

**New Development of Numerical Simulations in Low-Dimensional Quantum Systems:
From Density Matrix Renormalization Group to Tensor Network Formulations
October 27-29, 2010, Yukawa Institute for Theoretical Physics, Kyoto University**

Matrix-Product states: Properties and Extensions

Anders W Sandvik
Boston University

Collaborators

Chen Liu (Boston University)
Ying-Jer Kao (National Taiwan University, Taipei)
Yu-Cheng Su (National Taiwan University, Taipei)
Ling Wang (University of Vienna)

PHYSICAL REVIEW B 82, 060410R (2010)



Outline

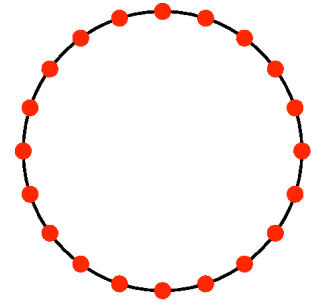
- **Variational optimization of periodic MPS**
- **Mechanism of symmetry breaking with MPS**
 - ▶ 1-d periodic transverse-field Ising model
 - ▶ critical form of the magnetization curve (finite N , $N=\infty$)
 - ▶ limitations of finite computer precision(?)
- **Criticality in 2D iPEPS (transverse-field Ising)**
- **MPS with variational Monte Carlo** (time permitting)

Matrix product states (MPS)

Consider a periodic chain of $S=1/2$ spins

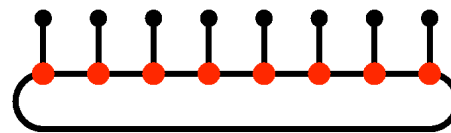
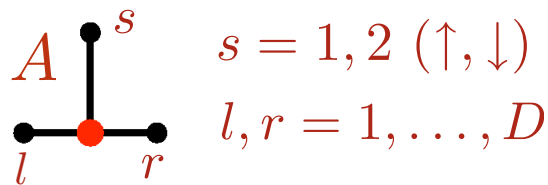
$$|\Psi\rangle = \sum_{\{s_i\}} W(s_1, s_2, \dots, s_N) |s_1, s_2, \dots, s_N\rangle, \quad s_i = \uparrow, \downarrow$$

$$W(s_1, s_2, \dots, s_N) = \text{Tr}[A(s_1)A(s_2)\cdots A(s_N)]$$



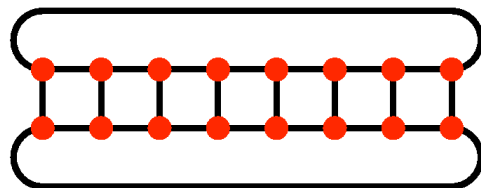
- MPSs can be implicitly generated by DMRG (Ostlund & Romer, 1995)
- Can be used independently of DMRG as a class of variational 1-d states

Graphical representation of a_{lr}^s and MPSs

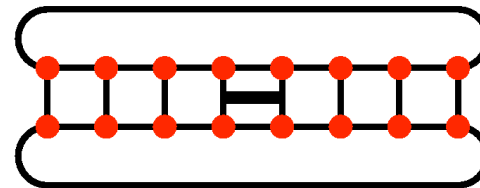


$$A(s) = \begin{pmatrix} a_{11}^s & \cdots & a_{1D}^s \\ a_{21}^s & \cdots & a_{2D}^s \\ \cdots & \cdots & \cdots \\ a_{D1}^s & \cdots & a_{DD}^s \end{pmatrix}$$

Normalization $\langle \Psi | \Psi \rangle$



Expectation value $\langle \Psi | S_i^a S_{i+1}^b | \Psi \rangle$



Can be easily evaluated; scaling for periodic chain: standard way costs ND^5

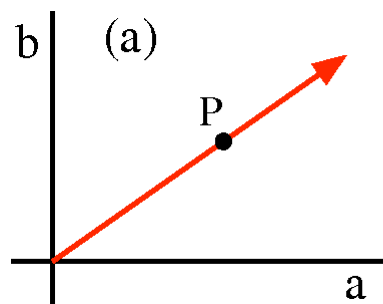
- Pippin, White, Evertz (PRB 2010); good approximation (SVD) with ND^3
- Monte Carlo sampling (Sandvik & Vidal, PRL 2007); ND^3

How to optimize the matrices in MPS calculations

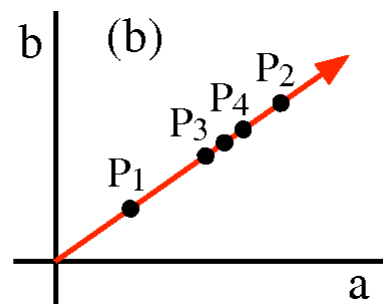
- Local energy minimization, “sweep” through the lattice (Verstraete et al., ...)
- Imaginary-time evolution (projecting out the ground state) (Vidal, ...)

Minimize the energy with maintained translational invariance?

