

# Matrix Product States: Basic properties

Matrix Product State (open boundary conditions (OBC));  
1D chain of  $N$   $d$ -level sytks:

$$|\psi\rangle = \sum_{i_1, \dots, i_N} A^{[1]i_1} A^{[2]i_2} \dots A^{[N]i_N} |i_1, i_2, \dots, i_N\rangle$$

with  $A^{[1]i_1}$   $1 \times D$  matrix (vector)

$A^{[s]i_s}$   $D \times D$  matrix,  $s = 2, \dots, N-1$

$A^{[N]i_N}$   $D \times 1$  matrix

$D$ : virtual dimension, bond dimension

Last lecture: Every state  $|\psi\rangle$  has an MPS representation with  
 $D = \max_k$  Schmidt rank of  $|\psi\rangle$  across any cut  $1 \dots k | k+1 \dots N$ .

General state: Schmidt rank  $D \sim d^{N/2} \Rightarrow$

$\Rightarrow$  # of parameters  $\sim NdD^2 = Nd^{2N+1}/2$  : exp. in  $N$ !

$\Rightarrow$  no improvement over  $|\psi\rangle = \sum c_{i_1 \dots i_N} |i_1 \dots i_N\rangle$  ( $d^N$  parameters)

State with bounded Schmidt rank:

# of parameters  $\sim NdD^2$ ,  $D$  const. : linear in  $N$ !

$\Rightarrow$  exponential improvement over  $c_{i_1 \dots i_N}$

Will see later: low entanglement (area law)  $\rightarrow$  good MPS approximation to  $|\psi\rangle$  with low D exists.

MPS: Efficient approximation of states with low entanglement,  
 (in particular ground states of gapped local Hamiltonians!)

Generalization: MPS with periodic boundary conditions (PBC):

$$|\psi\rangle = \sum_{i_1, \dots, i_N} \text{tr} [A^{[1]i_1} A^{[2]i_2} \dots A^{[N]i_N}] |i_1, \dots, i_N\rangle$$

Graphical notation for MPS:

$$|\psi\rangle = \sum_{i_1, \dots, i_N} c_{i_1, \dots, i_N} |i_1, \dots, i_N\rangle$$

$c_{i_1, \dots, i_N}$ : Tensor w/  $N$  indices (very large!)

MPS: Express  $c_{i_1, \dots, i_N}$  as summation ("contraction") over a set of simple tensors.

Graphical notation:

Tensor = "box w/ legs":

$$c_{i_1, \dots, i_N} = \begin{array}{c} i_1 \quad i_2 \quad \dots \quad i_N \\ | \quad | \quad \dots \quad | \\ \hline c \end{array}$$

$$A_{\alpha\beta}^{[s]i_s} = \begin{array}{c} i_s \\ | \\ \hline \alpha \quad [A^{[s]}] \quad \beta \end{array}$$

Summation (contraction) = connect legs:

$$\sum_{\alpha, \beta} A_{\alpha\beta}^{[s]i_s} A_{\beta\gamma}^{[s+1]i_{s+1}} = \begin{array}{c} i_s \quad i_{s+1} \\ | \quad | \\ \boxed{A^{[s]}} \text{---} \boxed{A^{[s+1]}} \\ \alpha \quad \beta \quad \gamma \end{array}$$

MPS (OBC):

$$\begin{array}{c} i_1 \quad i_2 \quad \dots \quad i_N \\ | \quad | \quad \dots \quad | \\ \boxed{c} \end{array} = \begin{array}{c} i_1 \quad i_2 \quad \dots \quad i_{N-1} \quad i_N \\ | \quad | \quad \dots \quad | \quad | \\ \boxed{A^{[1]}} \text{---} \boxed{A^{[2]}} \text{---} \dots \text{---} \boxed{A^{[N-1]}} \text{---} \boxed{A^{[N]}} \end{array}$$

MPS (PBC):

$$\begin{array}{c} i_1 \quad i_2 \quad \dots \quad i_N \\ | \quad | \quad \dots \quad | \\ \boxed{c} \end{array} = \begin{array}{c} i_1 \quad i_2 \quad \dots \quad i_N \\ | \quad | \quad \dots \quad | \\ \boxed{A^{[1]}} \text{---} \boxed{A^{[2]}} \text{---} \dots \text{---} \boxed{A^{[N]}} \\ \text{---} \text{---} \text{---} \text{---} \end{array}$$

Re-express  $c_{i_1 \dots i_N}$  as Tensor Network  $\rightarrow$  "Tensor Network States".

