

- 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]

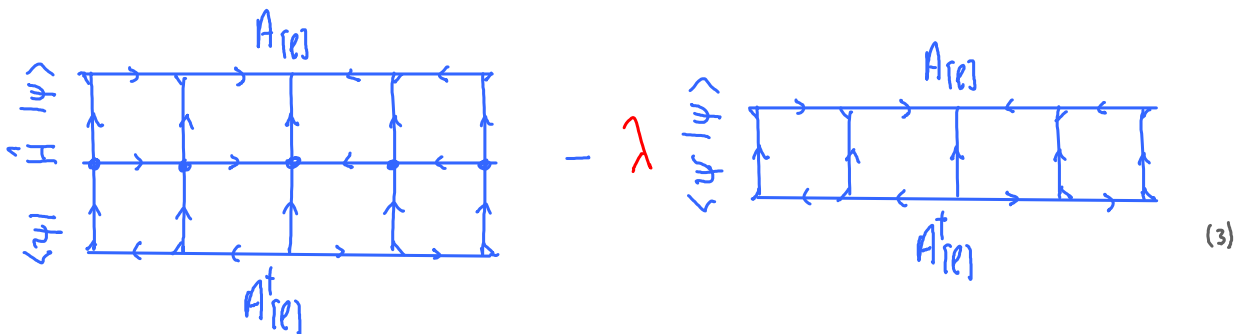
1. Iterative ground state search

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

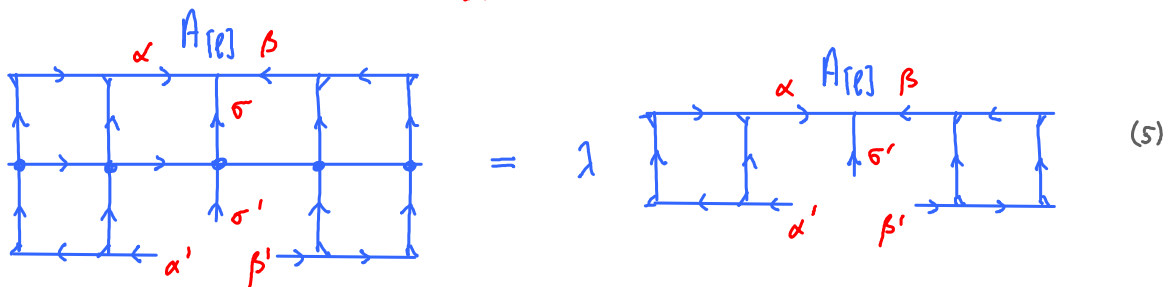
Minimize  $\langle \psi | \hat{H} | \psi \rangle$  in this space, subject to constraint of unit normalization,  $\langle \psi | \psi \rangle = 1$ . (1)

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  $l$  :



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



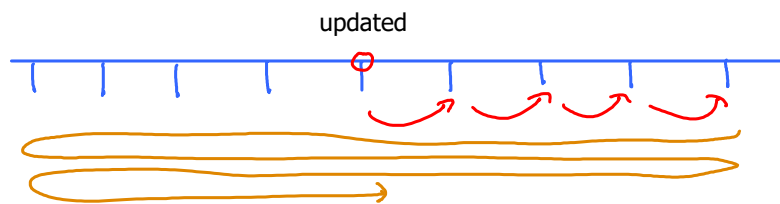
=  $\lambda$  (6)  
 close zippers from left and right

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.

## 2. Lanczos method

[Lanczos1950], [Ojalvo1970], [Paige1972], [Koch2011]  
 original idea                      stable version   nice discussion

DMRG-I.2

- Fast way of finding extremal eigenvalues of an Hermitian NxN 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)

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

Moving in opposite direction will thus lower the energy:

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

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

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

Starting from the random initial state  $|\psi\rangle$ , construct a normalized basis  $\{|\psi_0\rangle, |\psi_1\rangle\}$  for this space:

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

Second basis vector:  $b_1 |\psi_1\rangle := |\tilde{\psi}_1\rangle := H|\psi_0\rangle - |\psi_0\rangle \langle\psi_0|H|\psi_0\rangle$  (8)  
normalization factor, such that  $\langle\psi_1|\psi_1\rangle = 1$                       orthonormalize w.r.t. to  $|\psi_0\rangle$

$$b_1 := \sqrt{\langle\tilde{\psi}_1|\tilde{\psi}_1\rangle} \quad \langle\psi_1|\psi_1\rangle = \langle\psi_1|H|\psi_0\rangle + 0 \quad \text{since } \langle\psi_1|\psi_0\rangle \stackrel{(8)}{=} 0$$
 (9)

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

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

then

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

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

$$H_{K_1} = \begin{pmatrix} \langle v_0 | H | v_0 \rangle & \langle v_0 | H | v_1 \rangle \\ \langle v_1 | H | v_0 \rangle & \langle v_1 | H | v_1 \rangle \end{pmatrix} = \begin{pmatrix} a_0 & b_1 \\ b_1 & a_1 \end{pmatrix} \quad (12)$$

The ground state of  $H_{K_1}$ , say  $|g\rangle_{K_1}$ , yields the optimal choice for  $\alpha$ .

Now we could iterate: use  $|g\rangle_{K_1}$  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 2x2 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  $\{ |v_0\rangle, |v_1\rangle, \dots, |v_L\rangle \}$

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

normalize

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

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

where

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

real

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

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



## 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)  $b_n = \sqrt{\langle \hat{\psi}_n | \hat{\psi}_n \rangle}$

(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]

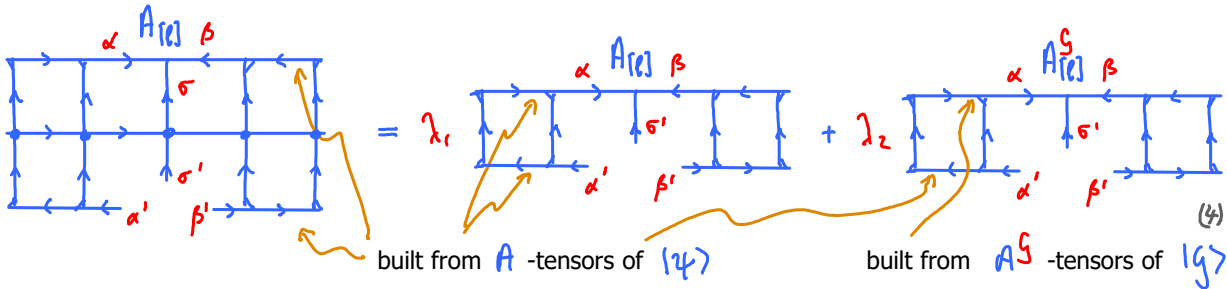
Suppose we have an MPS representation for ground state,  $|g\rangle = |\bar{\sigma}\rangle \prod_l A_{[\ell]}^{\sigma_l}$  (1)

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)]:

$$H_{a'}^a A_{[\ell]}^a = \lambda_1 A_{a'}^a + \lambda_2 g_{a'}^a, \quad \text{with} \quad \begin{matrix} A_a^\dagger A^a = 1 \\ g_a^\dagger g^a = 1 \end{matrix} \quad (3)$$

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

$$g_{a'}^a \equiv g^{\alpha' \sigma' \beta'} = L_{[\ell-1]}^{\alpha'} A_{[\ell]}^{\sigma'} R_{[\ell+1]}^{\beta'}$$

with  $L$  and  $R$  computed iteratively,

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

Index-free notation for (5):  $H|A\rangle = \lambda_1 |A\rangle + \lambda_2 |g\rangle, \quad \langle A|g\rangle = 0$  (8)

Define projector onto subspace orthogonal to  $|g\rangle$ :  $P_g = \mathbb{1} - |g\rangle\langle g|$  (9)

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

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

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

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

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

$$\begin{pmatrix} \begin{bmatrix} a_0 \\ b_1 \end{bmatrix} \begin{bmatrix} b_1^+ \\ a_1 \end{bmatrix} \\ \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \begin{bmatrix} a_1 \\ b_2^+ \end{bmatrix} \\ \vdots \end{pmatrix}$$



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

$$H_a^{\alpha'} = \quad (6)$$

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

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

Key point:  $S$  has  $Dd$  singular values, larger than the virtual bond dimension  $D$  of  $A$  and  $B$ . Hence, it explores a larger state space, in general also including more symmetry sectors!

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

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