- 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 $\stackrel{\checkmark}{\sim}$ $\stackrel{\checkmark}{\sim}$: 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]

DMRG.1 Iterative ground state search

View space of all MPS of given bond dimension, D, as <u>variational</u> space.

Graphical representation, assuming site-canonical form with orthogonality center at site ℓ :

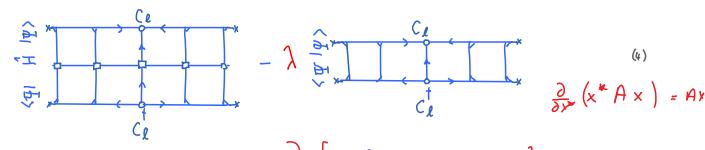
Arrow convention: use same arrow directions on virtual bonds for MPO as in MPS. Then, orientation of MPS triangles, \top , \bot , hence we henceforth drop most (soften all) arrows.

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

Hence extremize

$$\langle \Psi | \hat{H} | \Psi \rangle - \frac{1}{2} \langle \Psi | \Psi \rangle$$
(2)

Lagrange multiplier



Do this one tensor at a time:

tensor at a time:
$$\frac{\partial}{\partial C_{\ell}^{+}} \left[\langle \vec{q} | \hat{H} | \vec{q} \rangle - \lambda \langle \vec{q} | \vec{q} \rangle \right] = 0 \qquad (5)$$



$$= \lambda$$

Structure of (6): 1-site Schrödinger equation!

with
$$\psi_{a}^{(1)} = C_{\ell}$$
 (7)

with normalization

$$\psi_{\ell}^{(1)} \psi_{\ell}^{(1)} = 1$$

Here, ℓ_{ℓ} is viewed as vector, labeled by composite index $\alpha' = (\alpha', \sigma'_{\ell}, \beta')$, and $H_{\ell}^{(1)}$ as a matrix:

$$\left(\left|-\right|_{\ell}^{(1)}\right)_{\alpha}^{\alpha} \left(\left|-\right|_{\ell}\right)_{\alpha}^{\alpha} = \lambda \left(\left|-\right|_{\ell}\right|_{\alpha}^{\alpha} \quad \text{with normalization} \quad \left(\left|-\right|_{\ell}\right|_{\alpha}^{\alpha} \left|-\right|_{\alpha}^{\alpha} \left|-\right|-\right|_{\alpha}^{\alpha} \left|-\right|_{\alpha}^{\alpha} \left|-\right|-\right|_{\alpha}^{\alpha} \left|-\right|_{\alpha}^{\alpha} \left|-\right|_{\alpha}^{\alpha} \left|-\right|_{\alpha}^{\alpha} \left|-\right|_{\alpha}^$$

compare (MPS.15.11)

$$\left[H_{\ell}^{(1)}\right]^{a'} = \left[\begin{array}{c} \left(D^{2}d\right) \times \left(D^{2}d\right) \\ \left(D^{2}d\right) \times$$

(7) is an eigenvalue equation for \mathcal{C}_{ℓ} . The lowest eigenvalue and eigenvector can be found with standard linear algebra tools (e.g. Lanczos algorithm, next section), without having to construct $H_{\ell}^{(i)}$ fully. It suffices to know how to compute $H_{\ell}^{(i)}$ \mathcal{C}_{ℓ} .

More generally: if $|\Psi\rangle$ is not represented in site-canonical form, one obtains a generalized eigenvalue equation of the form $H_{\ell}^{(i)} = H_{\ell}^{(i)} =$

Use the 'eigenvector' with the lowest eigenvalue (= current estimate of ground state energy), say C_{ℓ}^{S} to 'update' MPS, then move to next site, use SVD on C_{ℓ}^{S} to shift orthogonality center to site $\ell \ell \ell$:

$$\frac{C_{\ell}^{q}}{L} = \frac{\tilde{A}_{\ell}}{L} \times \left(U \right) \left(S \times V^{\dagger} \times B_{\ell+1} \right)^{2} \times C_{\ell+1} = \frac{\tilde{A}_{\ell}}{L} \times C_{\ell+1} = \frac{\tilde{A}_{\ell}}{L}$$

Compute new environments \mathcal{L}_{ℓ} and $\mathcal{R}_{\ell+2}$ for site ℓ , then optimize $\mathcal{C}_{\ell+1}$, etc..

h until convergence of ground state energy has been achieved.

