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Introduction to quantum spin systems and quantum Monte Carlo simulations

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- Classical spins and classical Monte Carlo
- Path integrals in quantum statistical mechanics
- Stochastic Series expansion (incl technical details)
- Ground-state projection with valence bonds

Review article on quantum spin systems and numerical methods: ArXiv:1101.3281

Classical spin systems and Monte Carlo simulations

Classical spin models

Lattice models with "spin" degrees of freedom at the vertices

Classified by type of spin:

- Ising model: discrete spins, normally two-state $\sigma_i = -1, +1$
- XY model: planar vector spins (fixed length)
- Heisenberg model: 3-dimensional vector spins.

Statistical mechanics

- spin configurations C
- energy E(C)

C

- some quantity Q(C)
- temperature T (k_B=1)

$$\langle Q \rangle = \frac{1}{Z} \sum_{C} Q(C) e^{-E(C)/T}$$

$$Z = \sum e^{-E(C)/T}$$



$$E = \sum_{\langle ij \rangle} J_{ij} \sigma_i \sigma_j \quad \text{(Ising)}$$

$$E = \sum_{\langle ij \rangle} J_{ij} \vec{S}_i \cdot \vec{S}_j = \sum_{\langle ij \rangle} J_{ij} \cos(\Theta_i - \Theta_j) \quad (XY)$$

$$E = \sum_{\langle ij \rangle} J_{ij} \vec{S}_i \cdot \vec{S}_j \qquad \text{(Heisenberg)}$$

Phase transition in the Ising model



- first-order transition versus h (at h=0) for T<T_c
- continuous transition at h=0

Mean-field solution: $J = J_i = \sum J_{ij}$ $m = \tanh[(Jm + h)/T], \quad (m = \langle \sigma_i \rangle)$ 1.0 Here J is the 1.0 $T/T_c = 0.8$ -- $T/T_c = 1.0$ -- $T/T_c = 2.0$ h=0sum of local 0.5 0.8 couplings ε^{0.6} E 0.0 $J = \sum_{i} J_{ij}$ 0.4 -0.5 0.2 (a)(b) -1.0 0.0∟ 0.0 0.2 0.2 -0.2 0.0 0.8 0.4 0.6 1.0 0.4 -0.4h/JT/J

Monte Carlo simulation of the Ising model

The Metropolis algorithm

[Metropolis, Rusenbluth, Rosenbluth, Teller, and Teller, Phys. Rev. 1953]

Generate a series of configurations (Markov chain); $C_1 \rightarrow C_2 \rightarrow C_3 \rightarrow C_4 \rightarrow ...$

- C_{n+1} obtained by modifying (updating) C_n
- changes satisfy the **detailed-balance principle**

 $\frac{P_{\text{change}}(A \to B)}{P_{\text{change}}(B \to A)} = \frac{W(B)}{W(A)} \qquad W(A) = e^{-E(A)/T}$

$$P_{\text{change}}(A \to B) = P_{\text{select}}(B|A)P_{\text{accept}}(B|A)$$

$$P_{\text{select}} = 1/N, \quad P_{\text{accept}} = \min[W(B)/W(A), 1]$$



 $\frac{W(B)}{W(A)} = e^{-\Delta_E/T} = e^{[E(A) - E(B)]/T}$

is easy to calculate (only depends on spins interacting with lipped spin)

Starting from any configuration, such a repeated stochastic process leads to configurations distributed according to W

- the process has to be ergodic
 - any configuration reachable in principle
- it takes some time to reach equilibrium (typical configurations of the Boltzmann distribution)

Metropolis algorithm for the Ising model. For each update perform:

- select a spin i at random; consider flipping it $\sigma_i \rightarrow -\sigma_i$
- compute the ratio $R=W(\sigma_1,...,\sigma_i,...,\sigma_N)/W(\sigma_1,...,\sigma_i,...,\sigma_N)$
 - for this we need only the neighbor spins of i
- generate random number 0<r≤1; accept flip if r<R (stay with old config else)
- repeat (many times...)

Simulation time unit

(Monte Carlo step or sweep) - N spin flip attempts

"Measure" physical observables (averaged over time) on the generated configurations

 begin after equilibration (when configurations are typical representatives of the Boltzmann distribution)

Example

- 128×128 lattice (N=16384) at T/J=4 (> T_c/J \approx 2.27)





Going closer to Tc



Going below Tc....