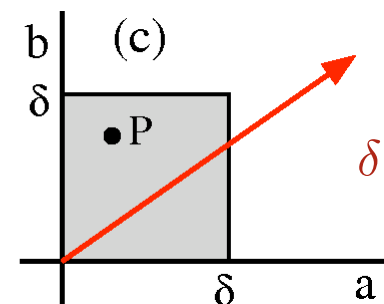
Stochastic Optimization (using first derivatives)



Steepest decent



Line minimization



Stochastic method

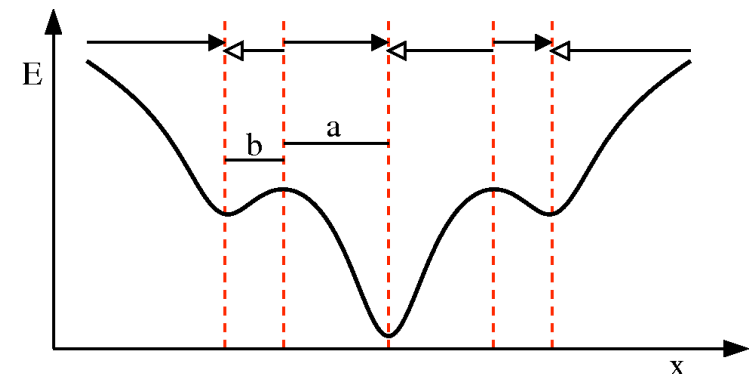
$$\delta = \frac{\delta_0}{k^\alpha}, \quad k = 1, 2, \dots$$

The **stochastic method** is guaranteed to reach the global minimum if:

- “cooled” sufficiently slowly
- if all local minima on “funnel walls”: $b < a$

Seems to work well for MPS optimization

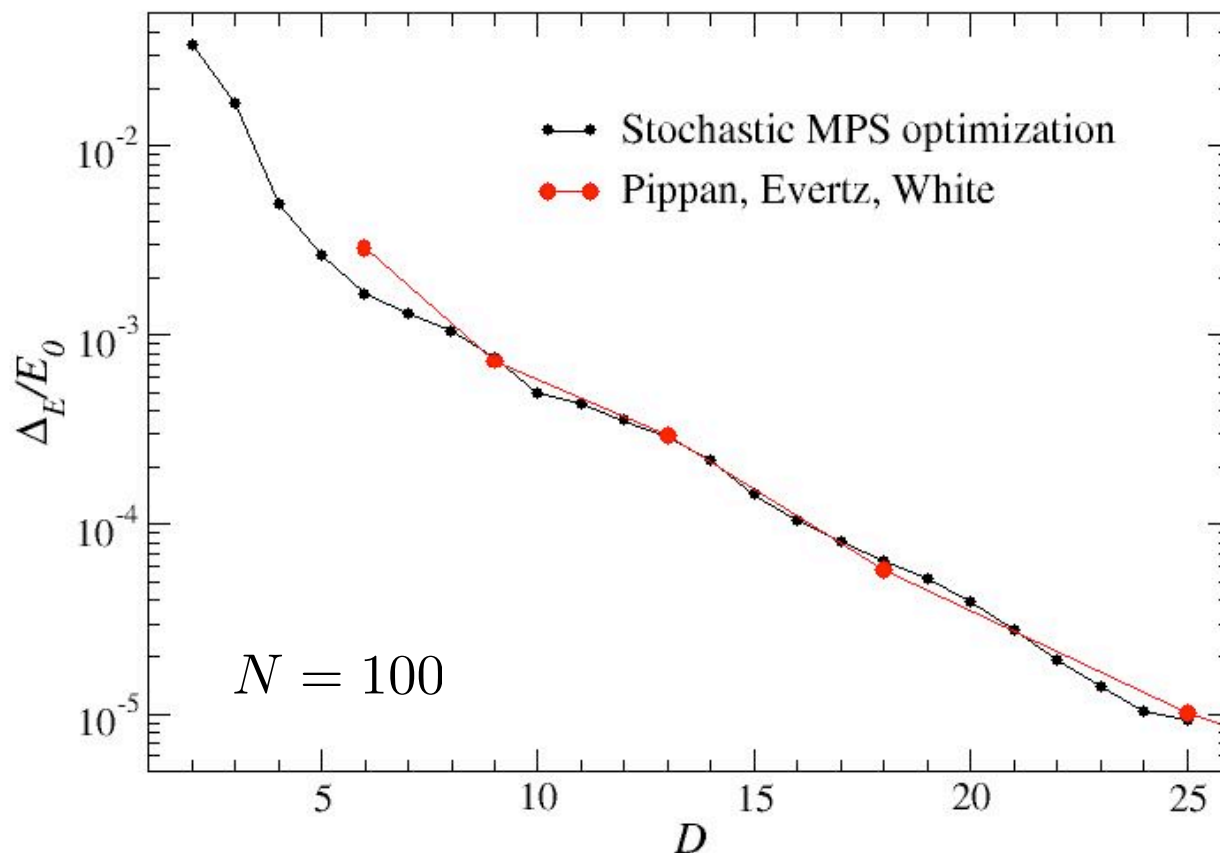
- Starting from random matrices or ones optimized for smaller D
- Steepest decent can be faster at final stages
- But much slower than conventional methods



