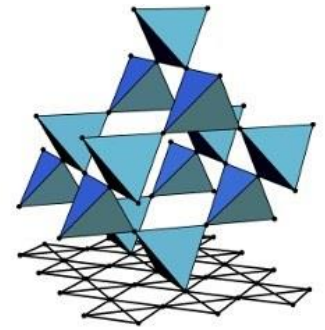




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SFB 1143

# Tensor Networks (SS2021)

## Lecture 7: MPO and single-site DMRG

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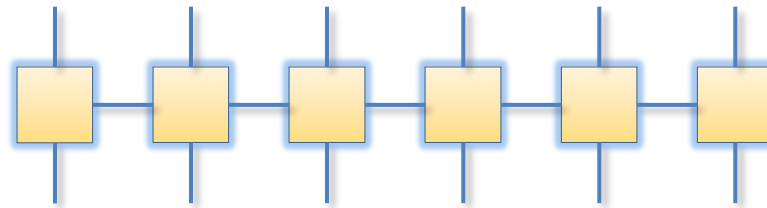
Zoom: [tuhonghao@gmail.com](mailto:tuhonghao@gmail.com)

May 10<sup>th</sup>, 2021

## § 2.1 Matrix Product Operator

Matrix Product Operator:

$$W = \sum_{s'_1, s_1, \dots, s'_N, s_N} \sum_{\alpha_1, \dots, \alpha_{N-1}} B_{\alpha_1}^{s'_1, s_1} B_{\alpha_1 \alpha_2}^{s'_2, s_2} \dots B_{\alpha_{N-1}}^{s'_N, s_N} |s'_1, \dots, s'_N\rangle \langle s_1, \dots, s_N|$$



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$O_{\alpha_{i-1}, \alpha_i} = \sum_{s'_i, s_i} B_{\alpha_{i-1} \alpha_i}^{s'_i, s_i} |s'_i\rangle \langle s_i|$

- Product of matrices whose matrix entries are **local** operators.

## § 2.1 Matrix Product Operator

Example:

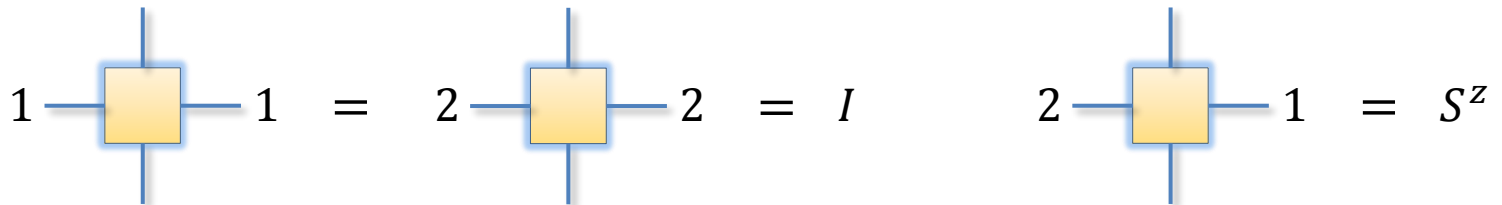
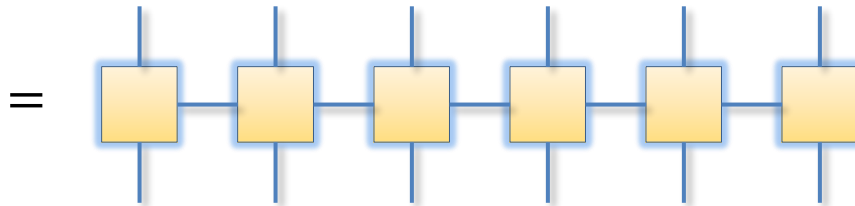
$$\begin{aligned} \sum_{j=1}^N S_j^Z &= \begin{pmatrix} 0 & 1 \end{pmatrix} \begin{pmatrix} I & 0 \\ S_1^Z & I \end{pmatrix} \begin{pmatrix} I & 0 \\ S_2^Z & I \end{pmatrix} \cdots \begin{pmatrix} I & 0 \\ S_{N-1}^Z & I \end{pmatrix} \begin{pmatrix} I & 0 \\ S_N^Z & I \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} S_1^Z & I \end{pmatrix} \begin{pmatrix} I & 0 \\ S_2^Z & I \end{pmatrix} \cdots \begin{pmatrix} I & 0 \\ S_{N-1}^Z & I \end{pmatrix} \begin{pmatrix} I \\ S_N^Z \end{pmatrix} \end{aligned}$$

