$$

- Hamiltonian matrix in 2-site basis

$$\{ |\uparrow\uparrow\rangle, |\uparrow\downarrow\rangle, |\downarrow\uparrow\rangle, |\downarrow\downarrow\rangle \}$$

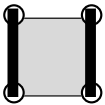
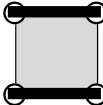
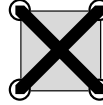
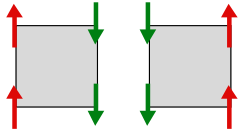
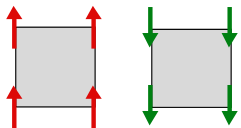
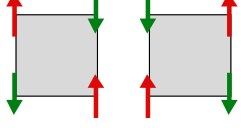
$$H_{XXZ} = \begin{pmatrix} \frac{J_z}{4} + h & 0 & 0 & 0 \\ 0 & -\frac{J_z}{4} & \frac{J_{xy}}{2} & 0 \\ 0 & \frac{J_{xy}}{2} & -\frac{J_z}{4} & 0 \\ 0 & 0 & 0 & \frac{J_z}{4} - h \end{pmatrix}$$

Cluster building rules: XY-like antiferromagnet

$$H_{XXZ} = \frac{J_{xz}}{2} \sum_{\langle i,j \rangle} (S_i^+ S_j^- + S_i^- S_j^+) + J_z \sum_{\langle i,j \rangle} S_i^z S_j^z$$

with $0 \leq J_z \leq J_{xy}$

$$W(C) = \sum_G W(C, G) = \sum_G \Delta(C, G) V(G)$$

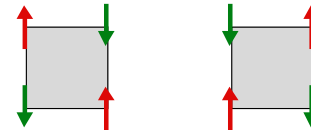
$\Delta(C, G)$				$W(C)$
	I	I	—	$1 + (J_z/4) d\tau$
	I	—	O	$1 - (J_z/4) d\tau$
	—	I	I	$(J_{xy}/2) d\tau$
$V(G)$	$1 - (J_z/4) d\tau$	$(J_z/2) d\tau$	$(J_{xy} - J_z)/2 d\tau$	

- Connected spins form a cluster and have to be flipped together

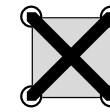
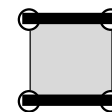
How to deal with infinitesimals ?

- How do we deal with the vanishing $d\tau$ terms in continuous time?

- First example: the exchange process



- Possible graph connections:



- Graph weights:

$$\frac{J_z}{2} d\tau$$

$$\frac{J_{xy} - J_z}{2} d\tau$$

- Probability to pick graph:
(divide weight by sum)

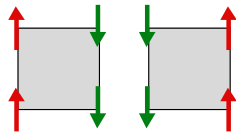
$$\frac{J_z}{J_{xy}}$$

$$\frac{J_{xy} - J_z}{J_{xy}}$$

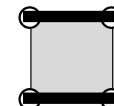
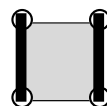
- The infinitesimal $d\tau$ terms cancel out
 - Randomly pick one of the graphs (with appropriate probabilities)

How to deal with infinitesimals ?

- How do we deal with the vanishing $d\tau$ terms in continuous time?

- Second example: the “decay” process 

- Possible graph connections:



- Graph weights:

$$1 - \frac{J_z}{4} d\tau$$

$$\frac{J_z}{2} d\tau$$

- Probability to pick graph:
(divide weight by sum)

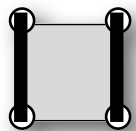
$$1 - \frac{J_z}{2} d\tau$$

$$\frac{J_z}{2} d\tau$$

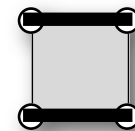
- The infinitesimal $d\tau$ terms remain
- Infinitesimal acceptance rate at infinitely many time steps?

How to deal with infinitesimals ?