Staying at same T, speeding up time by factor 10



Time series of simulation data; magnetization vs simulation time for T<T_c



Compute time-average of <m²> to carry out **finite-size scaling**



Squared magnetization for L×L Ising lattices

Quantum spin systems, quantum antiferromagnets

Quantum spin models

- the spins have three (x,y,z) components, satisfy commutation relations
- interactions may contain 1 (Ising), 2 (XY), or 3 (Heisenberg) components

$$\begin{split} H &= \sum_{\langle ij \rangle} J_{ij} S_i^z S_j^z = \frac{1}{4} \sum_{\langle ij \rangle} J_{ij} \sigma_i \sigma_j \quad \text{(Ising)} \\ H &= \sum_{\langle ij \rangle} J_{ij} [S_i^x S_j^x + S_i^y S_j^y] = \frac{1}{2} \sum_{\langle ij \rangle} J_{ij} [S_i^+ S_j^- + S_i^- S_j^+] \quad \text{(XY)} \\ H &= \sum_{\langle ij \rangle} J_{ij} \vec{S}_i \cdot \vec{S}_j = \sum_{\langle ij \rangle} J_{ij} [S_i^z S_j^z + \frac{1}{2} (S_i^+ S_j^- + S_i^- S_j^+)] \quad \text{(Heisenberg)} \end{split}$$

+ many modifications and extensions... and local spin S=1/2,1,3/2,....

Quantum antiferromagnets

Nearest-neighbor <i,j> interactions (Heisenberg) on some lattice

$$H = J \sum_{\langle i,j \rangle} \vec{S}_i \cdot \vec{S}_j, \quad J > 0$$

Lattices can be classified as

Bipartite

- nearest-neighbors i,j always on different sublattices
- compatible with Neel order
- but other states possible

Non-bipartite

- no bipartition is possible
- frustrated antiferromagnetic interactions
- different kinds of order or no long-range order (spin liquid)

Fully ordered Neel state (ground state of H for classical spins) is not an eigenstate of H even on a bipartite lattice

- if there is order at T=0 it is reduced by quantum fluctuations

Mermin-Wagner theorem (on breaking a continuous symmetry) implies:

- No Neel order in 1D Heisenberg model
- Neel order possible only at T=0 in 2D system
- Order possible also at T>0 in 3D





Path Integrals in quantum statistical mechanics

 $\langle Q \rangle = \frac{1}{Z} \operatorname{Tr} \left\{ Q e^{-H/T} \right\} \qquad Z = \operatorname{Tr} \left\{ e^{-H/T} \right\} = \sum_{n=1}^{M-1} e^{-E_n/T}$

Quantum statistical mechanics

Large size M of the Hilbert space; M=2^N for S=1/2

- difficult problem to find the eigenstates and energies
- we may be especially interested in the ground state (T→0) (for classical systems the ground state is often trivial)

Quantum Monte Carlo

Rewrite the quantum-mechanical expectation value into a classical form

$$\langle A \rangle = \frac{\operatorname{Tr}\{Ae^{-\beta H}\}}{\operatorname{Tr}\{e^{-\beta H}\}} \to \frac{\sum_{c} A_{c} W_{c}}{\sum W_{c}}$$

Monte Carlo sampling in the space {c} with weights Wc (if positive-definite...)

Different ways of doing it

- World-line methods for spins and bosons
- Stochastic series expansion for spins and bosons
- Fermion determinant methods

For ground state calculations we can also do projection from a "trial state"

$$\begin{split} |\Psi_m\rangle \sim H^m |\Psi_0\rangle & |\Psi_m\rangle \rightarrow |0\rangle \text{ when } m \rightarrow \infty \\ |\Psi_\beta\rangle \sim e^{-\beta H} |\Psi_0\rangle & |\Psi_\beta\rangle \rightarrow |0\rangle \text{ when } \beta \rightarrow \infty \end{split}$$

Particularly simple and efficient schemes exist for S=1/2 models

$$H = -J \sum_{b=1}^{N_b} \left(\frac{1}{4} - \mathbf{S}_{i(b)} \cdot \mathbf{S}_{j(b)} \right) \qquad (+ \text{ certain multi-spin terms})$$

No sign problem on bipartite lattices

("sign problem" if not the case)

Path integrals in quantum statistical mechanics

We want to compute a thermal expectation value

$$\langle A \rangle = \frac{1}{Z} \operatorname{Tr} \{ A \mathrm{e}^{-\beta H} \}$$

where $\beta = 1/T$ (and possibly T \rightarrow 0). How to deal with the exponential operator?