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

Cost of 1-site DMRG = cost of computing $H_{\ell}^{(i)} \psi_{\ell}^{(i)} : \mathcal{O}(\mathcal{D}_{d}^{3} \cup \mathcal{D}_{d}^{2} \cup \mathcal{D}_{d}^{2})$ (expensive!) need not be constructed explicitly!

DMRG.2

• Fast way of finding extremal eigenvalues of an Hermitian NxN matrix,

• Prerequiste: an algorithm for computing
$$| \cdot | \cdot | \cdot | \cdot |$$
 , for any vector $| \cdot | \cdot | \cdot |$.

Denote extremal value by
$$E_g = \min E[(\psi)] =: E[1\psi_g)$$
 (2)

The direction of steepest ascent of the functional $\mathcal{E}(\mathcal{V})$, evaluated at \mathcal{V} , is given by

'functional gradient':
$$\frac{\delta \in [14)}{\delta \langle 4|} := \frac{H147}{\langle 414\rangle} - \frac{\langle 4|H14\rangle}{\langle 4|4\rangle^2} [4] \tag{3}$$

$$= \frac{H - E[N47](4)}{\langle 4|4\rangle} (4)$$

Moving in opposite direction will thus lower the energy:

$$E[|\psi\rangle - \alpha |\psi_a\rangle] < E[|\psi\rangle]$$
 for small, positive α (5)

To find optimal value for α , minimize $\mathbb{E}\left[147 - \alpha | 4\alpha \right]$ w.r.t. the 'variational parameter' α , in the 'Krylov space' $K_1 := \text{Span}\left[147, | 4\alpha \right] = \text{Span}\left[147, | 4\alpha \right]$. (6)

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

First Krylov step: explore the second direction in Krylov space by applying H to $|v_b\rangle$:

Define
$$|\vec{v}_i\rangle := H(\vec{v}_i)$$
 (8)

Orthogonalize w.r.t.
$$|v_0\rangle$$
: = $|\tilde{v}_1\rangle$: = $|\tilde{v}_1\rangle$ - $|v_0\rangle\langle v_0|\tilde{v}_1\rangle$ (9)

ensuring
$$\langle v_o | v_i^{\perp} \rangle = 0$$
 (10)

2nd basis vector:
$$|v_i\rangle = |v_i^{\perp}\rangle/|b_i\rangle$$

Rewrite (9):
$$|v_1\rangle_b = |v_1\rangle_b = |v_2\rangle_b - |v_3\rangle_b = |$$

 $\frac{101}{\text{define } := a_0} = (v_0 | \hat{v_1})$.../

$$\langle \upsilon_{i} |$$
 (14) and (10) yield: $\langle \upsilon_{i} | H | \upsilon_{o} \rangle = 0 + \frac{1}{2} = \frac{\langle \upsilon_{o} | H | \upsilon_{i} \rangle}{\text{since } b_{i} \text{ is real, (11)}}$

Finally, define
$$a_1 := \langle v_i | H | v_i \rangle \stackrel{(z_i)}{=}$$

Now we have orthonormal basis for K1:= spun { (50), 15,7} = spun { 100), HIVE) } 2-dimensional Krylov space:

In the space K_1 , the Hamiltonian has the matrix representation

$$H_{K_{I}} = \begin{pmatrix} \langle v_{0}|H|v_{0} \rangle & \langle v_{0}|H|v_{1} \rangle \\ \langle v_{1}|H|v_{0} \rangle & \langle v_{1}|H|v_{1} \rangle \end{pmatrix} = \begin{pmatrix} a_{0} & b_{1} \\ b_{1} & a_{1} \end{pmatrix}$$
(18)

The ground state of H_{K_1} , say $\frac{1}{5}$, with energy $E_{K_1}^{9}$, yields the optimal choice for \propto

Now we could iterate: use $\frac{1}{5}$ as starting point for another optimization step. Convergence is rapid. Monitor quality of result by computing the residual energy variance,

$$\tau[147] = ||(H - E)|47||^2 = \langle 4|H^2|4\rangle - \langle 4|H|4\rangle^2$$
 (19)

for $|\psi\rangle = |\mathcal{G}\rangle_{\kappa_1}$, $\mathcal{E} = \mathcal{E}_{\kappa_1}^{\mathcal{G}}$ and stop when it drops below some threshold.

