

The loop update is an example of a cluster algorithm. Detour:

Cluster algorithm for the Ising model

Define bond index corresponding to pair of interacting spins

bond $b = 1, 2, \dots, N_b$, interacting spins $\sigma_{i(b)}, \sigma_{j(b)}$

Number of bonds $N_b = dN$ for a d -dimensional cubic lattice

Write the energy of the Ising ferromagnet as

$$E = -|J| \sum_{b=1}^{N_b} [\sigma_{i(b)} \sigma_{j(b)} + 1] = - \sum_{b=1}^{N_b} E_b$$

Write the partition function as

$$Z = \sum_{\sigma} e^{-E(\sigma)/T} = \sum_{\sigma} \prod_{b=1}^{N_b} e^{E_b/T} = \sum_{\sigma} \prod_{b=1}^{N_b} [1 + (e^{E_b/T} - 1)]$$

Define bond functions with arguments 0,1 (bond variable):

$$\begin{aligned} F_b(0) &= 1 \\ F_b(1) &= e^{E_b/T} - 1 \end{aligned}$$

$$Z = \sum_{\sigma} \prod_{b=1}^{N_b} [F_b(0) + F_b(1)]$$

Introduce **bond variables**

$$\tau_b = 0, 1, \quad \tau = \{\tau_1, \tau_2, \dots, \tau_{N_b}\}$$

Partition function can be written as **sum over spins and bonds**

$$Z = \sum_{\sigma} \prod_{b=1}^{N_b} [F_b(0) + F_b(1)] = \sum_{\sigma} \sum_{\tau} \prod_{b=1}^{N_b} F_b(\tau_b)$$

The functions F_b depend on the spins:

$$F_b(0) = 1$$

$$F_b(1) = e^{E_b/T} - 1 = \begin{cases} e^{2|J|/T} - 1, & \text{if } \sigma_{i(b)} = \sigma_{j(b)} \\ 0, & \text{if } \sigma_{i(b)} \neq \sigma_{j(b)} \end{cases}$$

$\tau_b = 1$ allowed only between parallel spins

Probabilities: For everything else fixed, probability for a given b

$$P(\tau_b) = \frac{F(\tau_b)}{F(0) + F(1)} = \frac{F(\tau_b)}{e^{2|J|/T}}$$

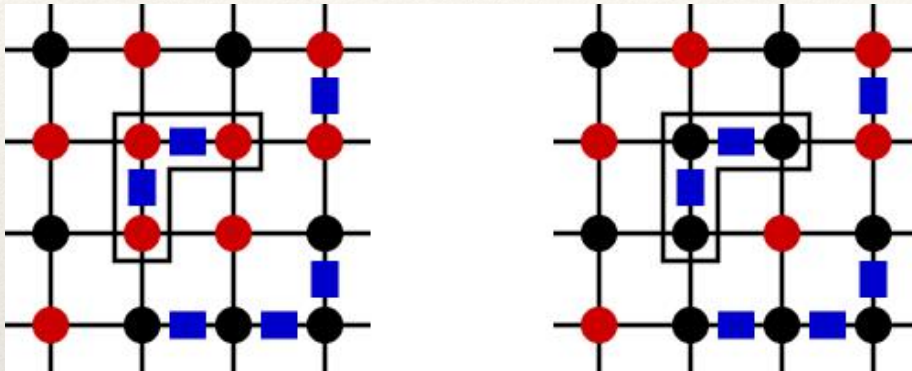
If **parallel spins on bond b** , probabilities for the bond variable

$$P(\tau_b = 0) = e^{-2|J|/T}, \quad P(\tau_b = 1) = 1 - e^{-2|J|/T}$$

If **anti-parallel spins on bond b**

$$P(\tau_b = 0) = 1, \quad P(\tau_b = 1) = 0$$

For a fixed bond configuration, **spins forming clusters** (spins connected by “filled” bonds) **can be flipped** and then give a configuration (term) with the same weight in Z ($F_b=1$ for all bonds between clusters, F_b unchanged inside cluster).



$$N(\tau_b = 1) = \text{No. of filled bonds}$$

$$W = (e^{2|J|/T} - 1)^{N(\tau_b=1)}$$

(unchanged after flip)