Test: Antiferromagnetic Heisenberg chain

$$H = \sum_{i=1}^N \mathbf{S}_i \cdot \mathbf{S}_{i+1} = \sum_{i=1}^N [S_i^z S_{i+1}^z + \frac{1}{2}(S_i^+ S_{i+1}^- + S_i^- S_{i+1}^+)]$$

Comparison with N=100 results by: Pippin, White, Evertz (PRB 2010)



Good results, but the method is very slow

Infinite chain MPS

Exactly as in classical transfer-matrix method;

- keep only largest eigenvalue of P when $N \rightarrow \infty$
- Imaginary-time evolution (ground state projection) or DMRG-type optimization can be applied (Vidal, Cirac, McCulloch,...)

$$(a) \quad \langle \Psi | \Psi \rangle = \begin{array}{c} \text{---} \text{---} \text{---} \text{---} \text{---} \\ | \quad | \quad | \quad \dots \quad | \quad | \\ \text{---} \text{---} \text{---} \text{---} \text{---} \end{array} = \begin{array}{c} \text{---} \text{---} \text{---} \text{---} \text{---} \\ | \quad | \quad | \quad \dots \quad | \quad | \\ \text{---} \text{---} \text{---} \text{---} \end{array} = \text{Tr}\{P^N\}$$

$$(b) \quad \begin{array}{c} c \text{---} d \\ | \\ a \text{---} b \end{array} = a+(c-1)D \text{---} \bullet \text{---} b+(d-1)D = A_{ab}(\uparrow)A_{cd}^*(\uparrow) + A_{ab}(\downarrow)A_{cd}^*(\downarrow) = P$$

For some operator M (single-site, e.g., magnetization)

$$\langle M \rangle = \frac{\text{Tr}\{MP^{N-1}\}}{\text{Tr}\{P^N\}} \rightarrow \frac{1}{\lambda_1} \sum_{i,j} v_{1i}^* v_{1j} M_{ij}$$

Question: How is symmetry breaking manifested in MPS?

- for finite N and $N \rightarrow \infty$

Test: transverse-field Ising model

- true critical magnetization exponent $\beta=1/8$
- how does this exponent emerge?
- what is the $h \rightarrow h_c$ behavior for finite D ?

$$H = - \sum_{i=1}^N \sigma_i^z \sigma_{i+1}^z - \frac{h}{2} \sum_{i=1}^N (\sigma_i^+ + \sigma_i^-)$$

Stochastic optimization

- Energy derivatives involve summing N different contributions
 - ▶ time-consuming for $N \rightarrow \infty$

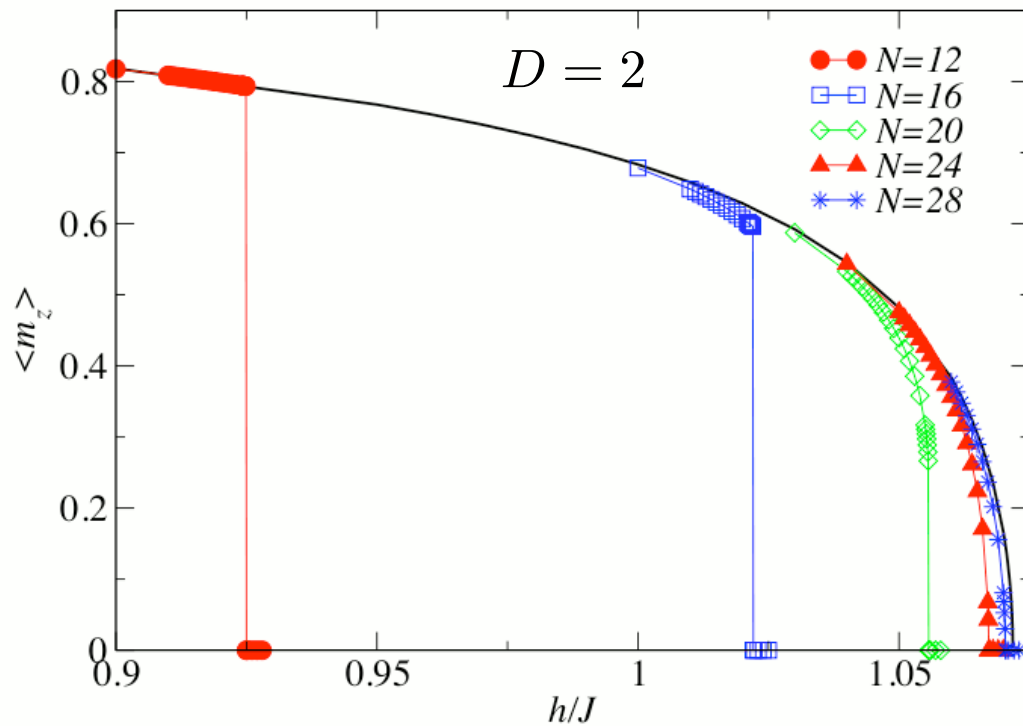
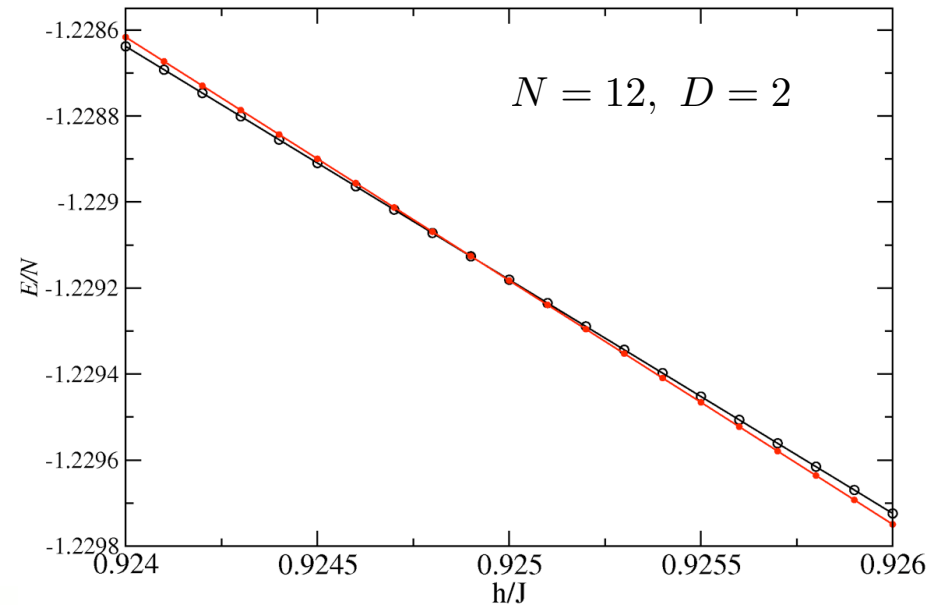
Optimize in a trivial (slow) way for $N=\infty$