"Time slicing" of the partition function

$$Z = \operatorname{Tr}\{\mathrm{e}^{-\beta H}\} = \operatorname{Tr}\left\{\prod_{l=1}^{L} \mathrm{e}^{-\Delta_{\tau} H}\right\} \qquad \Delta_{\tau} = \beta/L$$

Choose a basis and insert complete sets of states;

$$Z = \sum_{\alpha_0} \sum_{\alpha_1} \cdots \sum_{\alpha_L = 1} \langle \alpha_0 | e^{-\Delta_\tau H} | \alpha_{L-1} \rangle \cdots \langle \alpha_2 | e^{-\Delta_\tau H} | \alpha_1 \rangle \langle \alpha_1 | e^{-\Delta_\tau H} | \alpha_0 \rangle$$

Use approximation for imaginary time evolution operator. Simplest way

$$Z \approx \sum_{\{\alpha\}} \langle \alpha_0 | 1 - \Delta_{\tau} H | \alpha_{L-1} \rangle \cdots \langle \alpha_2 | 1 - \Delta_{\tau} H | \alpha_1 \rangle \langle \alpha_1 | 1 - \Delta_{\tau} H | \alpha_0 \rangle$$

Leads to error $\propto \Delta_{\tau}$. Limit $\Delta_{\tau} \to 0$ can be taken

Example: hard-core bosons

$$H = K = -\sum_{\langle i,j \rangle} K_{ij} = -\sum_{\langle i,j \rangle} (a_j^{\dagger} a_i + a_i^{\dagger} a_j) \qquad n_i = a_i^{\dagger} a_i \in \{0,1\}$$

Equivalent to S=1/2 XY model

$$H = -2\sum_{\langle i,j \rangle} (S_i^x S_j^x + S_i^y S_j^y) = -\sum_{\langle i,j \rangle} (S_i^+ S_j^- + S_i^- S_j^+), \quad S^z = \pm \frac{1}{2} \sim n_i = 0, 1$$

"World line" representation of

 $Z \approx \sum_{\{\alpha\}} \langle \alpha_0 | 1 - \Delta_\tau H | \alpha_{L-1} \rangle \cdots \langle \alpha_2 | 1 - \Delta_\tau H | \alpha_1 \rangle \langle \alpha_1 | 1 - \Delta_\tau H | \alpha_0 \rangle$



Expectation values

$$\langle A \rangle = \frac{1}{Z} \sum_{\{\alpha\}} \langle \alpha_0 | e^{-\Delta_\tau} | \alpha_{L-1} \rangle \cdots \langle \alpha_2 | e^{-\Delta_\tau H} | \alpha_1 \rangle \langle \alpha_1 | e^{-\Delta_\tau H} A | \alpha_0 \rangle$$

$$Z = \frac{1}{Z} \sum_{\{\alpha\}} \langle \alpha_0 | e^{-\Delta_\tau} | \alpha_{L-1} \rangle \cdots \langle \alpha_2 | e^{-\Delta_\tau H} | \alpha_1 \rangle \langle \alpha_1 | e^{-\Delta_\tau H} | \alpha_0 \rangle$$

We want to write this in a form suitable for MC importance sampling

$$\langle A \rangle = \frac{\sum_{\{\alpha\}} A(\{\alpha\}) W(\{\alpha\})}{\sum_{\{\alpha\}} W(\{\alpha\})} \longrightarrow \langle A \rangle = \langle A(\{\alpha\}) \rangle_W$$
$$W(\{\alpha\}) = \text{weight}$$
$$A(\{\alpha\}) = \text{estimator}$$

