

# How to make big Hilbert spaces small

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# Outline

- Motivation
- Compressing  $|\Psi\rangle$  with SVD
- Matrix Product States (MPS)
- Ground state calculation (roughly)

# Motivation

$N$  sites, each with spin-1/2. Find ground state of:

$$\mathcal{H} = -J \sum_{i=1}^N \sigma_i^z \sigma_{i+1}^z - h \sum_{i=1}^N \sigma_i^x$$

Hilbert space dimension:  $2^N$

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!! For many relevant Hamiltonians, haystack  $\ll$  full Hilbert space

e.g. haystack  $\sim$  subspace of states with low entanglement entropy

$\implies$  Clever parameterization + efficient algorithms =  $\odot$ ?

# Compressing $|\Psi\rangle$ ?

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- How to approximate  $|\Psi\rangle$  well without storing  $d^N$  coefficients?
- Possible to reduce entanglement entropy after approximation?

# Compressing $|\Psi\rangle$ with SVD

## Theorem (Singular value decomposition)

For any  $M$ , there are unitaries  $U, V$  for which  $M = USV^\dagger$ , with  $S = \text{diag}(s_1, s_2, \dots)$ .

$s_i$ : singular values of  $M \equiv$  eigenvalues of  $\sqrt{M^\dagger M}$ .  $s_i \geq 0$

## Theorem (Low-rank approximation)

The HS-distance from a rank- $m$  matrix  $A_{n \times n}$  to the nearest  $n \times n$  matrix of rank  $k \leq m$  is the square root of the sum of the squares of the smallest  $n - k$  singular values of  $A$ .

Hilbert-Schmidt norm:

$$\|A\|_{HS} = \sqrt{\sum |A_{ij}|^2} = \sqrt{\text{Tr}(A^\dagger A)} = \sqrt{\sum s_i^2}$$

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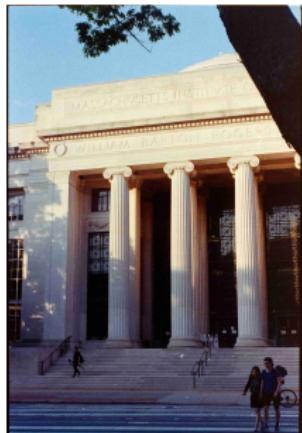
$$A = U \cdot \Sigma \cdot V^T \rightarrow A \approx U \cdot \Sigma \cdot V^T$$

The diagram illustrates the Singular Value Decomposition (SVD) of a matrix  $A$ . On the left,  $A$  is shown as a green rectangle. An equals sign follows, then  $U$  (a maroon rectangle), a dot, a grey rectangle containing a red diagonal line labeled  $\Sigma$ , another dot, and  $V^T$  (a blue rectangle). An arrow points to the right, followed by an approximation symbol ( $\approx$ ). To the right of the approximation symbol,  $A$  is shown as a green rectangle, followed by a maroon rectangle  $U$ , a dot, a grey rectangle containing a red diagonal line labeled  $\Sigma$  with a small orange square at the top-left corner labeled  $k$ , another dot, and a blue rectangle  $V^T$  with a small orange square at the top-left corner labeled  $k$ .

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The diagram shows the decomposition of a matrix  $A$  into  $U$ ,  $\Sigma$ , and  $V^T$ . On the left,  $A$  is shown as a green rectangle. An equals sign follows. To the right is a purple rectangle labeled  $U$ . This is followed by a dot, a gray rectangle containing a red diagonal line with orange dots, another dot, a blue rectangle labeled  $V^T$ , and a right-pointing arrow. To the right of the arrow is the symbol  $\approx$ . Further to the right is a green rectangle labeled  $A$ , followed by a dotted approximation symbol. After this is a purple rectangle labeled  $U$ , followed by a dot, a gray rectangle containing a red diagonal line with orange dots, another dot, and a blue rectangle labeled  $V^T$ .