- Propose random changes in the matrix elements
 - ▶ accept if and only if the energy improves
- easy to do in **quadruple precision** (but very slow)

Symmetry breaking for finite N

First-order transition (D fixed)

- discontinuity decreases with increasing N
- continuous for $N \rightarrow \infty$
- two E minimums
 - symmetric and symmetry-broken states
- “level” crossing



Behavior versus D

- for given N, $h_c(D) \rightarrow 0$
- no symmetry-breaking for $N < \infty, D = \infty$

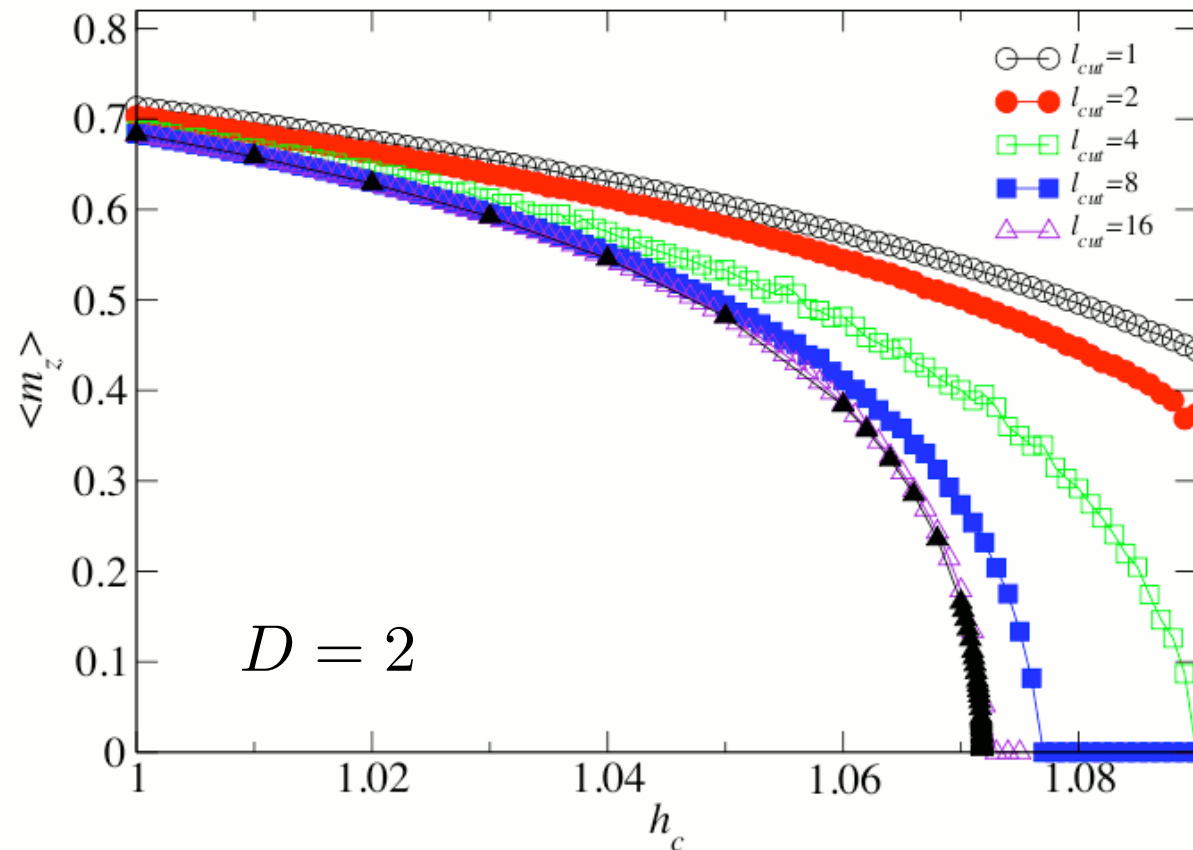
Infinite chain MPS - optimization using derivatives