For any quantity diagonal in the occupation numbers (spin z):

$$A(\{\alpha\}) = A(\alpha_n) \text{ or } A(\{\alpha\}) = \frac{1}{L} \sum_{l=0}^{L-1} A(\alpha_l)$$

Measure quantities on all slices and average

- in practice full averaging may take too long and OK to do partial averages

Off-diagonal expectation values

$$\langle A \rangle = \frac{1}{Z} \sum_{\{\alpha\}} \langle \alpha_0 | e^{-\Delta_{\tau}} | \alpha_{L-1} \rangle \cdots \langle \alpha_2 | e^{-\Delta_{\tau} H} | \alpha_1 \rangle \langle \alpha_1 | e^{-\Delta_{\tau} H} A | \alpha_0 \rangle$$

In general the states $a_1,...,a_n$ contributing to Z will not contribute to <A> - more complicated measurements

Special case: term K_{ij} in the kinetic energy

Multiply and divide by the weight

$$\langle A \rangle = \frac{1}{Z} \sum_{\{\alpha\}} \langle \alpha_0 | e^{-\Delta_\tau} | \alpha_{L-1} \rangle \cdots \langle \alpha_1 | e^{-\Delta_\tau H} | \alpha_0 \rangle \frac{\langle \alpha_0 | e^{-\Delta_\tau} | \alpha_{L-1} \rangle \cdots \langle \alpha_1 | e^{-\Delta_\tau H} K_{ij} | \alpha_0 \rangle}{\langle \alpha_0 | e^{-\Delta_\tau} | \alpha_{L-1} \rangle \cdots \langle \alpha_1 | e^{-\Delta_\tau H} K_{ij} | \alpha_0 \rangle}$$

$$= \frac{1}{Z} \sum_{\{\alpha\}} W(\{\alpha\}) \frac{\langle \alpha_0 | e^{-\Delta_\tau} | \alpha_{L-1} \rangle \cdots \langle \alpha_1 | e^{-\Delta_\tau H} K_{ij} | \alpha_0 \rangle}{\langle \alpha_0 | e^{-\Delta_\tau} | \alpha_{L-1} \rangle \cdots \langle \alpha_1 | e^{-\Delta_\tau H} | \alpha_0 \rangle}$$

$$e^{-\Delta_\tau K} K_{ij} \approx K_{ij} \quad K_{ij}(\{\alpha\}) = \frac{\langle \alpha_1 | K_{ij} | \alpha_0 \rangle}{\langle \alpha_1 | 1 - \Delta_\tau K | \alpha_0 \rangle} \in \{0, \frac{1}{\Delta_\tau}\}$$

$$Average over all slices \rightarrow count number of kinetic jumps$$

$$\langle K_{ij} \rangle = \frac{\langle n_{ij} \rangle}{\beta}, \quad \langle K \rangle = -\frac{\langle n_K \rangle}{\beta} \qquad \langle K \rangle \propto N \to \langle n_K \rangle \propto \beta N$$

There should be of the order βN "jumps" of the worldlines

Including interactions

For any diagonal interaction V (Trotter, or split-operator, approximation)

 $e^{-\Delta_{\tau}H} = e^{-\Delta_{\tau}K}e^{-\Delta_{\tau}V} + \mathcal{O}(\Delta_{\tau}^2) \to \langle \alpha_{l+1} | e^{-\Delta_{\tau}H} | \alpha_l \rangle \approx e^{-\Delta_{\tau}V_l} \langle \alpha_{l+1} | e^{-\Delta_{\tau}K} | \alpha_l \rangle$

Product over all times slices \rightarrow

$$W(\{\alpha\}) = \Delta_{\tau}^{n_{K}} \exp\left(-\Delta_{\tau} \sum_{l=0}^{L-1} V_{l}\right)$$

$$P_{\rm acc} = \min\left[\Delta_{\tau}^2 \exp\left(-\frac{V_{\rm new}}{V_{\rm old}}\right), 1\right]$$

The continuous time limit

Limit $\Delta_{\tau} \rightarrow 0$: number of kinetic jumps remains finite, store events only



Special methods (**loop and worm updates**) developed for efficient sampling of the paths in the continuum



local updates (problem when Δ_τ→0?)
consider probability of inserting/removing events within a time window