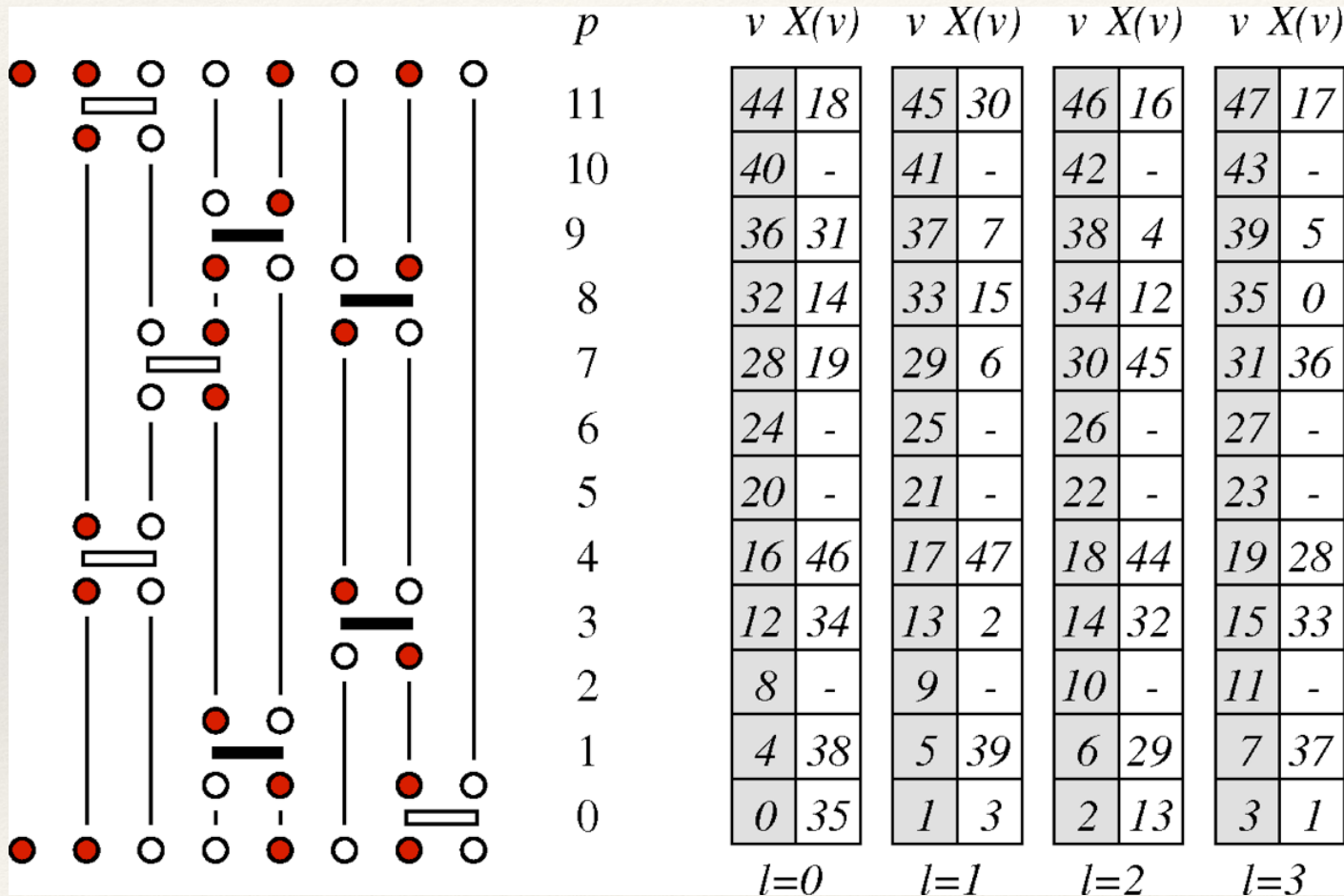
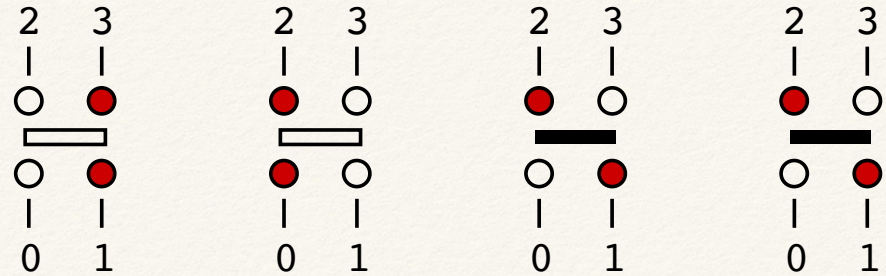
Spins not connected to any filled bonds are single-spin clusters

Swendsen-Wang algorithm

- Start from spin configuration
- Generate bond configuration
- Identify clusters of spins connected by bonds
- Flip each cluster with probability 1/2
- Generate new bonds with the current spins, etc