$\rightarrow$  Structure of TN reproduces locality structure of sysk!

## Examples:

78

Ex. 1: GHZ:  $\frac{1}{\sqrt{2}} (|0\dots 0\rangle + |1\dots 1\rangle)$

PBC:  $A^{[s],0} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} = |0\rangle\langle 0|$ ;  $A^{[s],1} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = |1\rangle\langle 1|$

$\rightarrow A^{[s],i_s} \cdot A^{[s+1],i_{s+1}} = \delta_{i_s, i_{s+1}} A^{[s+1],i_{s+1}}$  etc.

$\rightarrow \text{tr} [A^{[1],i_1} \dots A^{[N],i_N}] = \delta_{i_1 \dots i_N}$

Note: • MPS is not normalized.

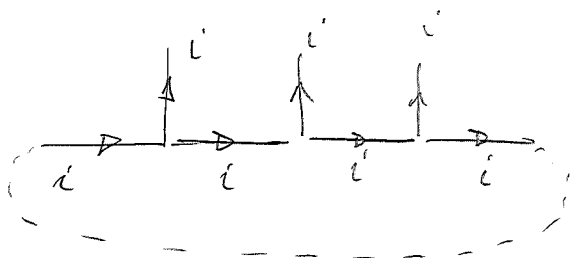
• We can normalize by e.g. choosing  $A^{[1],i} = \frac{1}{\sqrt{2}} (\dots)$ .

• Normalization breaks transl. invariance (or dep. on  $N$ ).

Graphical:

$\alpha \text{---} \square \text{---} \beta \equiv \delta_{\alpha\beta}^i$

"Algorithmic" interpretation: Tensor "transports" information on virtual index from left to right & copies it to physical index.



In particular: Map from  $\alpha$  to  $(i, j)$  is isometry, can be (79)  
 implemented deterministically  $\rightarrow$  preparation scheme (if we had OBC)

Can we work GHZ w/ OBC?

Yes:

$$\rightarrow A^{[1],0} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \langle 0 | = \langle + | A^{[s],0} \quad (\langle + | = (\langle 0 | + \langle 1 |))$$

$$A^{[1],1} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \langle 1 | = \langle + | A^{[s],1}$$

$$A^{[N],0} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = |0\rangle = A^{[s],0} |+\rangle$$

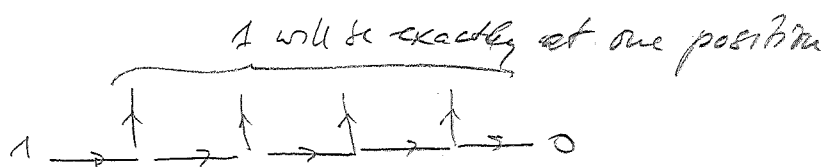
$$A^{[N],1} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} = |1\rangle = A^{[s],1} |+\rangle$$

Example 2: W State

$$|W\rangle \propto |100\dots\rangle + |010\dots\rangle + |0010\dots\rangle + \dots + |0\dots 01\rangle$$



Start with 1, end with 0:



$$A^{[1],0} = |0\rangle\langle 0| + |1\rangle\langle 1| = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \mathbb{1}$$

$$A^{[1],1} = |1\rangle\langle 0| = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} = \sigma^+$$

$$A^{[1],i} = \langle 1|A^{[i],i} ; \quad A^{[N],i} = A^{[1],i} |0\rangle$$

$$\Rightarrow c_{i_1 \dots i_N} = \langle 1 | \underbrace{A^{[1],i_1} A^{[2],i_2} \dots A^{[N],i_N}}_{\text{each: } \mathbb{1} \text{ or } \sigma^+, \text{ and } \sigma^+ \sigma^+ = 0} |0\rangle$$

$\Rightarrow$  exactly one  $\sigma^+$   $\Rightarrow$  W state!