After N steps, starting from (v_{ullet}) , the resulting vector will live in

$$\langle V_{N}(V_{0})\rangle = span \{|V_{0}\rangle, H|V_{0}\rangle, H^{2}(V_{0}), ..., H^{N}(V_{0})\}$$

$$= \text{'Krylov space of } H \text{ over } |V_{0}\rangle \text{'} \text{ (dimension } N+1).$$

Instead of repeatedly minimizing in 2x2 subspaces, we could $\underline{\text{first}}$ construct K_N , $\underline{\text{then}}$ compute its ground state. (This is faster, since it amounts to using | simultaneous variational parameters instead of $\,\,\,$ separate ones.) To do this, iteratively construct a 'Krylov basis' for $\,\,$ K $_{N}$, such that

$$\langle | \langle | v_{n} \rangle \rangle = \text{Span} \{ | v_{0} \rangle, | v_{1} \rangle, \dots, | v_{n} \rangle \}$$
 with $\langle v_{n} | v_{n'} \rangle = \delta_{nn'}$ (21)

We now elaborate this iteration strategy, first for the 2nd Krylov step, then for the (n+1)-th step.

<u>Second Krylov step:</u> explore a new direction in Krylov space by applying H to $| \psi_i \rangle$:

Define
$$|\hat{\mathbf{v}}_{i}\rangle_{i} = H(\mathbf{v}_{i})$$
 from $K_{i}(|\mathbf{v}_{0}\rangle)$ (22)

Define
$$|\widetilde{v}_{\ell}\rangle := H|v_{\ell}\rangle \qquad \text{Page in } K_{1}(|v_{0}\rangle) \qquad (22)$$
Orthogonalize:
$$|v_{2}^{\perp}\rangle := |\widetilde{v}_{\ell}\rangle - \sum_{j=0}^{\ell} |v_{j}\rangle\langle v_{j}|\widetilde{v}_{\ell}\rangle \qquad (23)$$

ensuring
$$\langle v_1 | v_2^{\perp} \rangle = 0$$
 $\int = 0, 1$ (24)

3rd basis vector:
$$|v_2\rangle = |v_2\rangle / |v_2\rangle$$
 (26)

Rewrite (23):
$$|v_2\rangle |_{z=0}^{(2l)} |v_2\rangle = ||(v_1)\rangle - |v_1\rangle\langle v_1|| + |v_1\rangle - |v_0\rangle\langle v_0|| + |v_1\rangle$$
 define $||v_1\rangle| + ||v_1\rangle| + ||v_1\rangle|$

Note:
$$\langle v_2 | H(v_0) = 0$$
, since $H(v_0) \in Span\{|v_0\rangle, |v_1\rangle\}$ (30)

Let K_2 : $H_{|v_0|} = \begin{pmatrix} a_0 & b_1 & 0 \\ b_1 & a_1 & b_2 \\ 0 & b_2 & a_2 \end{pmatrix}$ and we orthogonalized $|v_2\rangle$ w.r.t. $|v_0\rangle$, $|v_1\rangle$ [see (23,24)]

(2.11) the Know steps are provided in Know spaces by applying H to $|v_0\rangle$.

(n+1)-th Krylov step: explore a new direction in Krylov space by applying H to $|v_n\rangle$:

Define
$$|\widetilde{\mathcal{V}}_{n+1}\rangle := H|\mathcal{V}_n\rangle$$
 (31)

Define:
$$\alpha_n := \langle v_n | \tilde{v}_{n+1} \rangle = \langle v_n | H | v_n \rangle$$
 diagonal elements (32)

Orthogonalize:
$$|v_{n+1}^{\perp}\rangle := |\tilde{v}_{n+1}\rangle - \sum_{j=b}^{N} |v_{j}^{*}\rangle\langle v_{j}|\tilde{v}_{n+1}\rangle$$
 (33)

ensuring
$$\langle v_{n+1}^{\perp} \rangle = 0$$
 for $0 \leq j \leq N$ (34)

Normalize:
$$v_{n+1} := \sqrt{\langle v_{n+1}^{\perp} | v_{n+1}^{\perp} \rangle}$$
 (35)

(n+1)-th basis vector:
$$|v_{n+1}\rangle := |v_{n+1}^{\perp}\rangle / |v_{n+1}\rangle / |v_{n+1}\rangle$$
 (31)

[If it happens that $b_{N+1} = 0$, pick an arbitrary v_{N+1} orthonormal to all $v_{N+1} = 0$.]