- We have to tackle the problem of vanishing probabilities
 - Example: Heisenberg antiferromagnet



$$P_{\parallel} = 1 - \frac{J}{2} \Delta\tau \rightarrow 1$$



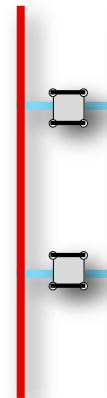
$$P_{\perp} = \frac{J}{2} \Delta\tau \rightarrow 0$$

- Interpret the  connection as a “decay process” where the loop jumps

$$P_{\perp} = \frac{J}{2} d\tau$$

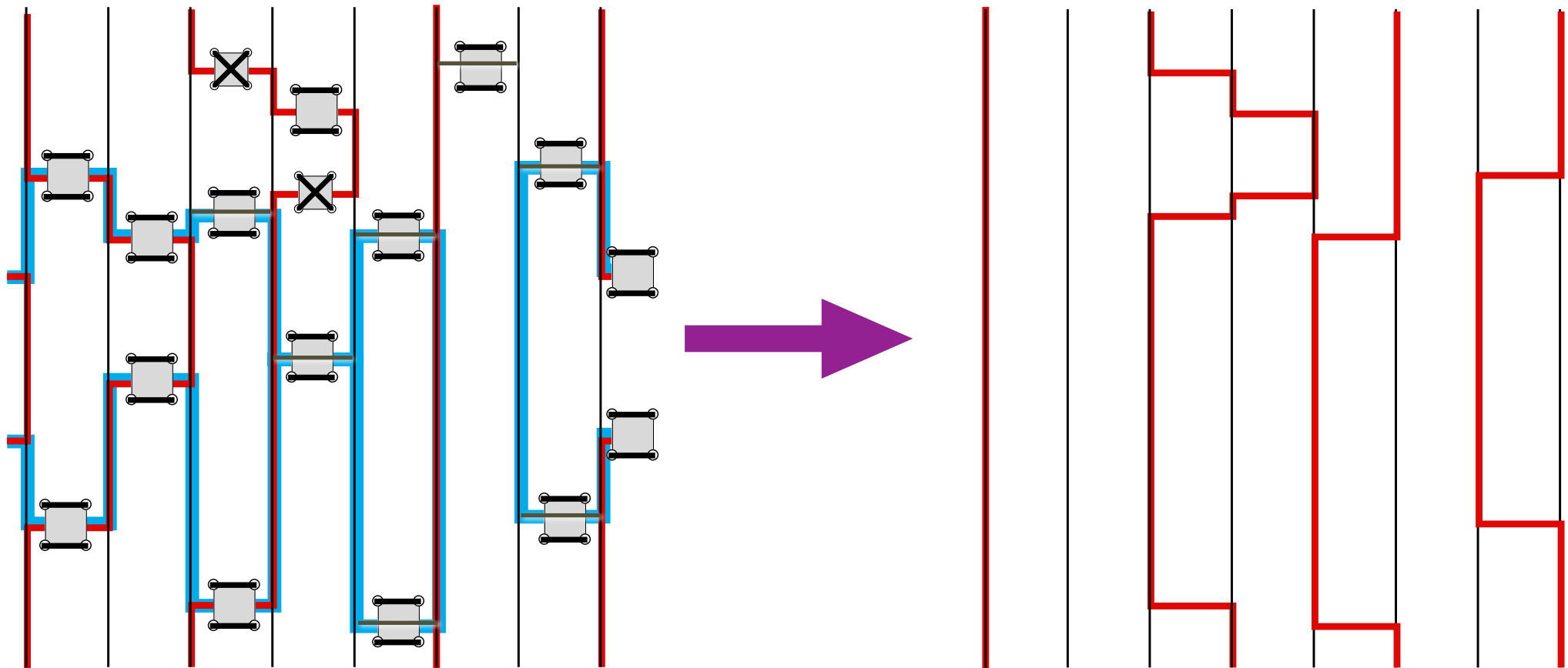
$$\text{decay constant } \lambda = \frac{J}{2}$$

$$\text{mean distance } d = \frac{2}{J}$$



Loop-cluster updates

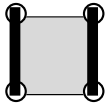
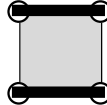
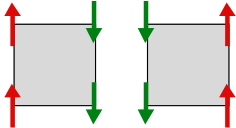
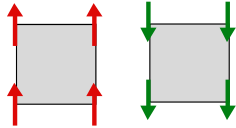
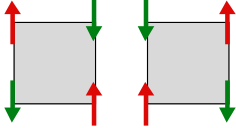
1. Connect spins according to loop-cluster building rules
2. Build and flip loop-cluster