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Example:

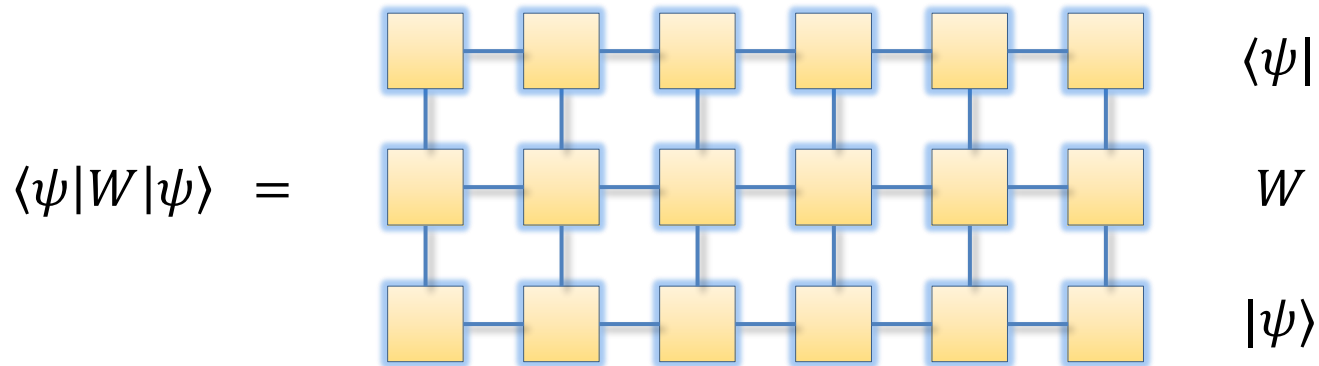
$$\sum_{j=1}^N S_j^Z = (0 \quad 1) \begin{pmatrix} I & 0 \\ S_1^Z & I \end{pmatrix} \begin{pmatrix} I & 0 \\ S_2^Z & I \end{pmatrix} \cdots \begin{pmatrix} I & 0 \\ S_{N-1}^Z & I \end{pmatrix} \begin{pmatrix} I & 0 \\ S_N^Z & I \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$= (S_1^Z \quad I) \begin{pmatrix} I & 0 \\ S_2^Z & I \end{pmatrix} \cdots \begin{pmatrix} I & 0 \\ S_{N-1}^Z & I \end{pmatrix} \begin{pmatrix} I \\ S_N^Z \end{pmatrix}$$



## § 2.1 Matrix Product Operator

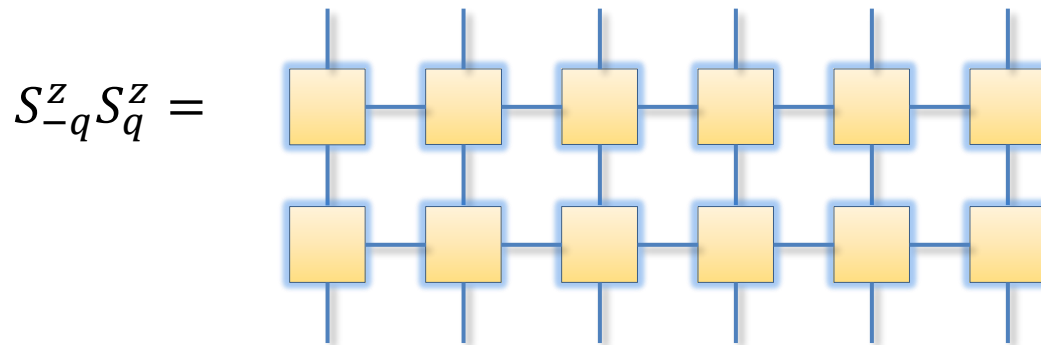
- Motivation: It's particularly convenient to compute **expectation values** (and matrix elements) of MPO **within MPS basis!**



## § 2.1 Matrix Product Operator

What is the MPO form for  $S_{-q}^z S_q^z$  ?