⇐ Evertz, Lana, Marcu (1993), Prokofev et al (1996) Beard & Wiese (1996)

Stochastic series expansion (SSE)

Alternative to path integral: Series expansion representation

Start from the Taylor expansion
$$e^{-\beta H} = \sum_{n=0}^{\infty} \frac{(-\beta)^n}{n!} H^n$$

(approximation-free method from the outset)

$$Z = \sum_{n=0}^{\infty} \frac{(-\beta)^n}{n!} \sum_{\{\alpha\}_n} \langle \alpha_0 | H | \alpha_{n-1} \rangle \cdots \langle \alpha_2 | H | \alpha_1 \rangle \langle \alpha_1 | H | \alpha_0 \rangle$$

Similar to the path integral; $1 - \Delta \tau H \rightarrow H$ and weight factor outside

For hard-core bosons the (allowed) path weight is $W(\{\alpha\}_n) = \beta^n/n!$

For any model, the energy is

$$E = \frac{1}{Z} \sum_{n=0}^{\infty} \frac{(-\beta)^n}{n!} \sum_{\{\alpha\}_{n+1}} \langle \alpha_0 | H | \alpha_n \rangle \cdots \langle \alpha_2 | H | \alpha_1 \rangle \langle \alpha_1 | H | \alpha_0 \rangle$$

$$= -\frac{1}{Z} \sum_{n=1}^{\infty} \frac{(-\beta)^n}{n!} \frac{n}{\beta} \sum_{\{\alpha\}_n} \langle \alpha_0 | H | \alpha_{n-1} \rangle \cdots \langle \alpha_2 | H | \alpha_1 \rangle \langle \alpha_1 | H | \alpha_0 \rangle = \frac{\langle n \rangle}{\beta}$$
relabel terms to "get rid of" extra slice

$$C = \langle n^2 \rangle - \langle n \rangle^2 - \langle n \rangle$$

From this follows: narrow n-distribution with $\langle n \rangle \propto N \beta$, $\sigma_n \propto \sqrt{N \beta}$

Fixed-length scheme

- n fluctuating \rightarrow varying size of the configurations
- the expansion can be truncated at some n_{max}=M (exponentially small error)
- cutt-off at n=M, fill in operator string with unit operators H₀=I

n=10 H₄ H₇ H₁ H₆ H₂ H₁ H₈ H₃ H₃ H₅ \Longrightarrow

 $\mathbf{M} = 14 \quad \mathbf{H}_4 \quad \mathbf{I} \quad \mathbf{H}_7 \quad \mathbf{I} \quad \mathbf{H}_1 \quad \mathbf{H}_6 \quad \mathbf{I} \quad \mathbf{H}_2 \quad \mathbf{H}_1 \quad \mathbf{H}_8 \quad \mathbf{H}_3 \quad \mathbf{H}_3 \quad \mathbf{I} \quad \mathbf{H}_5$

- conisider all possible locations in the sequence
- overcounting of actual (original) strings, correct by combinatorial factor:

$$\binom{M}{n}^{-1} = \frac{n!(M-n)!}{M!}$$

Here n is the number of H_i , i>0 instances in the sequence of M operators

$$Z = \sum_{\{\alpha\}_M} \sum_{\{H_i\}} \frac{(-\beta)^n (M-n)!}{M!} \langle \alpha_0 | H_{i(M)} | \alpha_{M-1} \rangle \cdots \langle \alpha_1 | H_{i(1)} | \alpha_0 \rangle$$

Stochastic Series expansion (SSE): S=1/2 Heisenberg model

Write H as a bond sum for arbitrary lattice

$$H = J \sum_{b=1}^{N_b} \mathbf{S}_{i(b)} \cdot \mathbf{S}_{j(b)},$$

Diagonal (1) and off-diagonal (2) bond operators

$$H_{1,b} = \frac{1}{4} - S_{i(b)}^{z} S_{j(b)}^{z},$$

$$H_{2,b} = \frac{1}{2} (S_{i(b)}^{+} S_{j(b)}^{-} + S_{i(b)}^{-} S_{j(b)}^{+}).$$