Heisenberg antiferromagnet

$$H_{\text{Heisenberg}} = J \sum_{\langle i,j \rangle} \vec{S}_i \vec{S}_j$$

$$W(C) = \sum_G W(C,G) = \sum_G \Delta(C,G) V(G)$$

$\Delta(C,G)$			$W(C)$
	I	I	$1 + (J/4) d\tau$
	I	O	$1 - (J/4) d\tau$
	O	I	$(J/2) d\tau$
$V(G)$	$1 - (J/4) d\tau$	$(J/2) d\tau$	

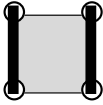
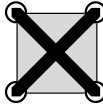
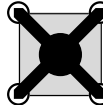
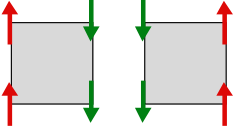
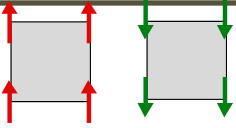
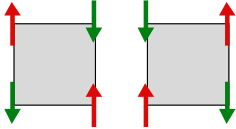
- Connected spins form a cluster and have to be flipped together
- Very simple and deterministic for Heisenberg model

Ising-like ferromagnet

$$H_{XXZ} = -\frac{J_{xz}}{2} \sum_{\langle i,j \rangle} (S_i^+ S_j^- + S_i^- S_j^+) - J_z \sum_{\langle i,j \rangle} S_i^z S_j^z$$

$$W(C) = \sum_G W(C,G) = \sum_G \Delta(C,G) V(G)$$

with $0 \leq J_{xy} \leq J_z$

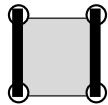
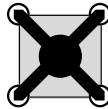
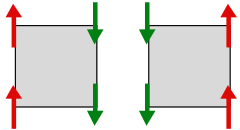
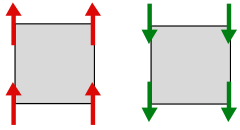
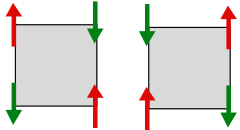
$\Delta(C,G)$				$W(C)$
	I	O	O	$1 - (J_z/4) d\tau$
	I	I	I	$1 + (J_z/4) d\tau$
	O	I	O	$(J_{xy}/2) d\tau$
$V(G)$	$1 - (J_z/4) d\tau$	$(J_{xy}/2) d\tau$	$(J_z - J_{xy})/2 d\tau$	

- Now 4-spin freezing graph is needed: connects (freezes) loops

Ising ferromagnet

$$H_{\text{Ising}} = -J \sum_{\langle i,j \rangle} S_i^z S_j^z = -\frac{J}{4} \sum_{\langle i,j \rangle} \sigma_i \sigma_j$$

$$W(C) = \sum_G W(C,G) = \sum_G \Delta(C,G) V(G)$$

$\Delta(C,G)$			$W(C)$
	I	O	$1 - (J/4) d\tau$
	I	I	$1 + (J/4) d\tau$
	O	O	0
$V(G)$	$1 - (J/4) d\tau$	$(J/2) d\tau$	

- Two spins are frozen if there is any freezing graph along the world line

$$P_{\text{no freezing}} = \lim_{M \rightarrow \infty} (1 - (\beta/M)J/2)^M = \exp(-\beta J/2) = \exp(-2\beta J_{\text{classical}})$$

- We recover the Swendsen Wang algorithm: probability for no freezing

Modern Monte Carlo algorithms

- Which system sizes can be studied?

temperature	Metropolis	modern algorithms
$3D T_c$	16'000 spins	16'000'000 spins
0.1 J	200 spins	1'000'000 spins
0.005 J	—	50'000 spins
$3D T_c$	32 bosons	1'000'000 bosons
0.1 t	32 bosons	10'000 bosons