SSE: Linked vertex storage for loop update

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



$X()$ = 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

Pseudocode: Sweep of loop updates

constructing all loops, flip probability 1/2

```
do  $v_0 = 0$  to  $4L - 1$  step 2
  if ( $X(v_0) < 0$ ) cycle
     $v = v_0$ 
    if (random[0 - 1] <  $\frac{1}{2}$ ) then
      traverse the loop; for all  $v$  in loop, set  $X(v) = -1$ 
    else
      traverse the loop; for all  $v$  in loop, set  $X(v) = -2$ 
      flip the operators in the loop
    endif
  enddo
```

- visited vertices are no longer needed and we set them to a negative value -1 or -2, to indicate that the loop has been visited (-1) or visited and flipped (-2)

construct and flip a loop

```
 $v = v_0$ 
do
   $X(v) = -2$ 
   $p = v/4$ ;  $s(p) = \text{flipbit}(s(p), 0)$ 
   $v' = \text{flipbit}(v, 0)$ 
   $v = X(v')$ ;  $X(v') = -2$ 
  if ( $v = v_0$ ) exit
enddo
```

- p is the location of the operator in the original length- L list of operators
- by flipping bit 0 of $s(p)$, the operator changes from diagonal to off-diagonal, or vice versa
- moving on the vertex to the adjacent spin is also done with a bit flip

We also have to modify the stored spin state after the loop update

- we can use the information in $V_{\text{first}}()$ and $X()$ to determine spins to be flipped
- spins with no operators, $V_{\text{first}}(i)=-1$, flipped with probability $1/2$

```
do  $i = 1$  to  $N$ 
   $v = V_{\text{first}}(i)$ 
  if ( $v = -1$ ) then
    if (random[0-1] < 1/2)  $\sigma(i) = -\sigma(i)$ 
  else
    if ( $X(v) = -2$ )  $\sigma(i) = -\sigma(i)$ 
  endif
enddo
```

$\mathbf{v=V_{first}(i)}$ is the location of the first vertex leg on site i

- flip the spin if $X(v)=-2$
- (do not flip it if $X(v)=-1$)
- no operation on i if $V_{\text{first}}(i)=-1$; then it is flipped with probability $1/2$

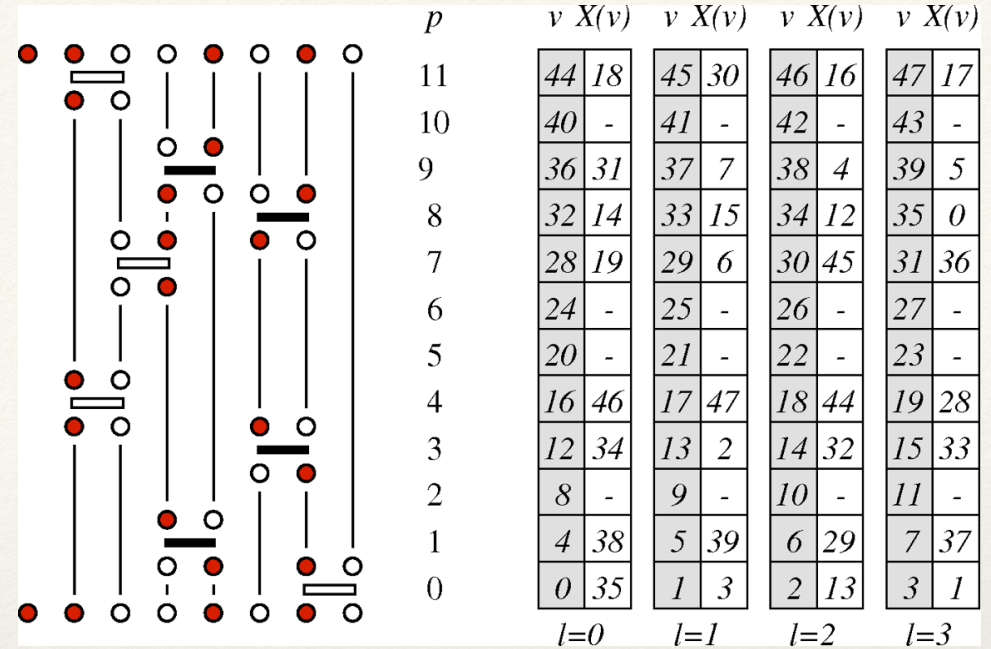
Constructing the linked vertex list

Traverse operator list $s(p)$, $p=0, \dots, L-1$

- vertex legs $v=4p, 4p+1, 4p+2, 4p+3$

Use arrays to keep track of the first and last (previous) vertex leg on a given spin