Note: No PBC representation of W state exists (unless D grows w/N).

AKLT state:  $d=3$  (spin-1),  $D=2$ .  $\rightarrow$  (states:  $|+1\rangle, |0\rangle, |-1\rangle$ )

$$A^{[1],+1} = \sqrt{2}\sigma^+ ; \quad A^{[1],-1} = -\sqrt{2}\sigma^- ; \quad A^{[1],0} = \sigma_z$$

Ground state:  $\Sigma$  of  $|(+1)0\dots 0(-1)0\dots 0(+1)0\dots 0(-1)\dots\rangle$

- $\rightarrow$  alternating  $(+1)$  and  $(-1)$  w/ arb. # of 0 in between!
- $\rightarrow$  sign dep. on # of zeros!
- $\rightarrow$  some type of long-range order/information?

- AKLT:
- o First example of topological phase.
  - o Gives interesting solvable model.
  - o Will meet it again later.

# Properties of Matrix Product States:

81

## ① Normalization:

$$|\psi\rangle = \sum_{i_1, \dots, i_N} c_{i_1, \dots, i_N} |i_1, \dots, i_N\rangle$$

$$\langle \psi | \psi \rangle = \sum_{\substack{i_1, \dots, i_N \\ j_1, \dots, j_N}} \langle j_1, \dots, j_N | \overline{c_{j_1, \dots, j_N}} c_{i_1, \dots, i_N} |i_1, \dots, i_N\rangle$$

$\delta_{i_1 j_1, \dots, i_N j_N}$

$$= \sum_{i_1, \dots, i_N} c_{i_1, \dots, i_N} \overline{c_{i_1, \dots, i_N}}$$

$$= \sum_{i_1, \dots, i_N} \text{tr} [A^{[1]i_1}, A^{[2]i_2}, \dots] \cdot \text{tr} [\overline{A}^{[1]i_1}, \overline{A}^{[2]i_2}, \dots]$$

NB:  $\text{tr}[A] \text{tr}[B] = \text{tr}[A \otimes B]$ ,  
 $(AC) \otimes (BD) = (AB) \otimes (CD)$

$$= \sum_{i_1, \dots, i_N} \text{tr} \left[ \left( A^{[1]i_1} \otimes \overline{A}^{[1]i_1} \right) \cdot \left( A^{[2]i_2} \otimes \overline{A}^{[2]i_2} \right) \cdot \dots \right]$$

$$= \text{tr} \left[ \left( \sum_{i_1} A^{[1]i_1} \otimes \overline{A}^{[1]i_1} \right) \cdot \dots \right]$$

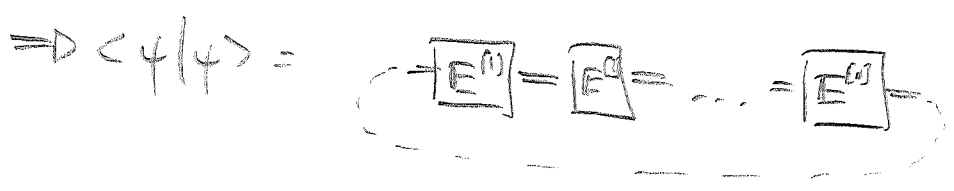
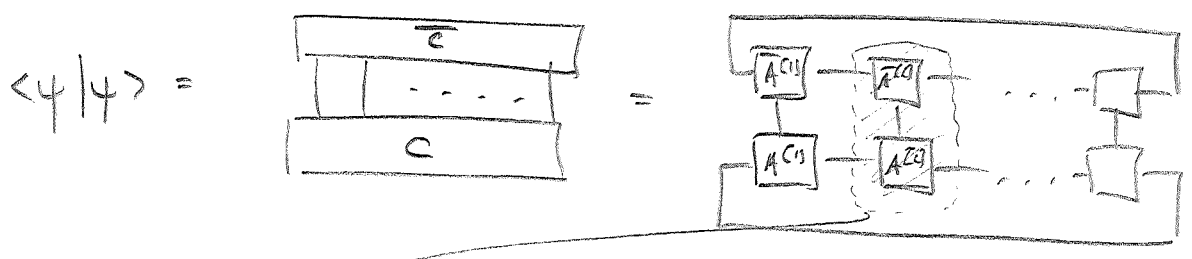
Define  $E^{[s]} := \sum_i A^{[s]i} \otimes \overline{A}^{[s]i}$  : "transfer operator"

$$\Rightarrow \langle \psi | \psi \rangle = \text{tr} [E^{[1]}, E^{[2]}, \dots, E^{[N]}]$$

