

- The Density Matrix Renormalization Group (DMRG) was invented by Steve White (student of Ken Wilson) to solve general quantum chain models. [White1992], [White1993]
- First realization of connection between MPS and DMRG in limit $N \rightarrow \infty$: Ostlund & Rommer [Ostlund1995]
- Realization that finite-size DMRG leads to MPS: Dukelsky, Martin-Delgado, Nishino, Sierra [Dukelski1998]
- Modern formulation: Vidal [Vidal2003], [Vidal2004], Cirac & Verstraete [Verstraete2004]
- Time evolution: Daley, Kollath, Schollwöck, Vidal [Daley2004], White, Feiguin [White2004]
- Connection to NRG: Weichselbaum, Verstraete, Schollwöck, Cirac, von Delft [arXiv:0504305], [Weichselbaum2009]

1. Iterative ground state search

View space of all MPS of given bond dimension, D , as variational space.

Minimize $\langle \psi | \hat{H} | \psi \rangle$ in this space, subject to constraint of unit normalization, $\langle \psi | \psi \rangle = 1$. (1)

Hence extremize $\langle \psi | \hat{H} | \psi \rangle - \lambda \langle \psi | \psi \rangle$ (2)
↑ Lagrange multiplier

Graphical representation, assuming mixed-canonical form w.r.t. site l :

(3)

Do this one A -tensor at a time: $\frac{\partial}{\partial A_{[el]}^t} [\langle \psi | \hat{H} | \psi \rangle - \lambda \langle \psi | \psi \rangle] = 0$ (4)

(5)

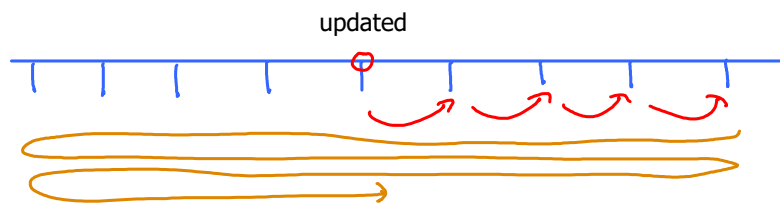
= λ (6)
 close zippers from left and right

In the notation of (MPS-V.3.11): $H_a^{a'} A_{[e]}^a = \lambda A_{[e]}^{a'}$ with $\alpha' = (\alpha', \sigma', \beta')$ (7)

This is an eigenvalue equation for $A_{[e]}$ and can be solved with standard linear algebra tools, e.g. Lanczos algorithm (next section).

More generally: if $|4\rangle$ is not represented in mixed-canonical form, one obtains a generalized eigenvalue equation of the form $HA = NA$, with N defined by r.h.s. of (5). (6)

Use that 'eigenvector' $A_{[e]}$ yielding lowest eigenvalue (= current estimate of ground state energy) to 'update' MPS, then move to next site, switch to mixed-canonical form of site $l+1$, optimize $A_{[l+1]}$, etc..



'Sweep' back and forth until convergence of ground state energy has been achieved. This works remarkably well for 1D chains with short-ranged interactions.

- Fast way of finding extremal eigenvalues of an Hermitian $N \times N$ matrix, H .
- Prerequisite: an algorithm for computing $H|\psi\rangle$, for any vector $|\psi\rangle$.

We seek the extremal value of $E[|\psi\rangle] = \frac{\langle \psi | H | \psi \rangle}{\langle \psi | \psi \rangle}$ (1)

Denote extremal value by $E_g = \min E[|\psi\rangle] \equiv E[|\psi_g\rangle]$ (2)

The direction of steepest ascent of the functional $E[|\psi\rangle]$, evaluated at $|\psi\rangle$, is given by

'functional gradient': $\frac{\delta E[|\psi\rangle]}{\delta \langle \psi |} \equiv \frac{H|\psi\rangle}{\langle \psi | \psi \rangle} - \frac{\langle \psi | H | \psi \rangle}{\langle \psi | \psi \rangle^2} |\psi\rangle$ (3)

$$= \frac{H - E[|\psi\rangle]}{\langle \psi | \psi \rangle} |\psi\rangle \equiv |\psi_a\rangle \quad (4)$$

Moving in opposite direction will thus lower the energy:

$$E[|\psi\rangle - \alpha |\psi_a\rangle] < E[|\psi\rangle] \quad \text{for small, positive } \alpha \quad (5)$$