$$H = -J \sum_{b=1}^{N_{b}} (H_{1,b} - H_{2,b}) + \frac{JN_{b}}{4}$$

2D square lattice bond and site labels



Four non-zero matrix elements

 $\langle \uparrow_{i(b)} \downarrow_{j(b)} | H_{1,b} | \uparrow_{i(b)} \downarrow_{j(b)} \rangle = \frac{1}{2} \qquad \langle \downarrow_{i(b)} \uparrow_{j(b)} | H_{2,b} | \uparrow_{i(b)} \downarrow_{j(b)} \rangle = \frac{1}{2}$ $\langle \downarrow_{i(b)} \uparrow_{j(b)} | H_{1,b} | \downarrow_{i(b)} \uparrow_{j(b)} \rangle = \frac{1}{2} \qquad \langle \uparrow_{i(b)} \downarrow_{j(b)} | H_{2,b} | \downarrow_{i(b)} \uparrow_{j(b)} \rangle = \frac{1}{2}$

Partition function

$$Z = \sum_{\alpha} \sum_{n=0}^{\infty} (-1)^{n_2} \frac{\beta^n}{n!} \sum_{S_n} \left\langle \alpha \left| \prod_{p=0}^{n-1} H_{a(p),b(p)} \right| \alpha \right\rangle$$

n₂ = number of a(i)=2 (off-diagonal operators) in the sequence

Index sequence: $S_n = [a(0), b(0)], [a(1), b(1)], \dots, [a(n-1), b(n-1)]$

For fixed-length scheme (string length = L now)

$Z = \sum_{\alpha} \sum_{S_L} (-1)^{n_2} \frac{\beta^n (I)}{\beta^n}$	$\left \frac{L-n)!}{L!} \right $	$\alpha \left \prod_{p=0}^{L-1} \right $	$H_{a(p)}$	$b(p) \bigg $	$\left. \alpha \right\rangle \qquad W(\alpha, S_L) = \left(\frac{\beta}{2}\right)^n \frac{(L-r)}{L!}$
Propagated states: $ \alpha(p) $	$ \rangle \propto \prod^{p-1}$	$H_{a(i)},$	$_{b(i)} _{\Omega}$	$\langle \rangle$	W>0 (n ₂ even) for bipartite lattice Frustration leads to sign problem
$i = 1 \ 2 \ 3 \ 4 \ 5 \ 6 \ 7 = 1$ $\sigma(i) = -1 \ +1 \ -1 \ -1 \ +1 \ -1 \ +1 \ +1 \$	$i=0$ 8 $\cdot 1$				$ \land \rightarrow \land $
	р	a(p)	b(p)	s(p)	
	$ \begin{array}{c} $	1	2	4	In a program:
••••••	b 9	2	4	9	
•••••••	D 8	2	6	13	s(p) = operator-index string
	D 7	1	3	6	• $s(p) = 2*b(p) + a(p)-1$
	6	0	0	0	 diagonal: s(p) = even
	5	0	0	0	• off-diagonal: $s(p) = off$
	4	1	2	4	
	3	2	6	13	$\sigma(i) = spin state i = 1$ N
	2	0	0	0	• only one bas to be stored
	1	2	4	9	• Only one has to be stored
	0	1	7	14	

SSE effectively provides a discrete representation of the time continuum

• computational advantage; only integer operations in sampling

Linked vertex storage

The "legs" of a vertex represents the spin states before (below) and after (above) an operator has acted





v X(v)

5

36

47

43

39

35

31

27

23

15

3

l=3

19 28

33

37

4

29

2

l=2



X() = vertex list• operator at $p \rightarrow X(v)$ v=4p+l, l=0,1,2,3

 links to next and previous leg

Spin states between operations are redundant; represented by links

network of linked vertices will be used for loop updates of vertices/operators

Ο

Monte Carlo sampling scheme

Change the configuration; $(\alpha, S_L) \rightarrow (\alpha', S'_L)$

$$P_{\text{accept}} = \min\left[\frac{W(\alpha', S_L)}{W(\alpha, S_L)} \frac{P_{\text{select}}(\alpha', S'_L \to \alpha, S_L)}{P_{\text{select}}(\alpha, S_L \to \alpha', S'_L)}, 1\right]$$