- $V_{\text{first}}(i)$ = location v of first leg on site i
- $V_{\text{last}}(i)$ = location v of last (currently) leg
- these are used to create the links
- initialize all elements to -1



$V_{\text{first}}(:) = -1; V_{\text{last}}(:) = -1$

do $p = 0$ **to** $L - 1$

if $(s(p) = 0)$ **cycle**

$v_0 = 4p; b = s(p)/2; s_1 = i(b); s_2 = j(b)$

$v_1 = V_{\text{last}}(s_1); v_2 = V_{\text{last}}(s_2)$

if $(v_1 \neq -1)$ **then** $X(v_1) = v_0; X(v_0) = v_1$ **else** $V_{\text{first}}(s_1) = v_0$ **endif**

if $(v_2 \neq -1)$ **then** $X(v_2) = v_0; X(v_0) = v_2$ **else** $V_{\text{first}}(s_2) = v_0 + 1$ **endif**

$V_{\text{last}}(s_1) = v_0 + 2; V_{\text{last}}(s_2) = v_0 + 3$

enddo

creating the last links across the “time” boundary

do $i = 1$ **to** N

$f = V_{\text{first}}(i)$

if $(f \neq -1)$ **then** $l = V_{\text{last}}(i); X(f) = l; X(l) = f$ **endif**

enddo

Determination of the cut-off L

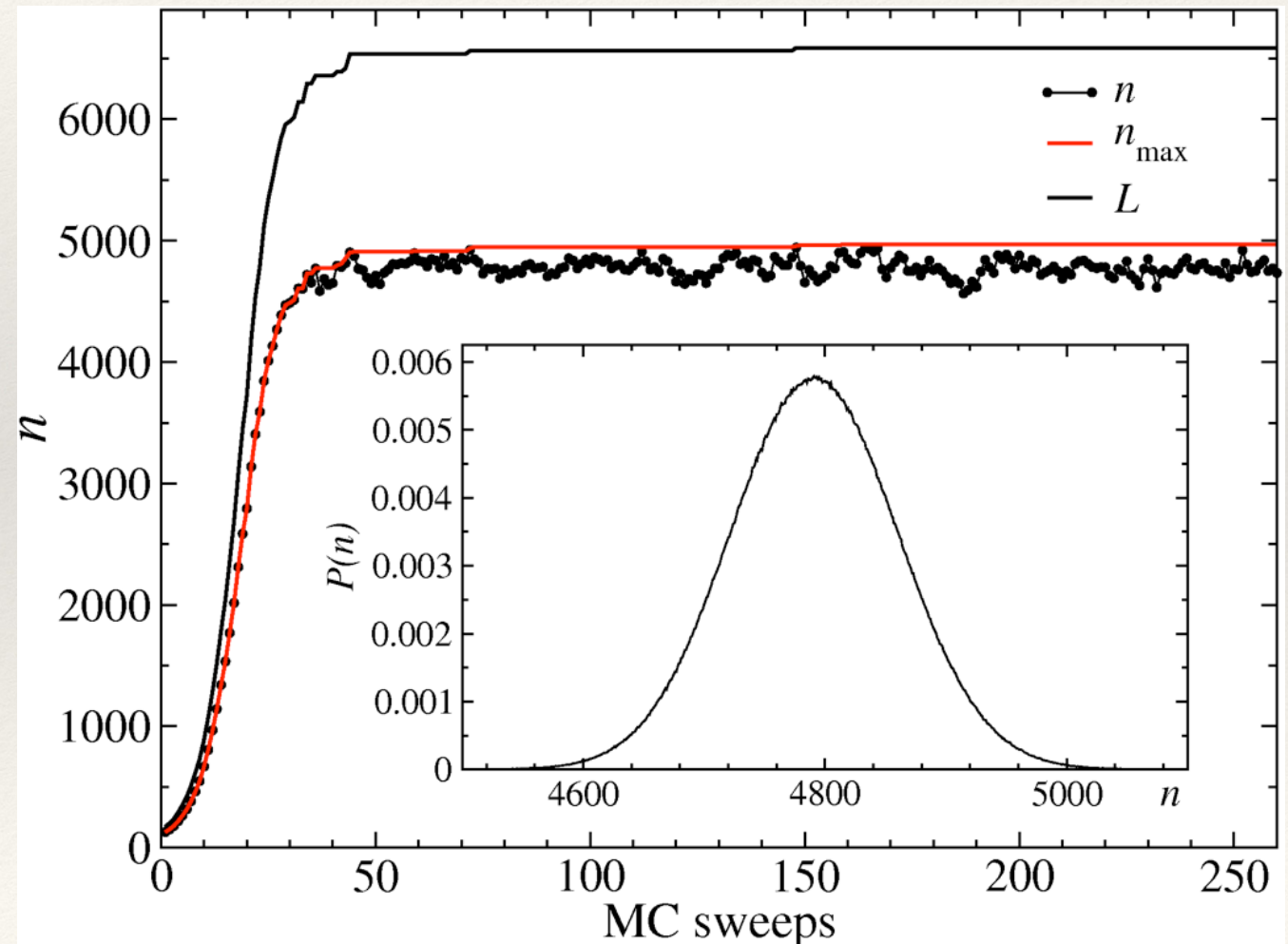
- adjust during equilibration
- start with arbitrary (small) n