To find optimal value for α , minimize $E[|\psi\rangle - \alpha |\psi_a\rangle]$ w.r.t. the 'variational parameter' α ,

in the space $K_1 \equiv \text{span}\{|\psi\rangle, |\psi_a\rangle\} = \text{span}\{|\psi\rangle, H|\psi\rangle\}$. (6)

Construct a normalized basis for this space (for a random initial state $|\psi\rangle$):

First basis vector: $|\psi_0\rangle = \frac{|\psi\rangle}{\sqrt{\langle \psi | \psi \rangle}}$ (7)

Second basis vector: $b_1 |\psi_1\rangle \equiv |\tilde{\psi}_1\rangle = H|\psi_0\rangle - |\psi_0\rangle \langle \psi_0 | H | \psi_0 \rangle$ (8)

normalization factor, such that $\langle \psi_1 | \psi_1 \rangle = 1$

orthonormalize w.r.t. to $|\psi_0\rangle$

$$\langle \psi_1 | \psi_1 \rangle: \quad b_1 = \sqrt{\langle \tilde{\psi}_1 | \tilde{\psi}_1 \rangle} = \langle \psi_1 | H | \psi_0 \rangle \quad (9)$$

Now find a matrix representation of H in this space: define

$$a_0 \equiv \langle \psi_0 | H | \psi_0 \rangle, \quad a_1 \equiv \langle \psi_1 | H | \psi_1 \rangle, \quad b_1^2 \equiv \langle \tilde{\psi}_1 | \tilde{\psi}_1 \rangle \quad (10)$$

then

$$H|\psi_0\rangle \stackrel{(8)}{=} |\psi_1\rangle b_1 + |\psi_0\rangle a_0 \quad (11)$$

hence in the space K_1 , the Hamiltonian has the matrix representation

$$\left(\begin{array}{cc} \langle \psi_0 | H | \psi_0 \rangle & \langle \psi_0 | H | \psi_1 \rangle \\ \langle \psi_1 | H | \psi_0 \rangle & \langle \psi_1 | H | \psi_1 \rangle \end{array} \right) \left(\begin{array}{c} a_0 \\ b_1^* \end{array} \right)$$

hence in the space \mathcal{K}_1 , the Hamiltonian has the matrix representation

$$H_{\mathcal{K}_1} = \begin{pmatrix} \langle v_0 | H | v_0 \rangle & \langle v_0 | H | v_1 \rangle \\ \langle v_1 | H | v_0 \rangle & \langle v_1 | H | v_1 \rangle \end{pmatrix} = \begin{pmatrix} a_0 & b_1^* \\ b_1 & a_1 \end{pmatrix} \quad (12)$$

The ground state of $H_{\mathcal{K}_1}$, say $|\psi\rangle_{\mathcal{K}_1}$, yields the optimal choice for α .

Now we could iterate: use $|\psi\rangle_{\mathcal{K}_1}$ as starting point for another optimization step. Convergence is rapid. Monitor quality of result by computing the residual energy variance,

$$\tau[|\psi\rangle] \equiv \| (H - E)|\psi\rangle \|^2 = \langle \psi | H^2 | \psi \rangle - \langle \psi | H | \psi \rangle^2 \quad (13)$$

and stop when it drops below some threshold.

Krylov space

After L steps, starting from $|v_0\rangle$, the resulting vector will live in

$$K_L(|v_0\rangle) = \text{span} \{ |v_0\rangle, H|v_0\rangle, H^2|v_0\rangle, \dots, H^L|v_0\rangle \} \quad (14)$$

$$= \text{'Krylov space of } H \text{ over } |v_0\rangle \text{' (dimension } L+1 \text{)}. \quad (15)$$

Instead of repeatedly minimizing in 2×2 subspaces, we could first construct K_L , then compute its ground state. (This is faster, since it amounts to using L simultaneous variational parameters instead of L separate ones.) To do this, iteratively construct a 'Krylov basis' for K_L :

Krylov basis

As before: $b_1 |v_1\rangle \equiv |\tilde{v}_1\rangle = H|v_0\rangle - a_0 |v_0\rangle \quad (16)$

normalize

Third vector: $b_2 |v_2\rangle \equiv |\tilde{v}_2\rangle = H|v_1\rangle - \sum_{j=0}^1 |v_j\rangle \langle v_j | H | v_1 \rangle \quad (17)$

$$= H|v_1\rangle - |v_1\rangle \underbrace{a_1}_{\langle v_1 | H | v_1 \rangle} - |v_0\rangle \underbrace{b_1^*}_{\langle v_0 | H | v_1 \rangle} \quad (18)$$

where $b_2 \stackrel{(17)}{=} \sqrt{\langle \tilde{v}_2 | \tilde{v}_2 \rangle} \stackrel{(18)}{=} \langle v_2 | H | v_1 \rangle \quad (19)$