Diagonal update: $[0,0]_p \leftrightarrow [1,b]_p$

$ \alpha(p+1)\rangle$	•	0	0	•	0	•	•	0	\leftrightarrow	•	0	0	•	0	•	•	0
$ \alpha(p)\rangle$	•	0	0	•	0	•	•	0		•	0	0	•	0	•	•	0

Attempt at p=0,...,L-1. Need to know |α(p)>
generate by flipping spins when off-diagonal operator

$$P_{\text{select}}(a = 0 \to a = 1) = 1/N_b, \quad (b \in \{1, \dots, N_b\})$$

 $P_{\text{select}}(a = 1 \to a = 0) = 1$

$$\frac{W(a=1)}{W(a=0)} = \frac{\beta/2}{L-n} \qquad \frac{W(a=0)}{W(a=1)} = \frac{L-n+1}{\beta/2}$$

Acceptance probabilities

$$P_{\text{accept}}([0,0] \to [1,b]) = \min\left[\frac{\beta N_b}{2(L-n)},1\right]$$
$$P_{\text{accept}}([1,b] \to [0,0]) = \min\left[\frac{2(L-n+1)}{\beta N_b},1\right]$$

$$W(\alpha, S_{L}) = \left(\frac{\beta}{2}\right)^{n} \frac{(L-n)!}{L!}$$

$$S_{L} \\ S_{L} \\ S_{L}$$

n is the current power

- n \rightarrow n+1 (a=0 \rightarrow a=1)
- n \rightarrow n-1 (a=1 \rightarrow a=0)

Off-diagonal updates





Local update

Change the type of two operators

- constraints
- inefficient
- cannot change winding numbers

Operator-loop update

- Many spins and operators can be changed simultaneously
- can change winding numbers

Determination of the cut-off L

- adjust during equilibration
- start with arbitrary (small) n

Keep track of number of operators n

- increase L if n is close to current L
- e.g., *L=n+n/3*

Example

- •16×16 system, β =16 \Rightarrow
- evolution of L
- n distribution after equilibration
- truncation is no approximation



Does it work? Compare with exact results

- 4×4 exact diagonalization
- Bethe Ansatz; long chains

Susceptibility of the 4×4 lattice \Rightarrow \approx

- SSE results from 10¹⁰ sweeps
- improved estimator gives smaller error bars at high T (where the number of loops is larger)





⇐ Energy for long 1D chains

- SSE results for 10⁶ sweeps
- Bethe Ansatz ground state E/N
- SSE can achieve the ground state limit (T→0)

Valence bonds and Ground State Projection

The valence bond basis for S=1/2 spins

Valence-bonds between sublattice A, B sites $(i, j) = (|\uparrow_i\downarrow_j\rangle - |\downarrow_i\uparrow_j\rangle)/\sqrt{2}$ Basis states; singlet products

$$|V_r\rangle = \prod_{b=1}^{N/2} (i_{rb}, j_{rb}), \quad r = 1, \dots (N/2)!$$

The valence bond basis is overcomplete and non-orthogonal • expansion of arbitrary singlet state is not unique

 $|\Psi
angle = \sum_{r} f_r |V_r
angle$ (all f_r positive for non-frustrated system)

All valence bond states overlap with each other

 $\langle V_l | V_r \rangle = 2^{N_{\circ} - N/2}$ $N_{\circ} =$ number of loops in overlap graph

Spin correlations from loop structure

$$\frac{\langle V_l | \vec{S}_i \cdot \vec{S}_j | V_r \rangle}{\langle V_l | V_r \rangle} = \begin{cases} \frac{3}{4} (-1)^{x_i - x_j + y_i - y_j} & \text{(i,j in same loop)} \\ 0 & \text{(i,j in different loops)} \end{cases}$$

More complicated matrix elements (e.g., dimer correlations) are also related to the loop structure

K.S.D. Beach and A.W.S., Nucl. Phys. B 750, 142 (2006)





Projector Monte Carlo in the valence-bond basis

Liang, 1991; AWS, Phys. Rev. Lett 95, 207203 (2005)