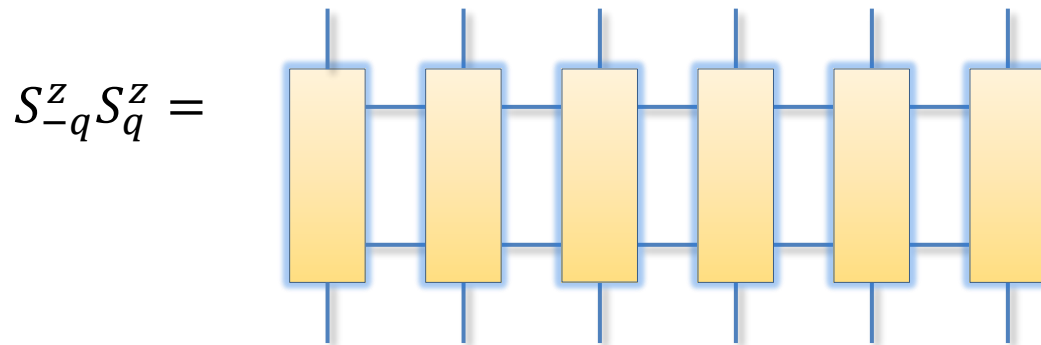
$$\begin{aligned} S_q^z &= \frac{1}{\sqrt{N}} \sum_{j=1}^N S_j^z e^{-iqj} \\ &= (0 \quad 1) \begin{pmatrix} I & 0 \\ S_1^z e^{-iq} & I \end{pmatrix} \begin{pmatrix} I & 0 \\ S_2^z e^{-2iq} & I \end{pmatrix} \cdots \begin{pmatrix} I & 0 \\ S_N^z e^{-iqN} & I \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \end{aligned}$$



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## § 2.1 Matrix Product Operator

$$\begin{aligned}
 H &= \sum_{i=1}^{N-1} \vec{S}_i \cdot \vec{S}_{i+1} \\
 &= \left( 0 \quad \frac{1}{2} S_1^- \quad \frac{1}{2} S_1^+ \quad S_1^Z \quad I \right) \cdots \begin{pmatrix} I & 0 & 0 & 0 & 0 \\ S_i^+ & 0 & 0 & 0 & 0 \\ S_i^- & 0 & 0 & 0 & 0 \\ S_i^Z & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{2} S_i^- & \frac{1}{2} S_i^+ & S_i^Z & I \end{pmatrix} \cdots \begin{pmatrix} I \\ S_N^+ \\ S_N^- \\ S_N^Z \\ 0 \end{pmatrix}
 \end{aligned}$$

## § 2.1 Matrix Product Operator

- How to obtain the MPO representation for a given operator?
  - Construction by hand: inspection + addition of the **simplest** ones


$$W = W_1 + W_2$$


$$O = \begin{pmatrix} O_1 & \\ & O_2 \end{pmatrix}$$

## § 2.1 Matrix Product Operator

- How to obtain the MPO representation for a given operator?
  - Construction by hand: inspection + addition of the **simplest** ones

$$W = W_1 + W_2$$


$$O = \begin{pmatrix} O_1 & \\ & O_2 \end{pmatrix}$$

- Automatic generation: see [this link](#) (“AutoMPO” in iTensor) and [this paper](#) [Hubig *et al.*, Phys. Rev. B 95, 035129 (2017)].

## § 2.2 Single-site DMRG

- For a given Hamiltonian, we want to find the best variational ansatz within the family of MPS:

$$E(A) = \min_{\{A\}} \frac{\langle \psi(A) | H | \psi(A) \rangle}{\langle \psi(A) | \psi(A) \rangle}$$

- This is a **highly nonlinear** optimization problem, which is in general rather difficult.

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- This is a **highly nonlinear** optimization problem, which is in general rather difficult.
- **Spirit of DMRG**: optimize  $A_{\alpha_i \beta_i}^{S_i}$  site by site (**locally optimal**)

S. R. White, Phys. Rev. Lett. 69, 2863 (1992);

U. Schollwöck, Ann. Phys 326, 96 (2011); C. Hubig, PhD Thesis at LMU (2017).

## § 2.2 Single-site DMRG

- Define energy functional:

$$E[A, \bar{A}, \lambda] = \langle \psi(A) | H | \psi(A) \rangle - \lambda (\langle \psi(A) | \psi(A) \rangle - 1)$$

↖  
Lagrange multiplier

- Energy minimization (with respect to site  $i$ ):