Rewrite (33):
$$|v_{n+1}\rangle = |v_{n+1}\rangle = |v_{n+1}\rangle = |v_{n}\rangle - |v_{n}\rangle \langle v_{n}|H|v_{n}\rangle - |v_{n-1}\rangle \langle v_{n-1}|H|v_{n}\rangle + |v_{n-1}\rangle \langle v_{n-1}|H|v_{n}\rangle - |v_{n-1}\rangle \langle v_{n-1}|H|v_{n}\rangle + |v_{$$

All other terms vanish:
$$\langle v_{j} \mid H \mid v_{n} \rangle = \langle v_{n} \mid H \mid v_{j} \rangle = 0$$
 for $j \leq h-1$ farther-than-next-to-diagonal since by construction, $H \mid v_{j} \rangle \in Span \{ \mid v_{i} \rangle , o \in i \in j+1 \}$ and for $j \leq h-1$, i.e. $i \leq j+1 \leq N$, $|v_{n}\rangle$ is orthogonal to them all: $\langle v_{n} \mid v_{i}\rangle = 0$ for $i \leq N$ (31)

orthonormal Krylov basis:
$$v_0$$
, ..., v_1 , v_2 , v_{n-2} , v_{n-2} , v_n

Rearrange (37):
$$H(v_n) = |v_{n-1}\rangle b_n + |v_n\rangle a_n + |v_{n+1}\rangle b_{n+1} \qquad (40)$$

$$\langle v_n | (40) \text{ and } (34) \text{ yield}$$
 $\langle v_n | H | v_n \rangle = \langle v_n | H | v_n \rangle$ next-to-diagonal elements (41)

(38) holds if computations are done using <u>exact</u> arithmetic. In numerical practice, it does not hold strictly (typical violations are $O(rb^{-12})$), and errors accumulate. Hence it is advisable to orthogonalize a second time, directly after (32), before proceeding. This will be made explicit see below.

Hence, in
$$K_N$$
, the stridiagonal form:

$$K_N = \begin{cases} 360 & \text{form} \\ 400 & \text{form} \end{cases}$$

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Ground state of
$$H_{K_N}$$
 satisfies the eigenvalue equation $(H_{K_N})^i_j (Y_g^N)^j_j = F_g^N (Y_g^N)^i_j$

Thus
$$\mathsf{E}_{\mathsf{S}}^{\mathsf{N}} \quad \mathsf{and} \quad \mathsf{I}_{\mathsf{S}}^{\mathsf{N}} \mathsf{I} = \sum_{\mathsf{j}=\mathsf{0}} \mathsf{I}_{\mathsf{S}} \mathsf$$

are the best approximations, within the Krylov space $\binom{1}{N}$, of true ground state energy and ground state. The Lanczos scheme converges exponentially fast, with a rate \sim [gap to first excited state] $\frac{1}{2}$.

Summary Construct Krylov space of dimension N+I as follows:

1. Initialization: start with arbitrary (normalized) state (ທ້າ

Then repeat steps 2 to 4 for N = 0, ..., N-1:

2. Explore new direction in Krylov space by applying H:

$$|\widetilde{\mathcal{V}}_{n+1}\rangle := H|\mathcal{V}_n\rangle$$
 and $:= \langle \widetilde{\mathcal{V}}_{n+1}|\mathcal{V}_n\rangle$ (45)

3. Orthogonalize w.r.t. the two previous Krylov vectors:

$$\frac{1}{2} \frac{1}{2} \frac{1}$$

3. Orthogonalize w.r.t. the two previous Krylov vectors:

$$|V_{n+1}^{\perp}\rangle := |\tilde{V}_{n+1}\rangle - \sum_{j=n-1}^{N} |v_{j}\rangle \langle v_{j}|\tilde{v}_{n+1}\rangle \tag{46}$$

Orthogonalize again, now w.r.t. all previous Krylov vectors (to minimize accumulation of errors, "ghosts"):

$$|U_{n+1}^{\perp \perp}\rangle := |U_{n+1}^{\perp}\rangle - \sum_{j=0}^{N} |U_{j}\rangle \langle U_{j}|U_{n+1}^{\perp}\rangle \tag{43}$$