$\Rightarrow \langle \psi | \psi \rangle \equiv$  matrix multiplication of  $D^2 \times D^2$ -matrices

$\Rightarrow$  efficient (as compared to  $\sum_{i_1, \dots, i_N} |c_{i_1, \dots, i_N}|^2$ ).

Alternative graphical derivation:



② Expectation values of local observables

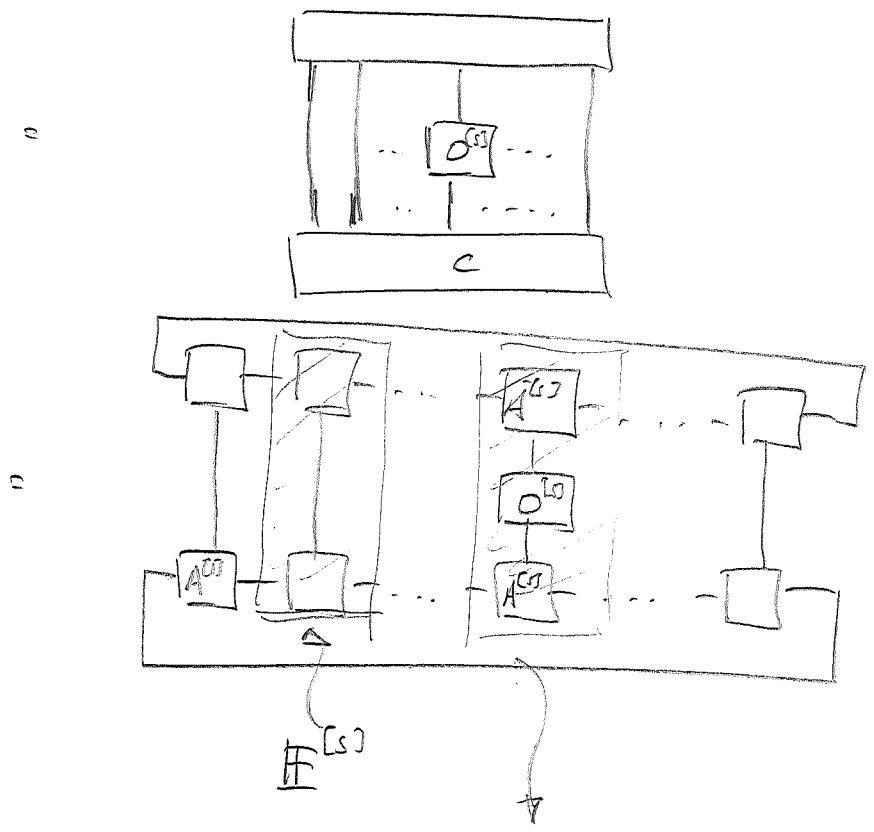
(for simplicity: single-site observables.

Otherwise: block sites, or write  $O = \sum A^i B^i$ )

$$\begin{aligned} \langle \psi | O^{[s]} | \psi \rangle &= \sum \langle j_1 - j_N | \bar{c}_{j_1 - j_N} \cdot O^{[s]} \cdot c_{i_1 - i_N} | i_1 - i_N \rangle \\ &= \sum \underbrace{\langle j_1 - j_N | O^{[s]} | i_1 - j_N \rangle}_{\prod_{k \neq s} \delta_{i_k j_k}} c_{i_1 - i_N} \bar{c}_{j_1 - j_N} \\ &= \langle j^s | O^{[s]} | i^s \rangle \prod_{k \neq s} \delta_{i_k j_k} \end{aligned}$$