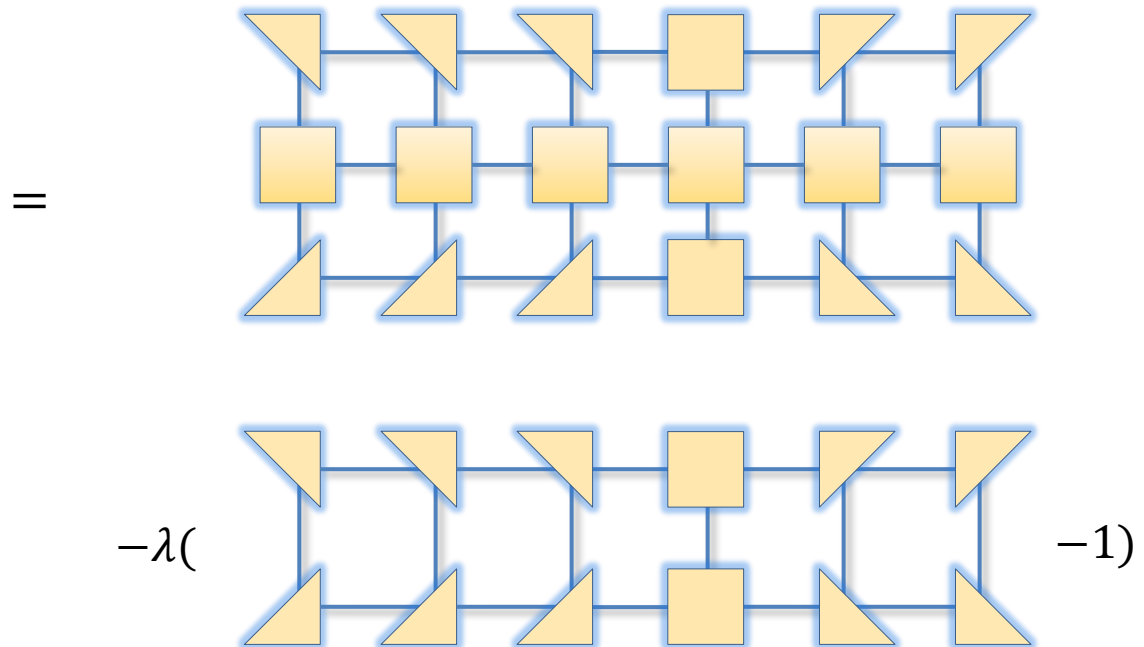
$$\frac{\partial}{\partial \bar{A}_{\alpha_i \beta_i}^{s_i}} E[A, \bar{A}, \lambda] = \frac{\partial}{\partial A_{\alpha_i \beta_i}^{s_i}} E[A, \bar{A}, \lambda] = 0$$

$$\frac{\partial}{\partial \lambda} E[A, \bar{A}, \lambda] = 0$$

## § 2.2 Single-site DMRG

- This is **best** done with **site** canonical form (for site  $i$ ):

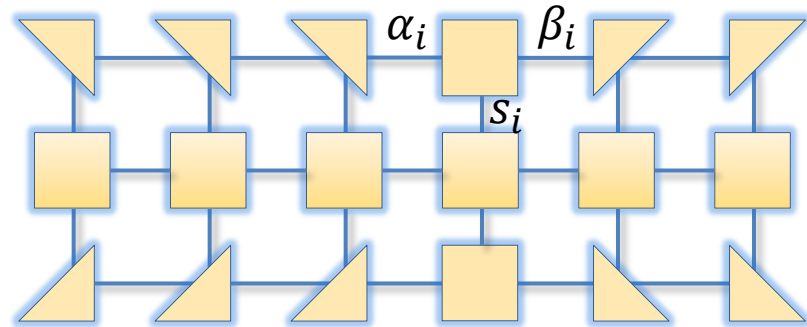
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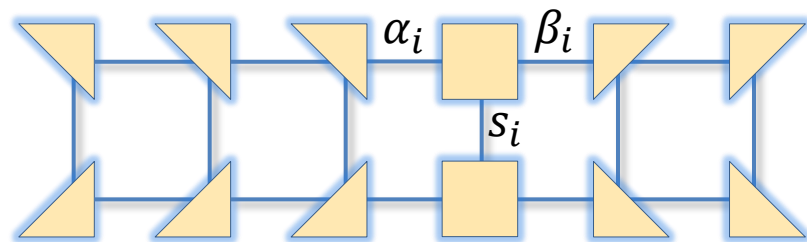
## § 2.2 Single-site DMRG