## Application: image compression



→



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Apply SVD:  $\psi_{lr} = [\mathbf{U} \mathbf{D} \mathbf{V}]_{lr}$

$\mathbf{U}, \mathbf{V}$  are unitary.  $\mathbf{D} = \text{diag}(s_1, s_2, \dots)$ :

$s_i$ 's = singular values of  $\psi_{lr}$

= eigenvalues of  $\sqrt{\psi^\dagger \psi} = \sqrt{\rho} \implies s_i^2 = \text{eigenvalues of } \rho$

# Compressing $|\Psi\rangle$ with SVD

After SVD:

$$\begin{aligned} |\Psi\rangle &= \sum_{l,r} \sum_i \mathbf{U}_{li} \mathbf{D}_{ii} \mathbf{V}_{ir} |l\rangle |r\rangle \\ &= \sum_i \sum_{\substack{l,r}} \mathbf{U}_{li} \mathbf{D}_{ii} \mathbf{V}_{ir} |l\rangle |r\rangle \\ &= \sum_i s_i |i\rangle_L |i\rangle_R \leftarrow \text{Schmidt decomposition} \end{aligned}$$

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Can simply read off reduced density matrices:

$$\rho_L = \psi \psi^\dagger = \sum_i s_i^2 |i\rangle_L \langle i|_L \quad \rho_R = \psi^\dagger \psi = \sum_i s_i^2 |i\rangle_R \langle i|_R$$

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Normalization:

$$\text{Tr}(\psi^\dagger \psi) = \sum_i s_i^2 = 1 \implies s_i^2: \text{probability for } i^{\text{th}} \text{ Schmidt state pair}$$

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So,

$$S = S(\rho_L) = S(\rho_R) = -\sum_i^{\sim 2^{N/2}} s_i^2 \ln s_i^2 \rightarrow -\sum_i^m s_i^2 \ln s_i^2$$

Drop small  $s_i$ 's  $\implies$  reduce  $S$  and exponential compression,  $m \sim \mathcal{O}(100)$

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Entanglement entropy is 0  $\implies$  not entangled (makes sense)

But wait...

We need  $|\Psi\rangle$  to compress. But we want to find such a  $|\Psi\rangle$  for some  $H$ .

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DMRG:

- numerical variational technique for finding ground states
- works with MPS, not  $|\Psi\rangle$
- most efficient method for 1d systems

# MPS

**Every  $|\Psi\rangle$  can be written as an MPS.** From Schmidt decomposition:

$$|\Psi\rangle = \sum_i s_i |i\rangle_L |i\rangle_R = \sum_{i_m} s_i |i_m\rangle_L |i_m\rangle_R$$

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$$|i_3\rangle_L = \sum_{\sigma_1, \sigma_2, \sigma_3} (A^{\sigma_1} A^{\sigma_2} A^{\sigma_3})_{1, i_3} |\sigma_1 \sigma_2 \sigma_3\rangle$$

⋮

Schmidt states:

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$$|i_m\rangle_R = \sum_{\{\sigma\}_{m+1}^N} (B^{\sigma_{m+1}} B^{\sigma_{m+2}} \dots B^{\sigma_N})_{i_m,1} |\sigma_{m+1} \sigma_{m+2} \dots \sigma_N\rangle$$

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Full wavefunction:

$$|\Psi\rangle = \sum_{\{\sigma\}} A^{\sigma_1} A^{\sigma_2} \dots A^{\sigma_m} S B^{\sigma_{m+1}} B^{\sigma_{m+2}} \dots B^{\sigma_N} |\sigma_1 \sigma_2 \dots \sigma_N\rangle$$

Matrix dimensions:

$$\underbrace{(1 \times d), (d \times d^2), \dots, (d^{N/2} \times d^{N/2})}_{A}, \underbrace{\dots}_{S}, \underbrace{(d^2 \times d), (d \times 1)}_{B}$$

# MPS

Example:

$$|\text{GHZ}_4\rangle = \frac{|0000\rangle + |1111\rangle}{\sqrt{2}} = \frac{1}{\sqrt{2}} \sum_{\{\sigma\}} A_1^{\sigma_1} A_2^{\sigma_2} A_3^{\sigma_3} A_4^{\sigma_4} |\sigma_1 \sigma_2 \sigma_3 \sigma_4\rangle$$

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$$A_1^0 = \begin{pmatrix} 1 & 0 \end{pmatrix}, \quad A_2^0 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}, \quad A_3^0 = \begin{pmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}, \quad A_4^0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

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To generalize, can make all matrices to  $D \times D$ , so that

$$|\Psi\rangle = \sum_{\{\sigma\}} \text{Tr} [A_1^{\sigma_1} A_2^{\sigma_2} \dots A_m^{\sigma_m} A_{m+1}^{\sigma_{m+1}} A_{m+2}^{\sigma_{m+2}} \dots A_N^{\sigma_N}] |\sigma_1 \sigma_2 \dots \sigma_N\rangle$$