(-H)ⁿ projects out the ground state from an arbitrary state

$$(-H)^n |\Psi\rangle = (-H)^n \sum_i c_i |i\rangle \to c_0 (-E_0)^n |0\rangle$$

S=1/2 Heisenberg model

$$H = \sum_{\langle i,j \rangle} \vec{S}_i \cdot \vec{S}_j = -\sum_{\langle i,j \rangle} H_{ij}, \quad H_{ij} = \left(\frac{1}{4} - \vec{S}_i \cdot \vec{S}_j\right)$$

Project with string of bond operators

$$\sum_{\{H_{ij}\}} \prod_{p=1}^{n} H_{i(p)j(p)} |\Psi\rangle \to r |0\rangle \qquad (r = \text{irrelevant})$$

Action of bond operators

$$H_{ab}|...(a,b)...(c,d)...\rangle = |...(a,b)...(c,d)...\rangle$$
$$H_{bc}|...(a,b)...(c,d)...\rangle = \frac{1}{2}|...(c,b)...(a,d)...\rangle$$



Simple reconfiguration of bonds (or no change; diagonal)

- no minus signs for A→B bond 'direction' convetion
- sign problem does appear for frustrated systems

Expectation values: $\langle A \rangle = \langle 0 | A | 0 \rangle$

Strings of singlet projectors

$$P_k = \prod_{p=1}^n H_{i_k(p)j_k(p)}, \quad k = 1, \dots, N_b^n \quad (N_b = \text{number of interaction bonds})$$

We have to project bra and ket states

$$\sum_{k} P_{k} |V_{r}\rangle = \sum_{k} W_{kr} |V_{r}(k)\rangle \rightarrow (-E_{0})^{n} c_{0} |0\rangle$$
$$\sum_{g} \langle V_{l} | P_{g}^{*} = \sum_{g} \langle V_{l}(g) | W_{gl} \rightarrow \langle 0 | c_{0} (-E_{0})^{n}$$

6-spin chain example:



$$\langle A \rangle = \frac{\sum_{g,k} \langle V_l | P_g^* A P_k | V_r \rangle}{\sum_{g,k} \langle V_l | P_g^* P_k | V_r \rangle}$$

$$= \frac{\sum_{g,k} W_{gl} W_{kr} \langle V_l(g) | A | V_r(k) \rangle}{\sum_{g,k} W_{gl} W_{kr} \langle V_l(g) | V_r(k) \rangle}$$

- Monte Carlo sampling of operator strings
- Estimators based on transition graphs

More efficient ground state QMC algorithm → larger lattices Loop updates in the valence-bond basis

AWS and H. G. Evertz, PRB 2010

Put the spins back in a way compatible with the valence bonds

 $(a_i, b_i) = (\uparrow_i \downarrow_j - \downarrow_i \uparrow_j) / \sqrt{2}$

and sample in a combined space of spins and bonds





Loop updates similar to those in finite-T methods (world-line and stochastic series expansion methods)

- good valence-bond trial wave functions can be used
- larger systems accessible
- sample spins, but measure using the valence bonds

T>0 and T=0 algorithms side-by-side

Finite-temperature QMC (world lines, SSE,...)



periodic time boundary conditions

Computer implementations similar

Ground state projection





open boundaries capped by valence bonds (2-spin singlets) [AWS, HG Evertz, 2010]

Trial state can conserve relevant ground state quantum numbers (S=0, k=0,...)

Starting point: S=1/2 antiferromagnetic Heisenberg model



Sublattice magnetization



 $\vec{m}_s = \frac{1}{N} \sum_{i=1}^{N} \phi_i \vec{S}_i, \quad \phi_i = (-1)^{x_i + y_i}$ (2D square lattice)

Long-range order: $< m_s^2 > > 0$ for $N \rightarrow \infty$

Quantum Monte Carlo

- finite-size calculations
- no approximations
- extrapolation to infinite size

Reger & Young 1988

 $m_s = 0.30(2)$

 $\approx 60~\%$ of classical value

<u>AWS & HG Evertz 2010</u> $m_s = 0.30743(1)$

L×L lattices up to 256×256, T=0 0.00002 0.13 0.00000 C-fit C(*L*/2,*L*/2), *M*² C(*L*/2,*L*/2), *M*² -0.000020.02 0.040.00.10 C(L/2,L/2)0.03 0.05 0.01 0.02 0.04 0.06 0

1/L