The derivative of the energy with respect to a matrix element is of the form

$$\frac{\partial E}{\partial a_{ij}^\sigma} = C_{ij}^\sigma + \sum_{l=1}^{N-2} D_{ij}^\sigma(l) \quad D(l) \sim \text{Tr}\{X B^l X B^{N-2-l}\}$$

$D(l)$ is a correlation function; $D(l) \rightarrow 0$ when $l \rightarrow \infty$

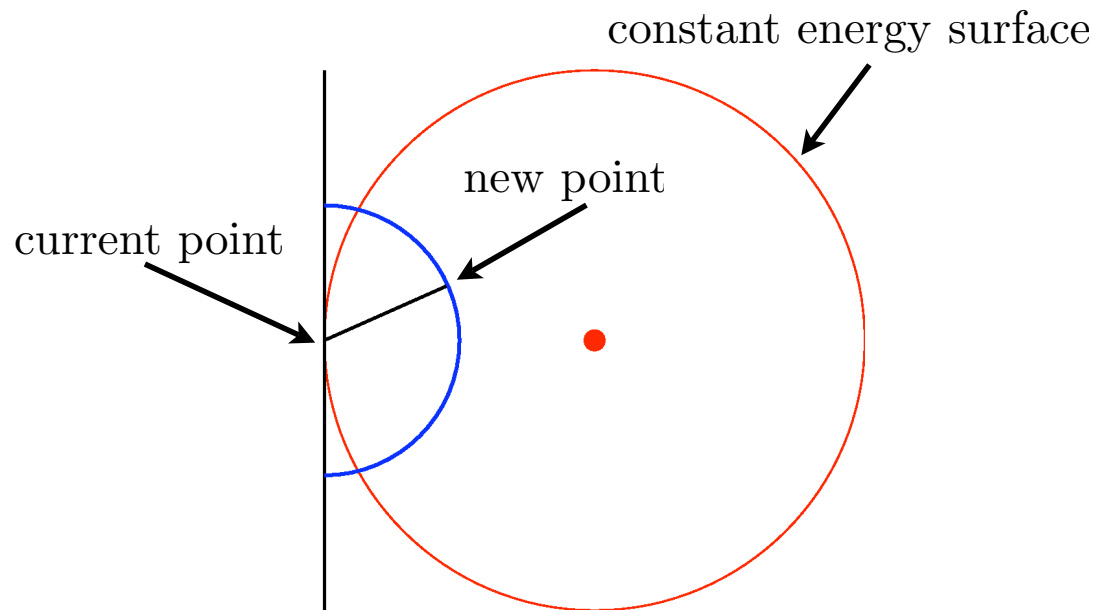
- impose cut-off l_{cut} in optimization for $N=\infty$; dependence on l_{cut}



$N=\infty$: Optimization using trivial random updates

Does not require derivatives

- propose random changes in all parameters; maximum change= δ
- accept only if the energy decreases
- for $\delta \rightarrow 0$ the acceptance rate should be 50%
- adjust δ to give (e.g.) 10% acceptance rate



- find largest eigenvalue of P using

$$P^m v = \lambda_1^m v, \quad m \rightarrow \infty$$

- efficient with $m=1,2,4,8,\dots \rightarrow P, P^2, P^4, P^8, \dots$
- numerically stable
- easy to go to high precision (quadruple, 128 bit)