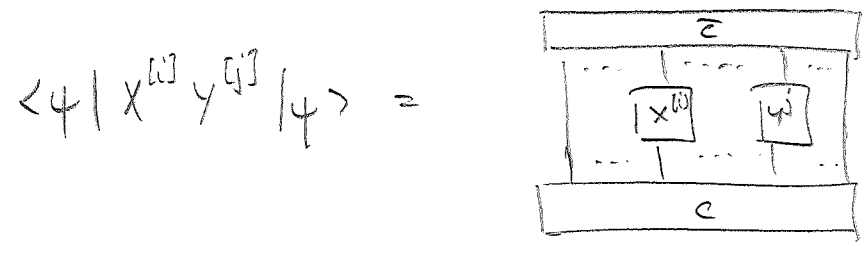
$$= \sum_{\substack{i_1 \dots i_N \\ j_s}} c_{i_1 \dots i_N} \overline{c_{i_1 \dots i_N}} \langle j_s | 0^{[s]} | i_s \rangle.$$

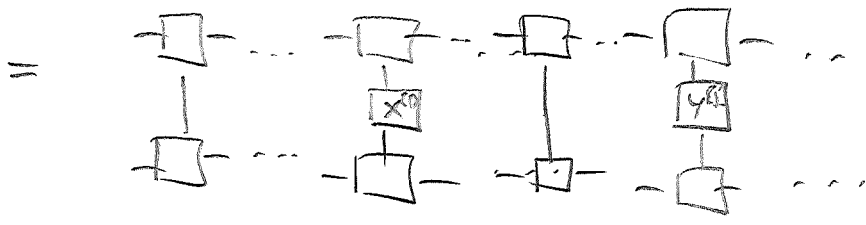


Define  $E_0^{[s]} = \sum_{i,j} A^{[s]i} \overline{A^{[s]j}} \langle j | 0^{[s]} | i \rangle$

$$\Rightarrow \langle \psi | 0 | \psi \rangle = \text{tr} [ E^{[1]} E^{[2]} \dots E_0^{[s]} E^{[s+1]} \dots E^{[N]} ]$$

③ Correlation functions:

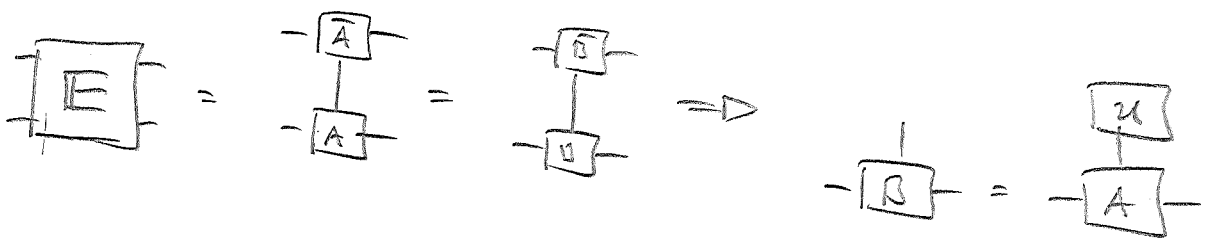




$$= \text{tr} [ E^{[0]} E^{[2]} \dots E_x^{[i]} \dots E_y^{[j]} E^{[k+1]} \dots ]$$

⇒ all information about correlation function etc. contained in transfer operator.

In fact: Transfer op.  $E^{[s]}$  determines state uniquely up to local basis transformation (i.e., same choice)



⇒ all non-local properties contained in  $E$ !

(Proof: Homework!)

## Scaling of correlations in MPS

(85)

Consider transl. invariant MPS (PBC), i.e.  $A^{[L]} \equiv A$ .

How do correlation functions scale?

$$c(i,j) = \frac{\langle \psi | X^{[i]} Y^{[j]} | \psi \rangle}{\langle \psi | \psi \rangle} \quad \text{as function of } i-j?$$

$$c(i,j) = \frac{\text{tr} \left[ \overbrace{E \dots E}^{i-1} E_x \overbrace{E \dots E}^{j-i-1} E_y \overbrace{E \dots E}^{N-j} \right]}{\text{tr} \left[ \underbrace{E \dots E}_N \right]}$$

$$= \frac{\text{tr} \left[ E_x \cdot E^{j-i-1} \cdot E_y \cdot E^{N+i-j-1} \right]}{\text{tr} \left[ E^N \right]}$$