4. Compute norm and normalize:
$$b_{n+1} := \sqrt{\langle v_{n+1}^{\perp} | v_{n+1}^{\perp} \rangle}$$
 (4.3)

If
$$b_{n+1} \neq 0$$
, then $| \mathbf{v}_{n+1} \rangle := | \mathbf{v}_{n+1}^{\perp \perp} \rangle / b_{n+1}$ (49)

else, pick $|v_{n+1}\rangle$ as arbitrary normalized vector orthogonal to all $|v_b\rangle$..., $|v_n\rangle$

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

In the resulting Krylov space
$$K_N = \text{Span} \{ | v_0 \rangle, | v_1 \rangle, ..., | v_N \rangle \}$$
 (56)

the Hamiltonian has the representation

$$H_{N} = \begin{pmatrix} a_{0} & b_{1} \\ b_{1} & a_{1} & b_{2} \\ b_{2} & a_{2} \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & \\ & & & \\ &$$

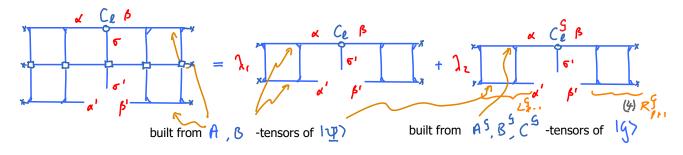
DMRG.3

Suppose we have an MPS representation for ground state, $g = \frac{A3}{7} + \frac{A3}{7} + \frac{C3}{7} + \frac{B3}{7} + \frac{B3$

Extremize:
$$\langle \Psi | H | \Psi \rangle - \frac{1}{4} \langle \Psi | \Psi \rangle - \frac{1}{4} \langle \Psi | \Psi \rangle$$
 (2)

Lagrange multipliers enforce
$$\langle \mathcal{I} | \mathcal{I} \rangle = 1$$
 and $\langle \mathcal{I} | \mathcal{I} \rangle = 0$. (3)

Extremization w.r.t. C yields



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

$$H_{\ell}^{(i)}C_{\ell} = \frac{\lambda}{\lambda}C_{\ell} + \frac{\lambda}{\lambda}C_{\ell}^{5}$$
 (5) with $C_{\ell}^{+}C_{\ell} = 1$, $C_{\ell}^{+}C_{\ell}^{(3)} = 0$ (6) cf. (DMRG-I.1.7)

Displaying indices: $\alpha' = (\alpha', 6', \beta')$ = ground state wave-function in local basis

$$\left(H_{\ell}^{(r)}\right)^{\alpha'}_{\alpha}\left[C_{\ell}\right]^{\alpha} = \lambda, \left[C_{\ell}\right]^{\alpha'}_{\alpha}, \lambda_{z}\left[C_{\ell}^{g}\right]^{\alpha'}_{\alpha} = 1, \left[C_{\ell}^{e}\right]_{\alpha}\left[C_{\ell}^{g}\right]^{\alpha} = 1$$

$$\begin{cases}
\left(C_{\ell}^{S}\right)^{\alpha'} = \left(C_{\ell}^{S}\right)^{\alpha'} \beta' = \left(C_{\ell}^{S}\right)^{\alpha'} \beta' \left(C_{\ell}^{S}\right)^{\alpha$$

Index-free notation for (5):
$$H(c) = \lambda_1(c) + \lambda_2(g)$$
, $(c(g) = 0)$

Projector onto subspace orthogonal to
$$\left| \mathcal{G} \right\rangle$$
: $\left| \frac{\overline{P}_{G}}{S} \right| = 1 - \left| \mathcal{G} \right\rangle \left| \mathcal{G} \right\rangle = 0$ (11)

[with indices:
$$\frac{\overline{R}_{3}}{R_{3}} = \frac{1}{a} = \frac{1}{a}$$

with indices:
$$\overline{P_g}^{\alpha'} = 1^{\alpha'} - g^{\alpha'} g^{\dagger}$$
, so that $\overline{P_g}^{\alpha'} = 0$ (12)

Project (10) onto this subspace:
$$\overline{P_g} + \overline{P_g} + \overline{P_$$

$$\overline{P_{g}} H \overline{P_{g}} C) = \lambda_{i} P_{g} C$$

$$= 0$$

$$(14)$$

This is simply an eigenvalue problem, for $\frac{\bar{P}_q}{4}$, in subspace orthogonal to $\frac{\bar{Q}}{4}$. It can be solved using straightforward generalization of Lanczos scheme, using Krylov subspace orthogonal to Given an arbitrary initial state $|v_0\rangle$, project it onto orthogonal subspace, $|v_0\rangle$ = $|v_0\rangle$ (15) and construct new Krylov vectors using

Why not simply use excited states in $\[\] \[? \]$ Because numerical noise can cause the to be not exactly orthogonal, hence for $\] \[\] \[\$ (17)

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 ${}^{}$ ${}^{}$, as indicated in (15).