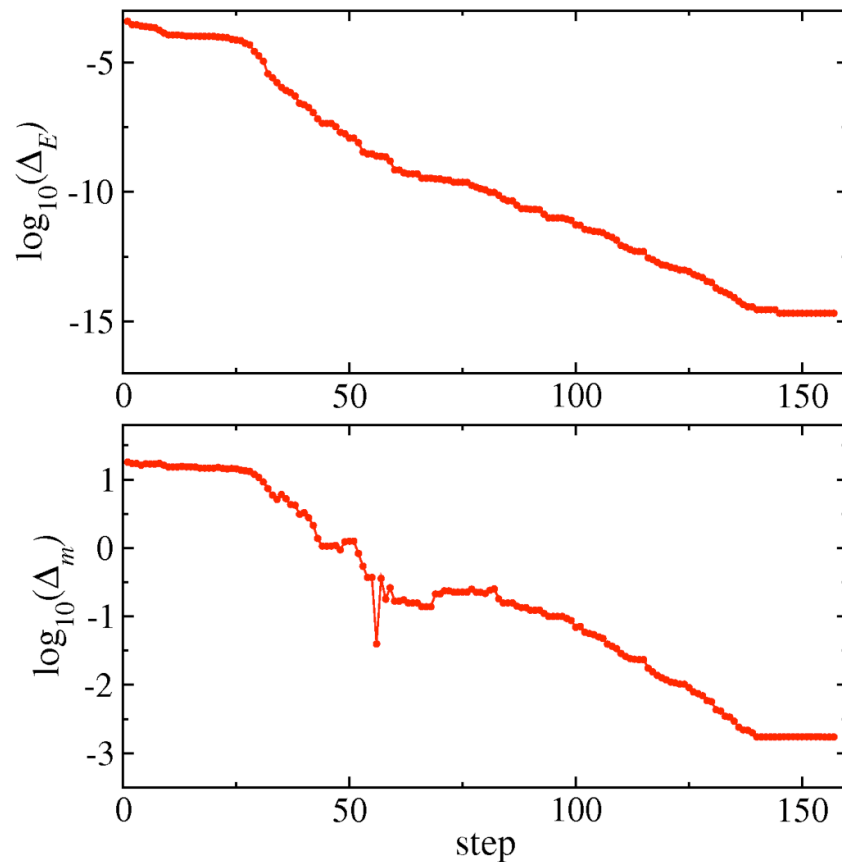
Example: $D=4$, $h=1.01432$

- 10^4 update attempts per “step”
- $\delta \rightarrow \delta/1.1$ after each step if $<10\%$ accepted updates
- **stage 1:** double precision, **stage 2:** quadruple precision
- $E = -1.282445246576107642\dots$, $M_z = 0.0318141670\dots$ (quad precision)

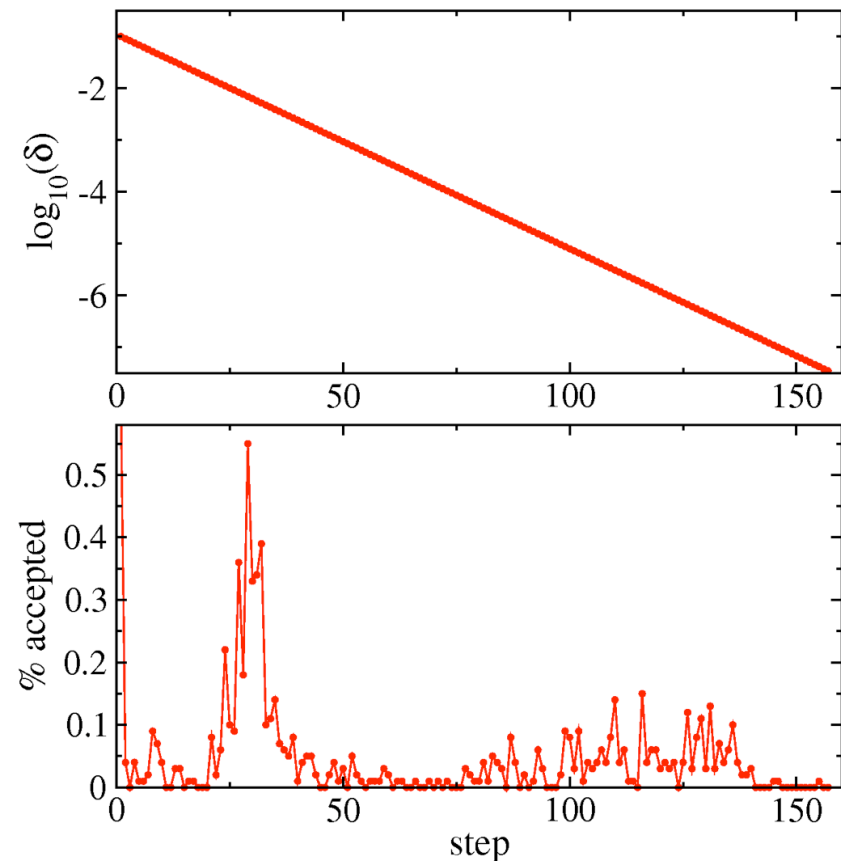
Errors relative to converged results (for given D)

$$\Delta_E = (E - E_{\text{conv}}) / E_{\text{conv}}$$

$$\Delta_m = (m - m_{\text{conv}}) / m_{\text{conv}}$$



evolution of δ and acceptance rate



E/N

$$m^z = \frac{1}{N} \sum_{i=1}^N S_i^z$$

δ = max change in matrix elements

-1.274624764007379568601740
 -1.274624764007380869060010
 -1.274624764007382167234777
 -1.274624764007383018499448
 -1.274624764007392878608852
 -1.274624764007400992225410
 -1.274624764007405559290561
 -1.274624764007410209949167
 -1.274624764007416468671250