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, $\beta = 16 \Rightarrow$
- evolution of L
- n distribution after equilibration
- truncation is no approximation



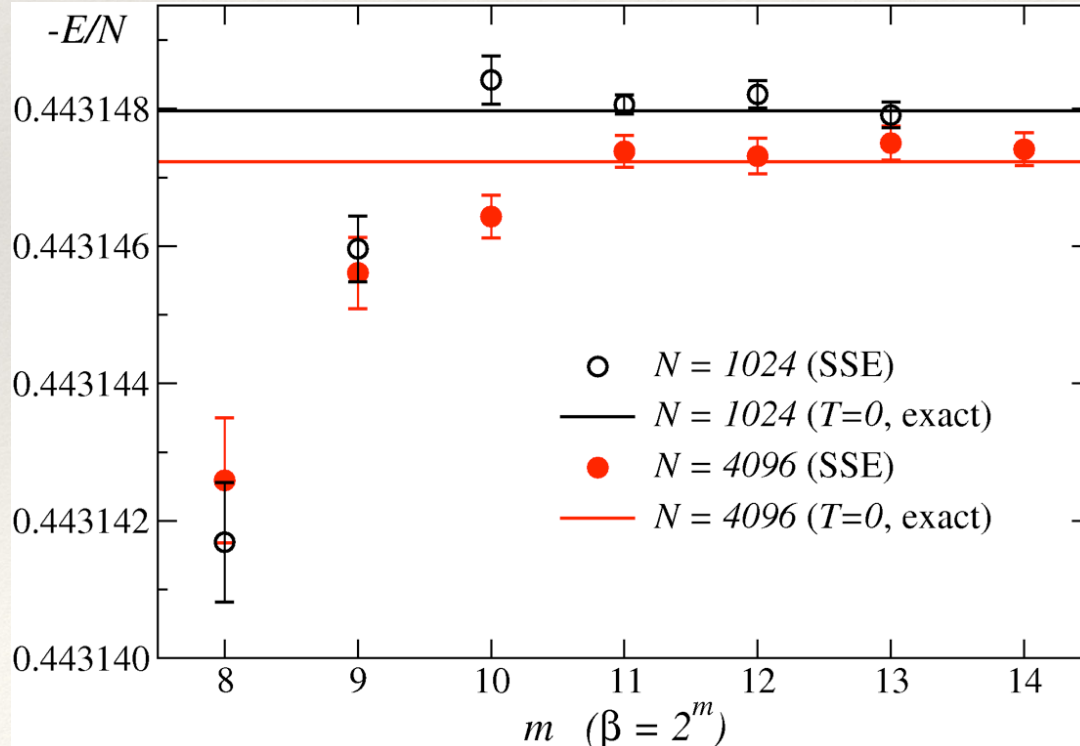
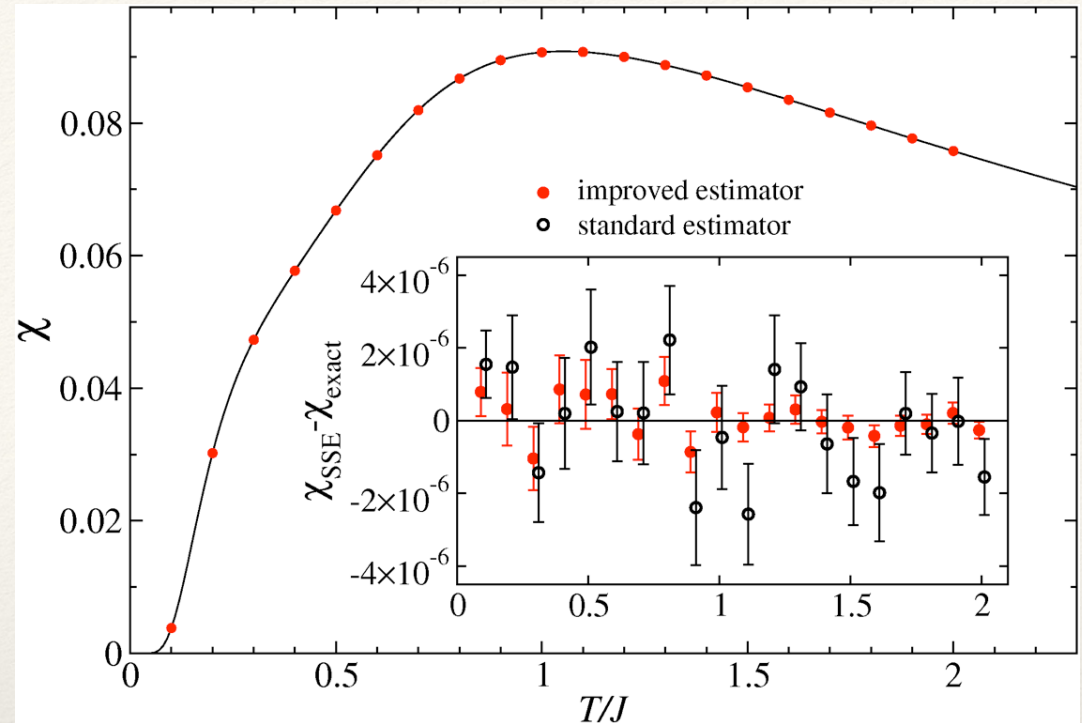
Does it work?