Assume  $E$  has no Jordan blocks ( $\rightarrow$  cf. HW):

$$E = \sum_{k=1}^D \lambda_k |r_k\rangle\langle l_k| \quad : \text{eigenvalue decomposition,} \\ \langle l_i | r_j \rangle = \delta_{ij}$$

wlog  $|\lambda_1| \geq |\lambda_2| \geq \dots$

Assume first  $|\lambda_1| > |\lambda_2| : \dots$

Then,  $E^N \xrightarrow{N \rightarrow \infty} \lambda_1^N |r_1\rangle\langle l_1|$  for  $N \rightarrow \infty$ .

$$\Rightarrow c(i,j) \approx \frac{\text{tr} [E_x E^{j-i+1} E_y \lambda_1^{N+i-j-1} |r_i\rangle\langle r_i|]}{\text{tr} [\lambda_1^N |r_i\rangle\langle r_i|]}$$

$$= \frac{\langle r_i | E_x E^{j-i+1} E_y | r_i \rangle}{\lambda_1^{j-i+1}}$$

We also know that  $E^{j-i+1} = \sum_k \lambda_k^{j-i+1} |r_k\rangle\langle r_k|$

$$\Rightarrow c(i,j) = \sum_{k=1}^{D^2} \left( \frac{\lambda_k}{\lambda_1} \right)^{j-i+1} \frac{\langle r_i | E_x | r_k \rangle \langle r_k | E_y | r_i \rangle}{\lambda_1^2}$$

$\Rightarrow$  all correlation functions decay exponentially.

Largest eigenvalue of  $E$  degenerate (upto phase,  $|\lambda_1| = |\lambda_2| = \dots$ ):

$$E^N \rightarrow \sum_{k=1}^K \lambda_k^N |r_k\rangle\langle r_k|$$

Sum over degen. eigenvalues

$$c(i,j) \approx \frac{\text{tr} [E_x \left( \sum_{k=1}^K \lambda_k^{j-i+1} |r_k\rangle\langle r_k| \right) E_y \left( \sum_{k=1}^K \lambda_k^{N+i-j-1} |r_k\rangle\langle r_k| \right)]}{\text{tr} \left[ \sum_{k=1}^K \lambda_k^N |r_k\rangle\langle r_k| \right]}$$

large N  
large  $|j-i|$

$$= \frac{\sum_{k,l=1}^K \lambda_k^{j-i-1} \lambda_l^{N-i-j-1} \langle e_l | E_x | r_k \rangle \langle e_k | E_y | r_l \rangle}{\sum_{k=1}^K \lambda_k^N}$$

⇒ This term is constant for  $|j-i| \rightarrow \infty$ ,  
 up to oscillations, if  $\arg(\lambda_k) \neq \arg(\lambda_l)$ .

Examples:

AKLT state:

$$A^{+1} = \sqrt{2} \sigma^+; \quad A^{-1} = \sqrt{2} \sigma^-; \quad A^0 = \sigma_z.$$

$$E = 2 \sigma^+ \otimes \sigma^+ + 2 \sigma^- \otimes \sigma^- + \sigma_z \otimes \sigma_z$$

$$= \begin{pmatrix} 1 & & & \\ & -1 & 0 & \\ & 0 & -1 & \\ 2 & & & 1 \end{pmatrix}.$$

$$\Rightarrow \lambda_1 = 3; \quad |r_1\rangle = |4\rangle = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix}; \quad |\lambda_2| = |\lambda_3| = |\lambda_4| = 1$$

$$\Rightarrow \text{Correlations decay as } \left(\frac{1}{3}\right)^L = e^{-L/\xi}$$

$$\Rightarrow \text{correlation length } \xi = -\frac{1}{\log(1/3)} \approx 0.910$$

GHZ state:

$$A^0 = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}; \quad A^1 = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

(88)

$$E = A^0 \otimes \bar{A}^0 + A^1 \otimes \bar{A}^1 = \begin{pmatrix} 1 & & & \\ & 0 & & \\ & & 0 & \\ & & & 1 \end{pmatrix}$$

$$\Rightarrow \lambda_1 = \lambda_2 = 1; \quad \lambda_3 = \lambda_4 = 0$$

$\Rightarrow$  long-range correlations!