Note: $\langle v_2 | H | v_0 \rangle = 0$, since $H|v_0\rangle \in \text{span} \{ |v_0\rangle, |v_1\rangle \}$ (20)

Fourth vector: $b_3 |v_3\rangle \equiv |\tilde{v}_3\rangle = H|v_2\rangle - \sum_{j=0}^2 |v_j\rangle \langle v_j | H | v_2 \rangle \quad (21)$

$$= H|v_2\rangle - |v_2\rangle \underbrace{a_2}_{\langle v_2 | H | v_2 \rangle} - |v_1\rangle \underbrace{b_2^*}_{\langle v_1 | H | v_2 \rangle} - |v_0\rangle \underbrace{0}_{\langle v_0 | H | v_2 \rangle} \quad (22)$$

Summary

1. Start with arbitrary $|\psi_0\rangle$

2. First iteration step: (i) $|\tilde{\psi}_1\rangle = H|\psi_0\rangle$

(ii) $a_0 = \langle \tilde{\psi}_1 | \psi_0 \rangle$

(iii) $|\hat{\psi}_1\rangle = |\tilde{\psi}_1\rangle - a_0|\psi_0\rangle$

3. General iteration step, for $n \geq 1$:

(i) $b_n = \langle \hat{\psi}_n | \hat{\psi}_n \rangle$

(ii) If $b_n \neq 0$, then $|\psi_n\rangle = |\hat{\psi}_n\rangle / b_n$

else, pick $|\psi_n\rangle$ as arbitrary normalized vector orthogonal to all $|\psi_0\rangle, \dots, |\psi_{n-1}\rangle$

(iii) $|\tilde{\psi}_{n+1}\rangle = H|\psi_n\rangle$

(iv) $a_n = \langle \tilde{\psi}_{n+1} | \psi_n \rangle$

(v) $|\hat{\psi}_{n+1}\rangle = |\tilde{\psi}_{n+1}\rangle - |\psi_n\rangle a_n - |\psi_{n-1}\rangle b_n^*$

and back to 3(i).

There are other ways of organizing this iteration loop, but the one shown here is numerically the most stable.
[Paige1972]

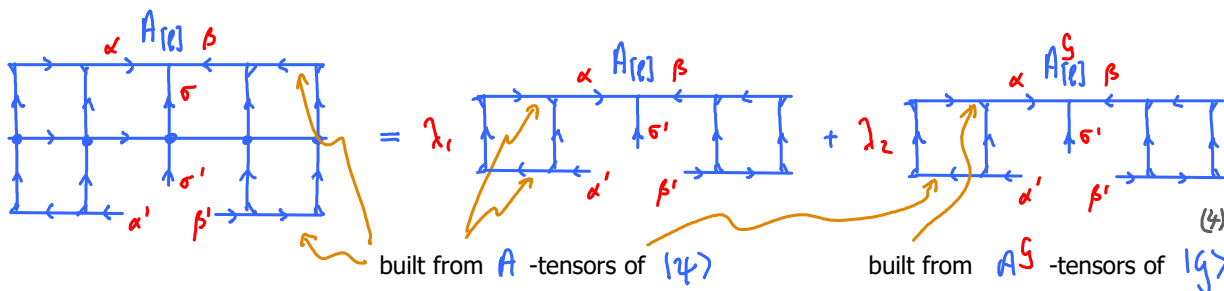
Suppose we have an MPS representation for ground state, $|g\rangle = |\bar{\sigma}\rangle \prod_l A_{[\ell]}^{\sigma_l}$ (1)

found by DMRG. Excited states can be constructed repeating a DMRG sweep in space orthogonal to $|g\rangle$.