Block-Lanczos for excited states

Standard Lanczos: represent action of H as

$$H(v_0) = (v_0) a_0 + (v_1) b_1 \Rightarrow$$

Block-Lanczos: start with set of

✓ orthogonal vectors,

$$v_{0,1}$$
, $v_{0,1}$, and represent action of H as (19)

$$H(v_{0,i}) = |v_{0,j}\rangle \frac{1}{2} (\alpha_0)^{j} + |v_{i,j}\rangle (b_i)^{j}; \qquad (20)$$

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

and
$$(\alpha_0 \dot{\gamma}_i = \langle \sigma_0, j | H | \sigma_0, i \rangle, (b, \dot{\beta}_i) = \langle \sigma_1, j | H | \sigma_0, i \rangle$$
 (22)

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

$$\begin{cases}
 a_{\circ} \mid b_{1} \\
 b_{1} \mid a_{1} \mid b_{2} \\
 b_{2} \mid a_{1}
\end{cases}$$
(23)

bi ai bz

(18)

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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.

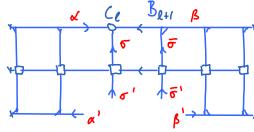
Represent MPS in site-canonical two-site basis:

14) = (α) (σε)(σε+) (β) (ς) «σε (Βε+) «σε β

14) = (α) (σε)(σε+) (β) (ς) «σε (Βε+) «σε β (σε) (σε

Then extremize simultaneously w.r.t.

$$\frac{\partial}{\partial S_{\ell+1}^{+}} \frac{\partial C_{\ell}^{+}}{\partial C_{\ell}^{+}} \left[\langle \vec{q} | \vec{H} | \vec{q} \rangle - \lambda \langle \vec{q} | \vec{q} \rangle \right] = 0$$



$$= \lambda \int_{\alpha'}^{\alpha'} \left(e^{-\frac{1}{2} \frac{1}{2} \frac{1}{2}} \right) \int_{\alpha'}^{\alpha'} \left(e^{-\frac{1}{2} \frac{1}{2}} \right) \int_{\alpha'}^{\alpha'} \left(e^{-\frac{1}{2} \frac{1}{2} \frac{1}{2}} \right) \int_{\alpha'}^{\alpha'} \left(e^{-\frac{1}{2} \frac{1}{2} \frac{1}{2}} \right) \int_{\alpha'}^{\alpha'} \left(e^{-\frac{1}{2} \frac{1}{2} \frac{1}{2}} \right) \int_{\alpha'}^{\alpha'} \left(e^{-\frac{1}{2} \frac{1}{2}} \right) \int$$

(4) close zippers from left and right

Compact notation:

$$\left[H_{\ell}^{(2)}\right]^{\alpha'} \left[\Psi_{\ell}^{(1)}\right]^{\alpha} = \lambda \left[\Psi_{\ell}^{(1)}\right]^{\alpha'} \quad \text{with composite index } \alpha = (0, 6, 6, \beta)$$

Use Lanczos to find lowest eigenvalue of eigenvalue equation (5), and reshape updated $\psi_{\ell}^{(\epsilon)}$:

updated

Key point: S has DI singular values, <u>larger</u> than the virtual bond dimension D of C_I and C_I Hence, it explores a larger state space, in general also including more symmetry sectors!

Truncate down to \bigcirc and reshape: This get rid of 'bad' symmetry sectors.

This concludes optimization of site ℓ . Now move one site to the right and repeat. Sweep back and forth until convergence of full chain (i.e. ground state energy converges).

Cost of 1-site DMRG: $\mathcal{O}(D^3 dw + D^2 d^2w^2)$ Cost of 2-site DMRG: $\mathcal{O}(D^3 d^3 + D^3 d^2w)$ (1)