- This is **best** done with **site** canonical form (for site  $i$ ):

$$\frac{\partial}{\partial \bar{A}_{\alpha_i \beta_i}^{s_i}} E[A, \bar{A}, \lambda] = \frac{\partial}{\partial \bar{A}_{\alpha_i \beta_i}^{s_i}}$$



$$-\lambda \frac{\partial}{\partial \bar{A}_{\alpha_i \beta_i}^{s_i}}$$

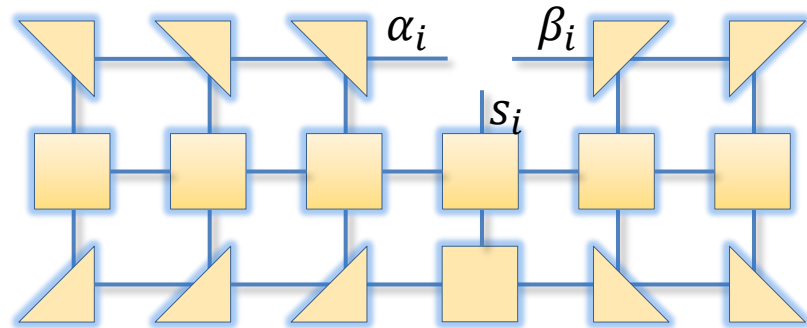




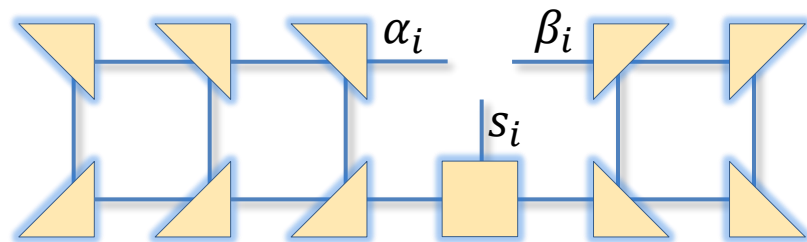
## § 2.2 Single-site DMRG

- This is **best** done with **site** canonical form (for site  $i$ ):

$$\frac{\partial}{\partial \bar{A}_{\alpha_i \beta_i}^{s_i}} E[A, \bar{A}, \lambda] =$$



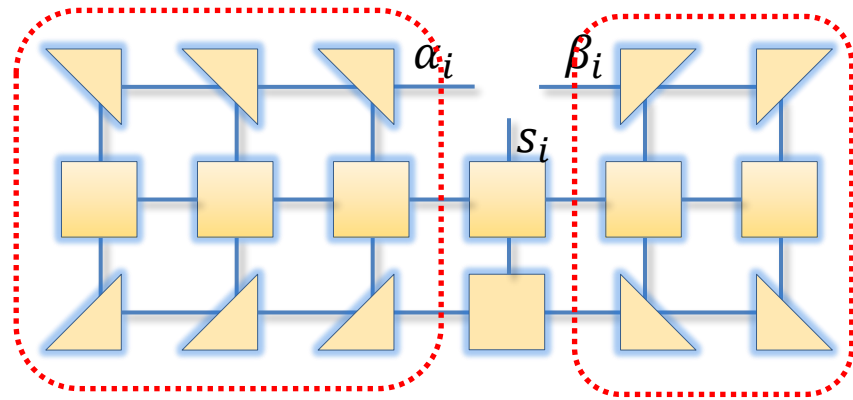
$-\lambda$



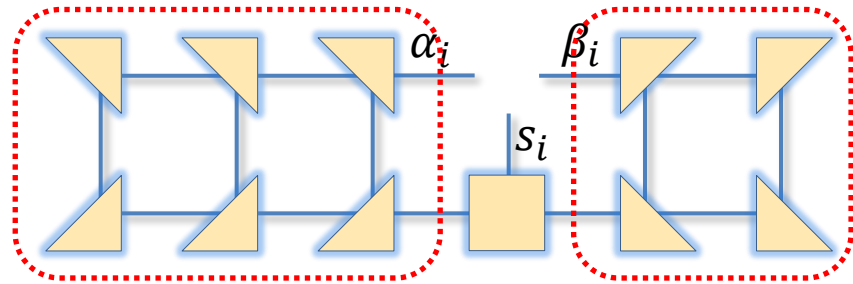
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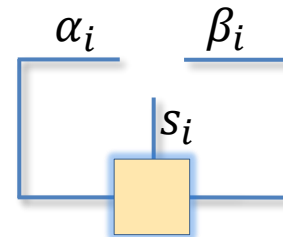
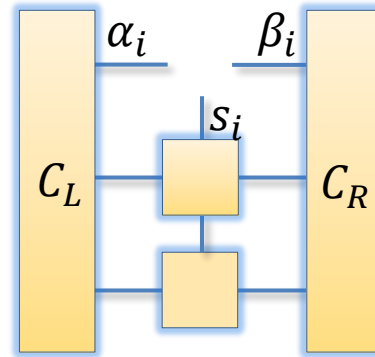