Extremize: $\langle \psi | H | \psi \rangle - \lambda_1 \langle \psi | \psi \rangle - \lambda_2 \langle \psi | g \rangle$ (2)

Lagrange multipliers enforce $\langle \psi | \psi \rangle = 1$ and $\langle \psi | g \rangle = 0$. (3)

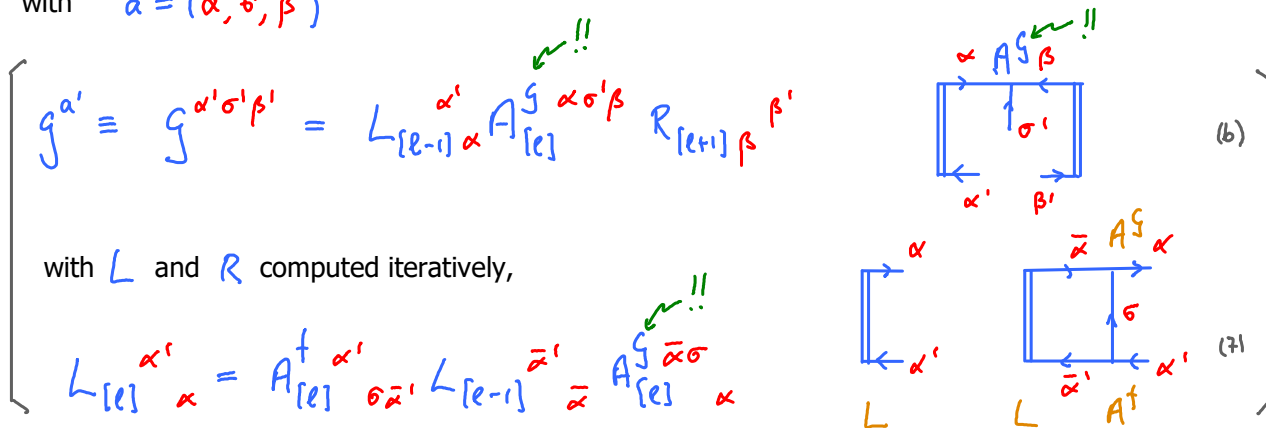
Extremization w.r.t. $A_{[\ell]}^\dagger$ yields



Generic structure of this equation, in mixed-canonical representation of site l [compare (DMRG-I.1.7)]:

$$H_a^{a'} A_{[\ell]}^a = \lambda_1 A_a^{a'} + \lambda_2 g_a^{a'}, \quad \text{with} \quad A_a^\dagger A^a = 1, \quad A_a^\dagger g^a = 0, \quad (5)$$

with $a' = (\alpha', \sigma', \beta')$



with L and R computed iteratively,

$$L_{[\ell]}^{\alpha'} = A_{[\ell]}^{\dagger \alpha'} \bar{\sigma} \bar{\alpha}' L_{[\ell-1]}^{\bar{\alpha}'} \bar{\alpha} \bar{\sigma} A_{[\ell]}^{\sigma} \alpha$$

Index-free notation for (5): $H|A\rangle = \lambda_1 |A\rangle + \lambda_2 |g\rangle, \quad \langle A|g\rangle = 0$ (6)

Define projector onto subspace orthogonal to $|g\rangle$: $P_g = \mathbb{1} - |g\rangle\langle g|$ (7)

[with indices: $P_g^{a'}{}_a = \mathbb{1}^{a'}{}_a - g^{a'} g_a$, so that $P_g^{a'}{}_a A^a = 0$] (10)

Project (8) onto this subspace: $P_g H (P_g + |g\rangle\langle g|) |A\rangle = \lambda_1 P_g |A\rangle + 0$ (11)

$$\overbrace{(8)} = 0$$

$$P_g H P_g |A\rangle = \lambda_1 P_g |A\rangle \quad (12)$$

This is simply an eigenvalue problem, for $P_g H$, in subspace orthogonal to $|g\rangle$. It can be solved

using straightforward generalization of Lanczos scheme, using Krylov subspace orthogonal to $|g\rangle$:

Given an arbitrary initial state $|v_0\rangle$, project it onto orthogonal subspace, $|v_0'\rangle = P_g |v_0\rangle$, (13)

and construct new Krylov vectors using

$$|\tilde{v}_{n+1}\rangle = P_g H |v_n\rangle - |v_n\rangle a_n - |v_{n-1}\rangle b_n^* \quad (14)$$

Why not simply use excited states in K_L ? Because numerical noise can cause the $|v_n\rangle$

to be not exactly orthogonal, hence for $j < n-2$, $\langle v_n | v_j \rangle \approx 10^{-16}$ rather than 0.