0.0107523120526
 0.0107537678231
 0.0107581552419
 0.0107499540804
 0.0107457713217
 0.0107333530705
 0.0107279581661
 0.0107265535218
 0.0107241180809

0.0000001652
 0.0000001502
 0.0000001366
 0.0000001241
 0.0000001128
 0.0000001026
 0.0000000933
 0.0000000848
 0.0000000771

Example
 Evolution of the
 energy and the
 magnetization

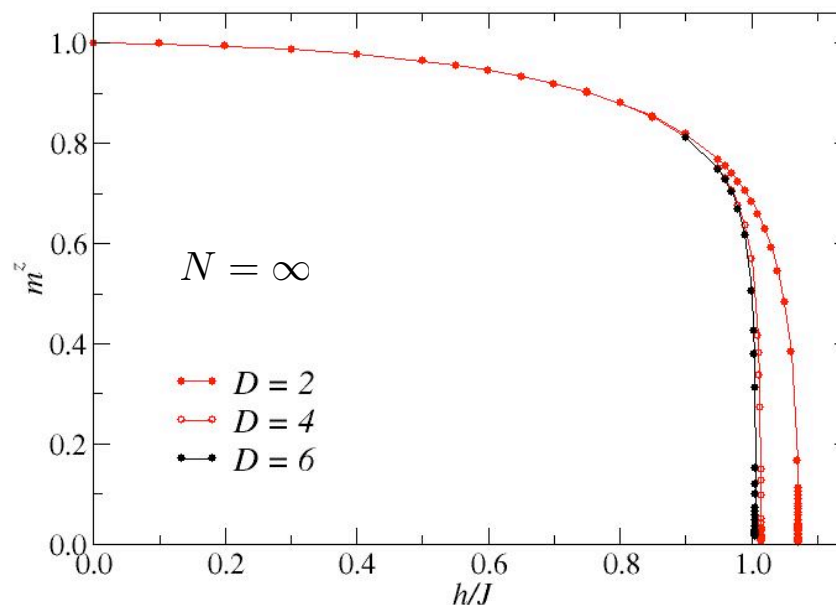
relative change = 3×10^{-14}

3×10^{-3}

Close to the critical point:

Small change in $E \rightarrow$
large (relative) change in m^z

- can be a serious issue when analyzing the critical behavior



Comparison with imaginary-time projection (TEBD); D=4, h=1.014334

$E = -1.282454538906097$

$m = 0.004589923026775$ (I. McCulloch, standard)