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where

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

Can see that

$$\text{Tr}[A^{\sigma_1} A^{\sigma_2} A^{\sigma_3} A^{\sigma_4}] = 1 \iff \sigma_1 = \sigma_2 = \sigma_3 = \sigma_4$$

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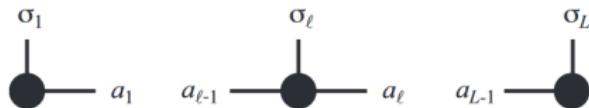
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- Decimate Hilbert space by keeping matrices in MPS at  $(D \times D)$
- Smaller Hilbert space ( $ND^2$  vs  $d^N$ ) and MPS corresponds to  $|\Psi\rangle$  with lower entanglement entropy
- MPS: natural way to describe ground states of relevant Hamiltonians

# Graphical notation

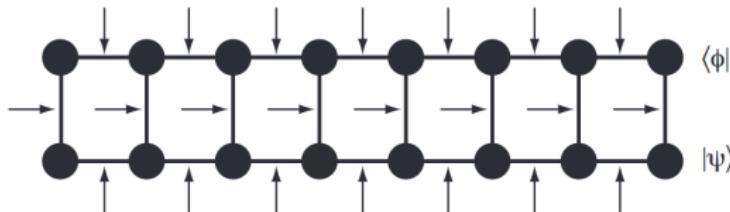
Matrices:



MPS:



Overlap for two MPS's (arrows indicate sum over indices)



# Graphical notation

MPO: Matrix product operators

If  $\langle \sigma_1 \sigma_2 \dots \sigma_N | \Psi \rangle = A^{\sigma_1} A^{\sigma_2} \dots A^{\sigma_N}$ , then

$$\langle \sigma_1 \dots \sigma_N | \hat{O} | \sigma'_1 \dots \sigma'_N \rangle = W^{\sigma_1 \sigma'_1} W^{\sigma_2 \sigma'_2} \dots W^{\sigma_{N-1} \sigma'_{N-1}} W^{\sigma_N \sigma'_N}$$

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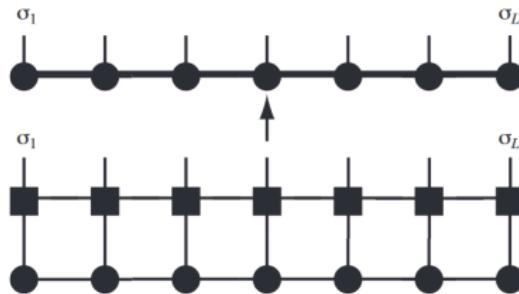
$$\hat{O} = \sum_{\{\sigma, \sigma'\}} W^{\sigma_1 \sigma'_1} W^{\sigma_2 \sigma'_2} \dots W^{\sigma_{N-1} \sigma'_{N-1}} W^{\sigma_N \sigma'_N} |\sigma\rangle \langle \sigma'|$$

⇒ can calculate  $\langle \Psi | H | \Psi \rangle$  in MPS language

# Graphical notation

MPO on MPS:

$$\hat{O} |\Psi\rangle = \sum_{\{\sigma, \sigma'\}} (W^{\sigma_1 \sigma'_1} W^{\sigma_2 \sigma'_2} \dots) (A^{\sigma_1} A^{\sigma_2} \dots) |\sigma\rangle = \sum_{\{\sigma\}} N^{\sigma_1} N^{\sigma_2} \dots |\sigma\rangle$$



# Ground state search

Variationally extremize

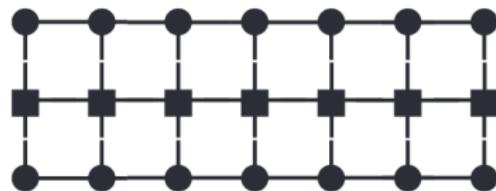
$$\langle \Psi | H | \Psi \rangle - \lambda \langle \Psi | \Psi \rangle, \text{ so that } |\Psi\rangle \rightarrow |\Psi_g\rangle, \lambda \rightarrow E_0$$

# Ground state search

Variationally extremize

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Graphically,



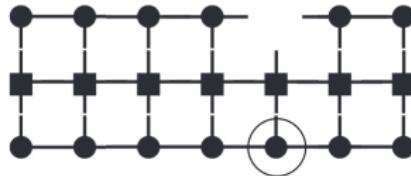
Idea: Keep matrices on all sites but one ( $\ell$ ) constant, optimize elements of  $M_{\ell}^{\sigma\ell}$ . Sweep through  $\ell$ .