Compare with exact results

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

Susceptibility of the 4×4 lattice ⇒

- SSE results from 10^{10} sweeps
- improved estimator gives smaller error bars at high T (where the number of loops is larger)



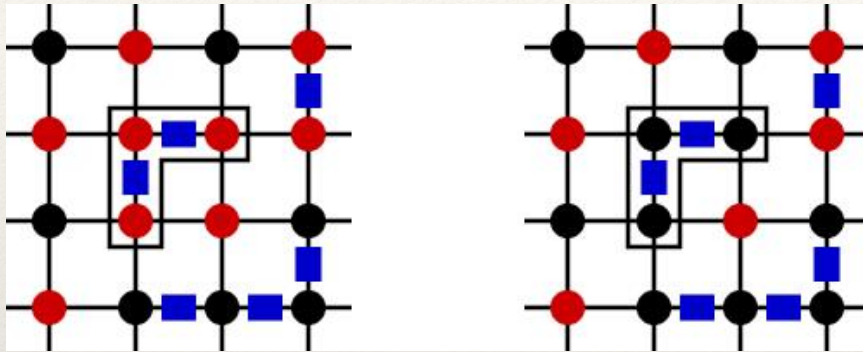
⇐ Energy for long 1D chains

- SSE results for 10^6 sweeps
- Bethe Ansatz ground state E/N
- SSE can achieve the ground state limit ($T \rightarrow 0$)

Improved Estimators

SSE with loop updates is an example of a cluster algorithms
- we can utilize improved estimators for many observables

Classical example: Swendsen-Wang Ising cluster algorithm



$N(\tau_b = 1) =$ No. of filled bonds

$$W = (e^{2|J|/T} - 1)^{N(\tau_b=1)}$$

(unchanged after flip)

Write magnetization as sum over clusters of size n_C , sign s_C :

$$M = \sum_{i=1}^N \sigma_i = \sum_{C=1}^{N_{\text{clus}}} \sum_{i \in C} \sigma_i = \sum_{C=1}^{N_{\text{clus}}} s_C n_C \quad \langle M^2 \rangle = \sum_{C=1}^{N_{\text{clus}}} \sum_{C'=1}^{N_{\text{clus}}} \langle n_C n_{C'} s_C s_{C'} \rangle$$

All cluster orientations (signs) have same weight

- average over all $2^{N_{\text{clus}}}$ orientations \rightarrow

$$\langle M^2 \rangle = \sum_{C=1}^{N_{\text{clus}}} \langle n_C^2 \rangle$$

This is the improved estimator of $\langle M^2 \rangle$
- only depends on cluster structure