This leads to spurious multiple copies of eigenstates ('ghost states'). For the ground state, the variational principle ensures that the loss of orthogonality does not become a severe problem. But for excited states, it does. To prevent this, explicit reorthogonalization is needed at every step, using P_g , as indicated above.

Block-Lanczos for excited states

Standard Lanczos: represent action of H as

$$H |v_0\rangle = |v_0\rangle a_0 + |v_1\rangle b_1 \Rightarrow$$

$$\begin{pmatrix} a_0 & b_1^* & & \\ b_1 & a_1 & b_2^* & \\ & b_2 & \ddots & \end{pmatrix}$$

Block-Lanczos: start with set of M orthogonal vectors,

$|v_{0,i}\rangle$, $i = 1, \dots, M$, and represent action of H as

$$H |v_{0,i}\rangle = |v_{0,j}\rangle \delta_{ij}^j (a_1)^i + |v_{1,j}\rangle (b_1)^j_i$$

with $\langle v_{0,j} | v_{1,i} \rangle = 0$, $\langle v_{1,j} | v_{1,i} \rangle = \delta_{ij}^j$

and $(a_1)^i = \langle v_{0,i} | H |v_{0,i}\rangle$, $(b_1)^j_i = \langle v_{1,j} | H |v_{0,i}\rangle$

etc. Then the lowest M eigenstates of block-tridiagonal matrix give the Lanczos approximation for lowest M eigenstates of H

$$\begin{pmatrix} [a_0] & [b_1^+] \\ [b_1] & [a_1] [b_2^+] \\ & [b_2] & \ddots \end{pmatrix}$$

If one encodes symmetries (see Sym-I to Sym-III), then 'one-site update' (discussed above) can get stuck: if one starts in the wrong symmetry sector, one stays there, because one-site update offers no way of enlarging the Hilbert space during the variational search to explore other symmetry sectors.

Cure: 'two-site' update, which variationally optimizes two A-tensors at a time.

Represent MPS in mixed-canonical two-site basis:

$$|\psi\rangle = |\beta\rangle |\sigma_{2l+1}\rangle |\sigma_{2l}\rangle |\alpha\rangle A_{[2l]}^{\alpha\sigma_{2l}} B_{[2l+1]}^{\sigma_{2l}\beta} \quad (1)$$

Then extremize simultaneously w.r.t.

$$A_{[2l]}^\dagger \text{ and } B_{[2l+1]}^\dagger \quad \frac{\partial}{\partial B_{[2l+1]}^\dagger} \frac{\partial}{\partial A_{[2l]}^\dagger} \left[\langle \psi | \hat{H} | \psi \rangle - \lambda \langle \psi | \psi \rangle \right] = 0 \quad (2)$$

$$= \lambda \quad (3)$$

close zippers from left and right $= \lambda$

$$= \lambda \quad (4)$$

Compact notation: $H_a^{\alpha'} (AB)^a = \lambda (AB)^{\alpha'}$ with composite index $a = (\alpha, \sigma, \bar{\sigma}, \beta)$ (5)

and

$$H_a^{\alpha'} = \quad \text{[Diagram of contracted two-site tensor network]} \quad = \quad \text{[Diagram of contracted two-site tensor network]} \quad (6)$$

$L_{[2l-1]} W_{[2l]} W_{[2l+1]} R_{[2l+2]}$

Use Lanczos to find lowest eigenvalue of eigenvalue equation (5), and reshape updated (\widetilde{AB}) :

updated $(\widetilde{AB})^a \xrightarrow{\text{reshape}} \alpha, \sigma \text{---} \sigma, \bar{\sigma} \xrightarrow{\text{SVD}} \alpha, \sigma \text{---} U \text{---} S \text{---} V^\dagger \text{---} \sigma, \bar{\sigma}$ (7)

$D_d \quad D_d \quad D_d \quad D_d$

Key point: S has dimensions $D_d \times D_d$, hence explores a larger state space than previously, in general also including different symmetry sectors!

Truncate down to D and reshape: $\approx \alpha, \sigma \text{---} U \text{---} (\widetilde{S}) \text{---} V^\dagger \text{---} \sigma, \bar{\sigma} \equiv \alpha \text{---} \widetilde{A} \text{---} \gamma \text{---} \widetilde{B} \text{---} \beta$ (8)

$D_d \quad D \quad D \quad D_d$

This concludes optimization of $\widetilde{A}_{[2l]}$. Now move one site to the right and repeat. Sweep back and forth until convergence of full chain.