# Ground state search

Minimizing



≡ solving (gradient descent)



- λ



= 0

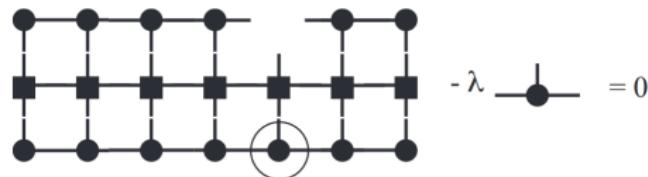
⇒ a generalized eigenvalue problem, where we treat  $M_\ell^{\sigma\ell}$  as a vector  $v$

# Ground state search

Aside: left-/right-normalized

$$\sum_{\sigma_\ell} A^{\sigma_\ell \dagger} A^{\sigma_\ell} = I \quad \sum_{\sigma_\ell} B^{\sigma_\ell \dagger} B^{\sigma_\ell} = I$$

If  $|\Psi\rangle$  is both left- and right-normalized,



⇒ standard eigenvalue problem

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- Repeat right and left sweeps until convergence

# Ground state search

Algorithm, formalized: Subscript denotes the number of updates

$$\begin{aligned} M_0 B_0 B_0 B_0 &\xrightarrow{\text{diag}} M_1 B_0 B_0 B_0 \xrightarrow{\text{SVD}} A_1 M_0 B_0 B_0 \\ &\xrightarrow{\text{diag}} A_1 M_1 B_0 B_0 \xrightarrow{\text{SVD}} A_1 A_1 M_0 B_0 \\ &\xrightarrow{\text{diag}} A_1 A_1 M_1 B_0 \xrightarrow{\text{SVD}} A_1 A_1 A_1 M_0 \\ &\xrightarrow{\text{diag}} A_1 A_1 A_1 M_1 \xrightarrow{\text{SVD}} A_1 A_1 M_1 B_1 \\ &\xrightarrow{\text{diag}} A_1 A_1 M_2 B_1 \xrightarrow{\text{SVD}} A_1 M_1 B_2 B_1 \\ &\xrightarrow{\text{diag}} A_1 M_2 B_2 B_1 \xrightarrow{\text{SVD}} M_1 B_2 B_2 B_1 \\ &\vdots \end{aligned}$$

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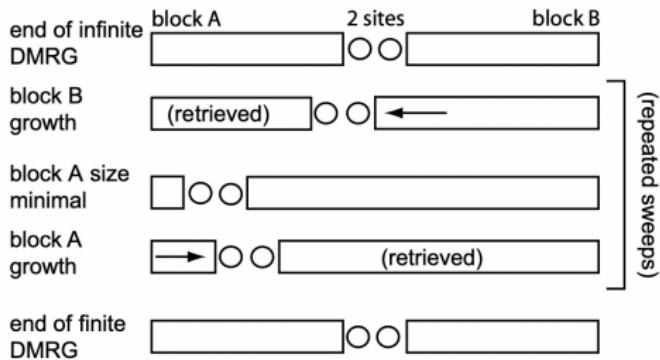
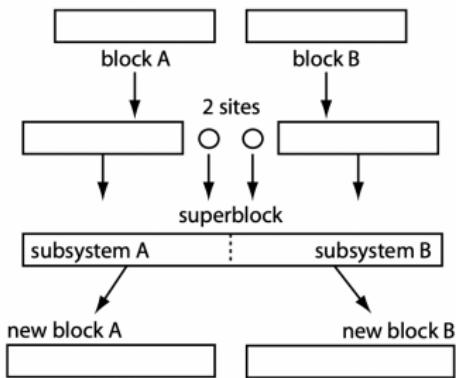
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- Repeat

# DMRG



# References

-  U. Schollwöck.  
The density-matrix renormalization group.  
*Rev. Mod. Phys.*, 77:259–315, Apr 2005.
-  Ulrich Schollwoeck.  
The density-matrix renormalization group in the age of matrix product states.  
*Annals of Physics*, 326:96–192, 2011.