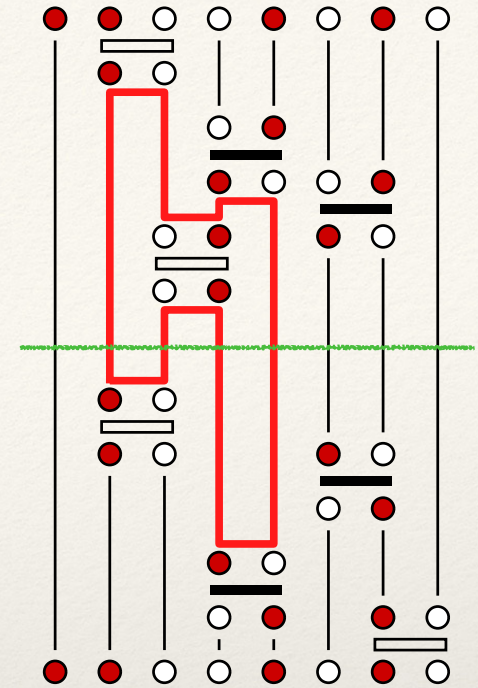
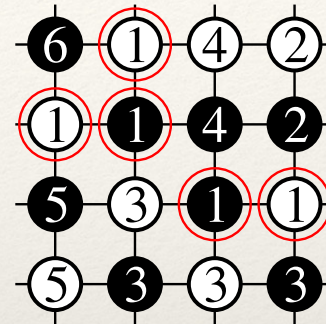
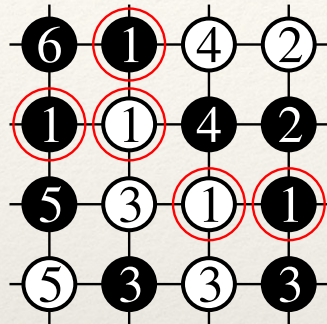
Improved estimators in SSE

Consider a given slice (propagated state) of an SSE configuration

- label the sites according to the loops passing through \rightarrow clusters

In given loop

All spins on given sub-lattice A or B are same, different on A, B



Staggered magnetization on a cluster is 1/2 of the size of the cluster

- changes sign when loop flipped
- similar to magnetization in SW algorithm

$$\langle M_{z,\text{stagg}}^2 \rangle = \frac{1}{4} \sum_{C=1}^{N_{\text{clus}}} \langle n_C^2 \rangle$$

The uniform magnetization requires the staggered phases

$$\chi = \frac{\beta}{4N} \left\langle \sum_{j=1}^C \left(\sum_{i=1}^{n_j} \phi_i \right)^2 \right\rangle \quad \phi_i = \begin{cases} +1 & i \text{ on A site} \\ -1 & i \text{ on B site} \end{cases}$$

Transverse-field Ising model [PRE 68, 056701]

$$H = \sum_{i,j} J_{ij} \sigma_i^z \sigma_j^z - h \sum_i \sigma_i^x$$

Arbitrary interactions (incl. random, long-range,...)

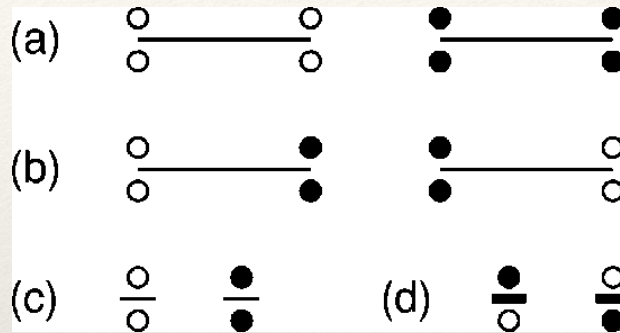
Elementary operators for the SSE strings:

$$H_{0,0} = 1,$$

$$H_{i,0} = h(\sigma_i^+ + \sigma_i^-), \quad i > 0,$$

$$H_{i,i} = h, \quad i > 0,$$

$$H_{i,j} = |J_{ij}| - J_{ij} \sigma_i^z \sigma_j^z, \quad i, j > 0, i \neq j$$



$$H = - \sum_{i=1}^N \sum_{j=0}^N H_{i,j}$$

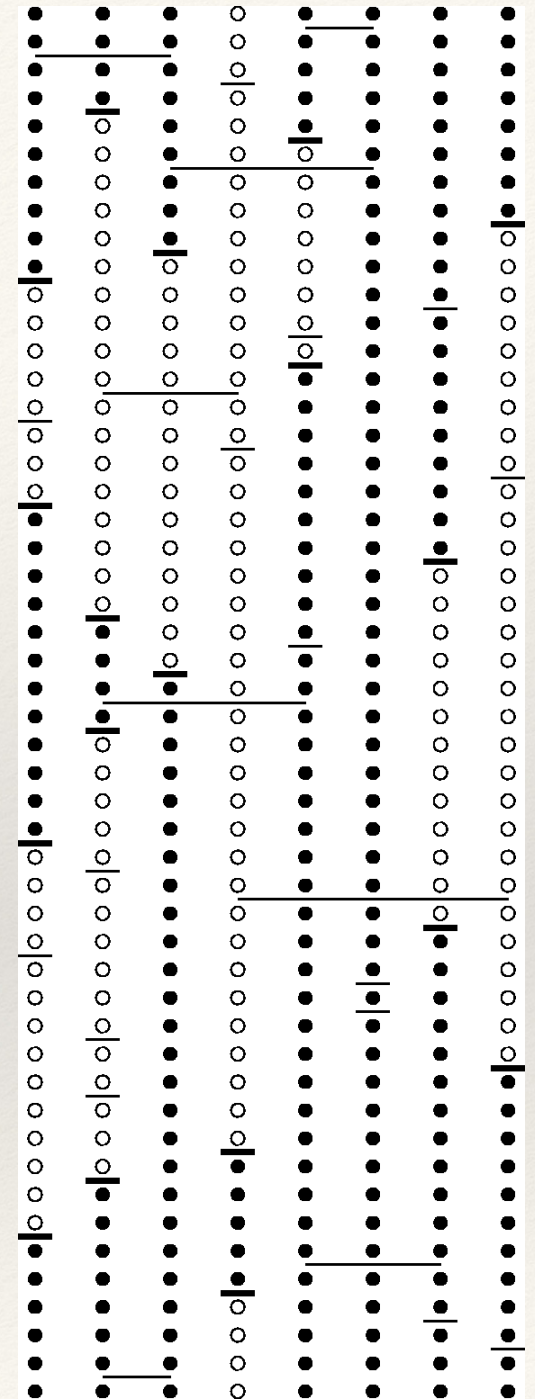
$$S_n = [i(1), j(1)], \dots, [i(n), j(n)]$$

$$Z = \frac{1}{L!} \sum_{\alpha} \sum_{S_L} \beta^n (L-n)! \langle \alpha | \prod_{l=1}^L H_{i(l), j(l)} | \alpha \rangle$$

Possible local updates:

$$[0,0]_p \leftrightarrow [i,j]_p, \quad i, j \neq 0, \quad (\text{diagonal})$$

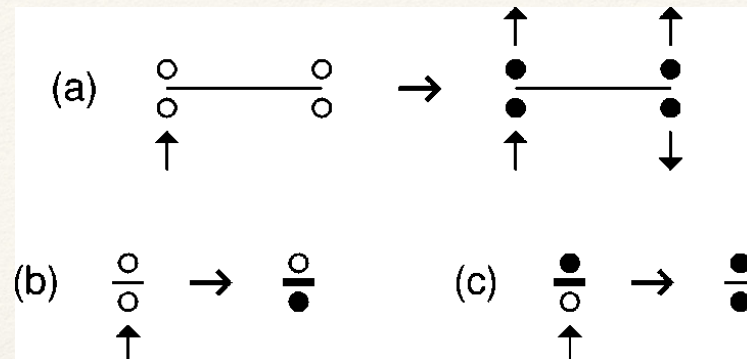
$$[i,i]_{p_1} [i,i]_{p_2} \leftrightarrow [i,0]_{p_1} [i,0]_{p_2}, \quad i \neq 0 \quad (\text{off-diagonal})$$



Cluster update for Transverse-field Ising models

Start at arbitrary in-leg and apply weight-preserving operator changes

- growing “to do” list (stack) from each out-leg
- until stack is empty
- divides the entire system into clusters
- similar to classical S-W algorithm



Long-range interactions

Goal: avoid summations over interactions J_{ij}

- we can first assume that all diagonal operators are allowed, then

$$P([0,0] \rightarrow [i,j]) = \frac{\beta \left(Nh + 2 \sum_{ij} |J_{ij}| \right)}{L - n + \beta \left(Nh + 2 \sum_{ij} |J_{ij}| \right)} \quad P([i,j] \rightarrow [0,0]) = \frac{L - n + 1}{L - n + 1 + \beta \left(Nh + 2 \sum_{ij} |J_{ij}| \right)}$$

where $[i,j]$ in $[0,0] \rightarrow [i,j]$ is a still undetermined interaction

- the actual $[i,j]$ is chosen in a second step using cumulative probabilities:

$$P_c(k) = \frac{\sum_{i=1}^k P(i)}{\sum_{i=1}^N P(i)}$$

N is the total number of diagonal ops

Choose op # k using bisection in the table $P_c()$

- time scales as $\ln(N)$

Cluster update remains the same for any kind of interaction

- total time for updating sweep scales as $\beta N \ln(N)$ with long-range interactions