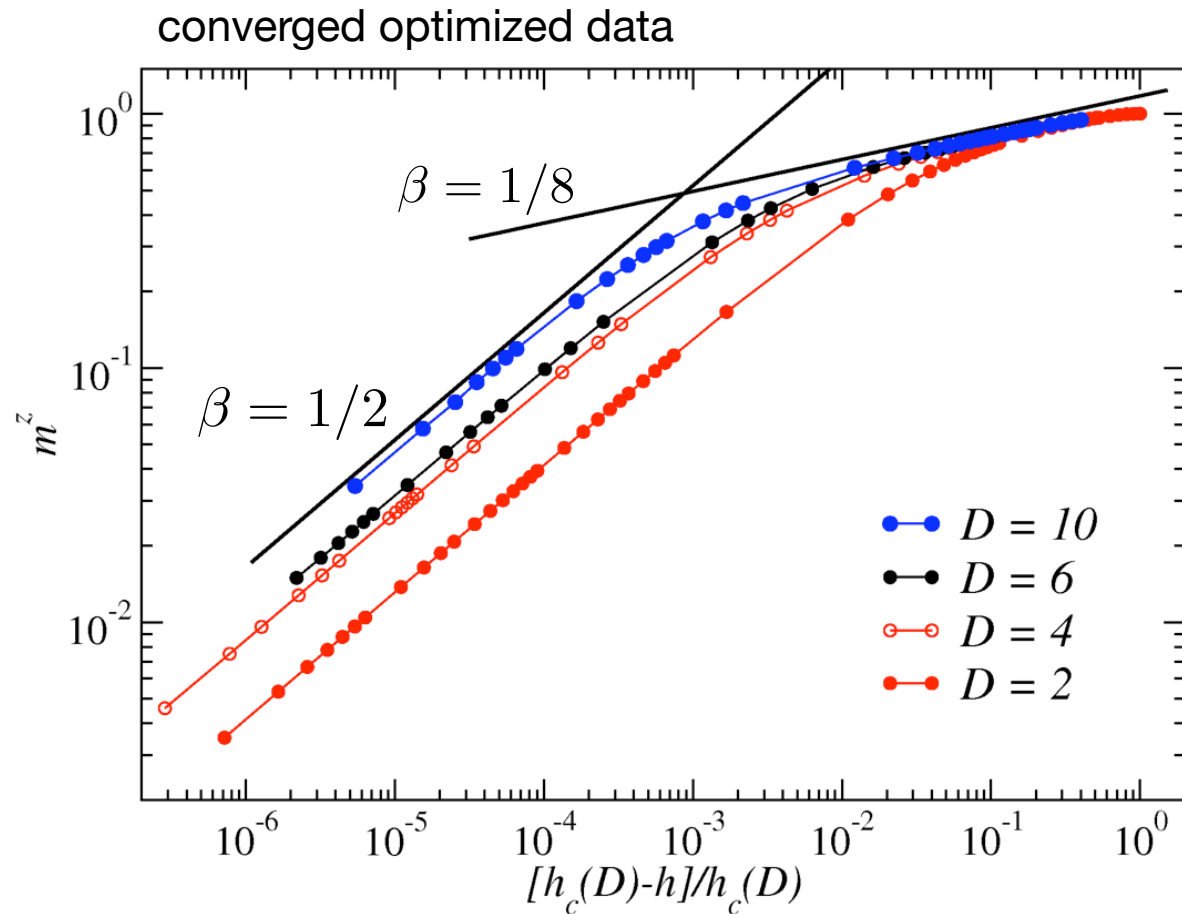
$E = -1.28245453890609554713490$

$m = 0.004589765790234$ (Random optimization)

Analysis of the critical behavior

Power-law fit for small m^z always gives $\beta \approx 0.50$

- indicates asymptotic mean-field behavior



- For finite D , asymptotic critical behavior is of mean-field type
 - cross-over to the true critical exponent
- numerical precision may limit access to critical behavior

The asymptotic mean-field behavior for MPS is not surprising

- finite $D \rightarrow$ maps to classical 1D transfer matrix
- criticality in 1D classical system requires long-range interactions

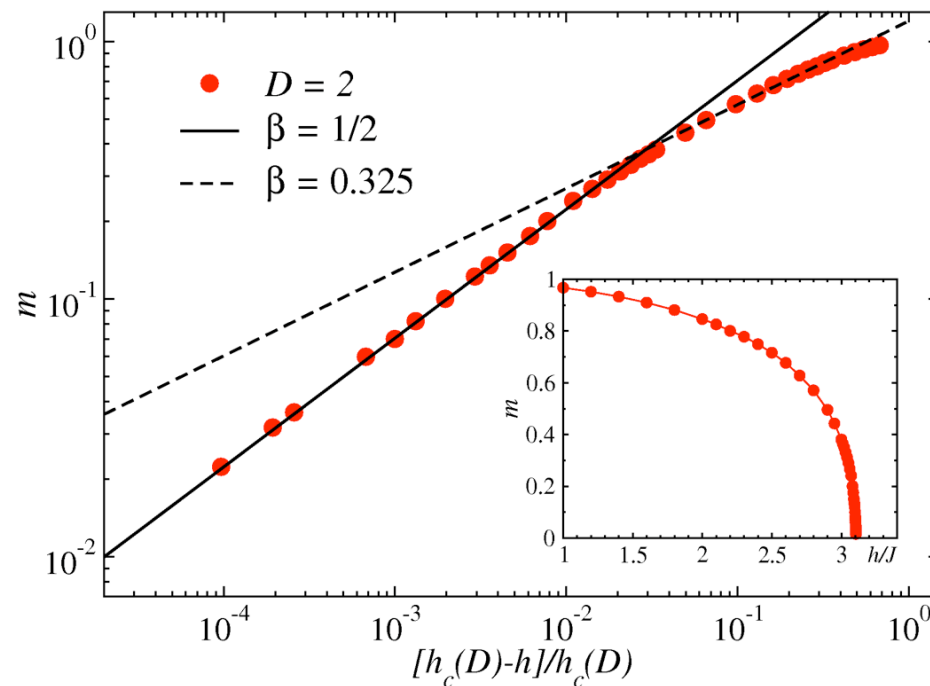
How about 2D PEPS?

Finite $D \rightarrow$ classical 2D partition function; critical points exist

- non-trivial exponents have been seen (?) for $D=2,3$ iPEPS

Infinite-size PEPS (iPEPS) [Orus & Vidal (2009)]

- Generalization of the $N=\infty$ MPS (but more complicated, approximations)
- We use new stable optimization/contraction [Wang & Verstraete]



← 2D transverse-field Ising

- mean-field cross-over
- to extract the true exponent requires careful check of convergence with D
- the true exponent emerges in a window away from h_c
- similar cross-overs in classical systems (Baxter, Nishino et al.,...)

Conclusions

Symmetry-breaking in MPS

- **first-order for finite N , finite D**
- **continuous mean-field transition for $N=\infty$, finite D**
 - mean-field window shrinks as $D \rightarrow \infty$
 - true exponents emerge through cross-over behavior

Numerical precision issues (variational optimization)

- **$N=\infty$ optimization difficult close to phase transition**
 - relative error of order parameter large even if E converged
 - difficult to extract the critical point precisely

$N=\infty$ PEPS (IPEPS)

- **mean-field criticality for finite D**
 - existence of non-trivial critical points for finite D does not mean that one automatically obtains correct critical behavior for a given H