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## § 2.2 Single-site DMRG

$$\frac{\partial}{\partial \bar{A}_{\alpha_i \beta_i}^{s_i}} E[A, \bar{A}, \lambda] =$$

Effective Hamiltonian:

$$(H_{\text{eff}})_{(\alpha_i \beta_i s_i); (\alpha'_i \beta'_i s'_i)} =$$

## § 2.2 Single-site DMRG

$$\frac{\partial}{\partial \bar{A}_{\alpha_i \beta_i}^{s_i}} E[A, \bar{A}, \lambda] =$$

Standard eigenvalue problem!

$$\sum_{\alpha'_i \beta'_i s'_i} (H_{\text{eff}})_{(\alpha_i \beta_i s_i); (\alpha'_i \beta'_i s'_i)} A_{\alpha'_i \beta'_i s'_i} = \lambda A_{\alpha_i \beta_i s_i} \quad \Rightarrow \quad H_{\text{eff}} A = \lambda A$$

## § 2.2 Single-site DMRG

Standard eigenvalue problem (at site  $i$ ):

$$H_{\text{eff}}A = \lambda A$$

$$A^\dagger A = 1$$



- Keep the lowest eigenvalue  $\lambda_{\min}$  and its corresponding (normalized) eigenvector  $A_{\alpha_i \beta_i}^{s_i}$ .

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variational ground-state energy  
(see next slide)

This ensures normalization:

$$\frac{\partial}{\partial \lambda} E[A, \bar{A}, \lambda] = \langle \psi | \psi \rangle - 1 = 0$$

## § 2.2 Single-site DMRG

Standard eigenvalue problem (at site  $i$ ):

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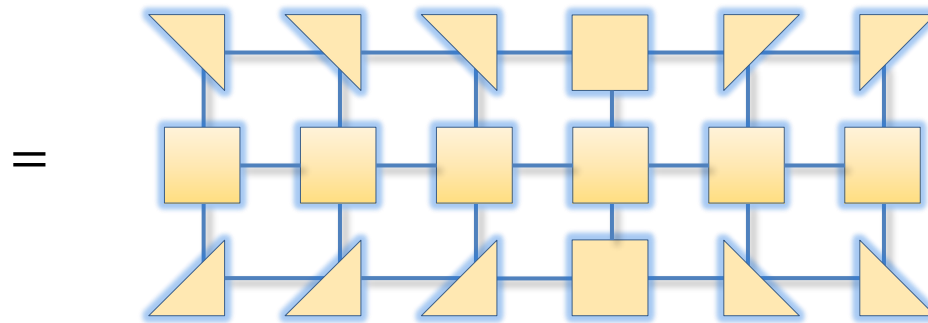
- Use **Lanczos algorithm** (next lecture), since  $H_{\text{eff}}$  is a large matrix and only its lowest eigenvalue/eigenvector is needed.



## § 2.2 Single-site DMRG

Substitute the **obtained solution**  $A_{\alpha_i\beta_i}^{s_i}$  into the energy functional:

$$E[A, \bar{A}, \lambda] = \langle \psi(A) | H | \psi(A) \rangle - \lambda (\langle \psi(A) | \psi(A) \rangle - 1)$$



$$= \lambda_{\min} \quad (\text{So } \lambda_{\min} \text{ is already the variational energy!})$$

## § 2.2 Single-site DMRG

Workflow:

- Generate a **random** MPS (to initialize DMRG).
- Start with site 1: bring the MPS into **site canonical form (at site 1)**, optimize the MPS tensor at site 1 (by diagonalizing the effective Hamiltonian), save the variational energy (to compare with the variational energies in the next steps), use the obtained MPS tensor to **calculate/update the environment tensors  $C_L$  and  $C_R$  (this saves computing time)**.
- Go to site 2, repeat the above steps for site 1.
- Do the same for site 3, site 4, ... , site  $N$ , site  $N - 1$ , ... , site 1, ...
- Sweep a few times, until the variational energy is converged.