

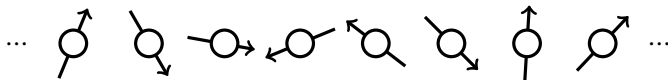
Introduction to Matrix Product States

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Quantum many-body physics

Consider an N -body system:

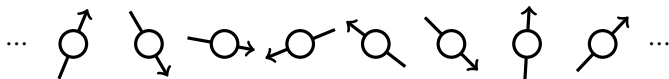


Classical: $\mathcal{O}(N)$ DOFs.

Quantum: $\mathcal{O}(\exp N)$ DOFs.

For classical computers, need to encode important DOFs.

Why matrix product states?



Represent state as a **matrix product state** (MPS)

$$|\Psi\rangle = \bigcirc - \bigcirc - \bigcirc - \dots - \bigcirc.$$

- Discard ‘long-range’ entanglement, keep local entanglement.
- Good for certain states (area law of entanglement).
- Limited for ‘general’ states (e.g. out of equilibrium).
- Works for 1D or quasi-1D systems.
- Also other varieties of **tensor networks**.

Singular value decomposition

The fundamental building block of MPS is the

Singular value decomposition (SVD)

For an $n \times m$ matrix M , we can write

$$M = USV^\dagger,$$

where U and $V^\dagger$ are unitary, and S is nonnegative real diagonal (singular value matrix).

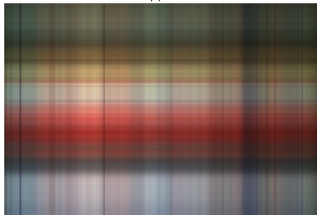
- Analogous (but not equivalent) to eigenvalue decomposition.
- Can form an *approximate* SVD by only keeping n largest singular values.

SVD image compression

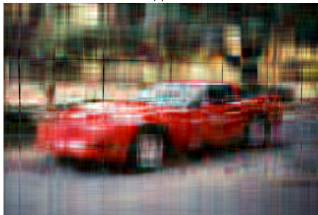
Original



Rank-1 Approximation



Rank-10 Approximation



Rank-100 Approximation



Image taken from Wikipedia: CC BY-SA 4.0 by Samir.beall.
See also: <https://timbaumann.info/svd-image-compression-demo/>

Schmidt decomposition

Given a pure state for an N -body system

$$|\Psi\rangle = \sum_{s_1 s_2 \dots s_N} c^{s_1 s_2 \dots s_N} |s_1\rangle \otimes |s_2\rangle \otimes \dots \otimes |s_N\rangle,$$

we make a bipartition between sites n and $n + 1$,
view this c as a matrix, and take an SVD

$$c^{(s_1 \dots s_n)(s_{n+1} \dots s_N)} = \sum_{ij} U^{(s_1 \dots s_n)i} S^{ij} (V^\dagger)^{j(s_{n+1} \dots s_N)}.$$

Schmidt decomposition

Can use graphical notation for tensors:

$$|\Psi\rangle = \boxed{}^c.$$

Taking the Schmidt decomposition:

$$|\Psi\rangle = \boxed{}^U \text{---} \bigcirc^S \text{---} \boxed{}^{V^\dagger}.$$

1 n n+1 N

Entanglement entropy $\mathcal{S}$ given in terms of singular values:

$$\mathcal{S} = -\text{Tr}(\rho_A \ln \rho_A) = -\sum_i S_i^2 \ln S_i^2;$$

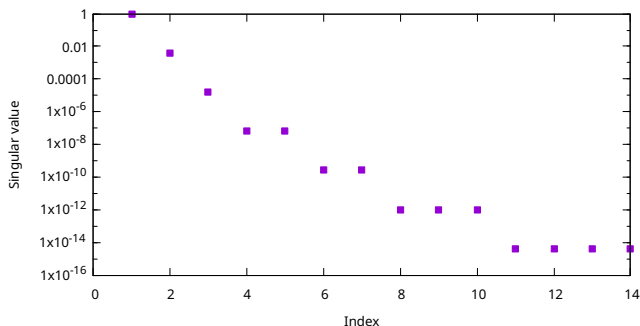
Can we use an approximate SVD here?

Singular value spectrum

Transverse field Ising chain

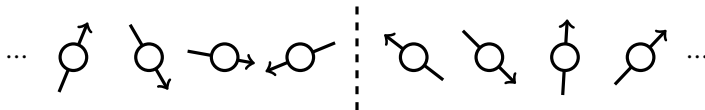
$$\hat{H} = -J \sum_j \hat{\sigma}_j^z \hat{\sigma}_{j+1}^z + h \sum_j \hat{\sigma}_j^x.$$

Ground state at $h/J = 0.8$:



Spectrum collapses at critical point $h/J = 1$.

Area law of entanglement entropy



Area law

For ground states of local Hamiltonians with an energy gap in *one* spatial dimension, the entanglement entropy of the ground state scales as the *boundary area* of the bipartition (i.e. constant).

- Only need a *constant* number of singular values as $N \rightarrow \infty$.
- Critical (gapless) states: area law plus logarithmic correction.
- Holds in some ≥ 2 D systems (less rigorous proofs).
- Entropy grows (linearly) out of equilibrium.

Matrix product states

$$|\Psi\rangle = \begin{array}{c} U \qquad S \qquad V^\dagger \\ \text{---} \text{---} \text{---} \\ | \quad \dots \quad | \quad \quad | \quad \dots \quad | \\ 1 \qquad n \qquad \qquad n+1 \quad N \end{array}$$

Take a Schmidt decomposition between every pair of sites:

$$|\Psi\rangle = \begin{array}{c} A_1^{s_1} \quad A_2^{s_2} \quad A_3^{s_3} \quad \dots \quad A_N^{s_N} \\ \circ \text{---} \circ \text{---} \circ \text{---} \dots \text{---} \circ \\ | \quad \quad | \quad \quad | \quad \quad \quad | \\ 1 \quad \quad 2 \quad \quad 3 \quad \quad \quad N \end{array} ,$$

obtain a **matrix product state** (MPS).

In general, each $A_n^{s_n}$ is a $D \times D$ matrix (D : bond dimension).

Matrix product states

$$|\Psi\rangle = \begin{array}{ccccccc} & A_1^{s_1} & A_2^{s_2} & A_3^{s_3} & \dots & A_N^{s_N} & \\ & \bigcirc & \bigcirc & \bigcirc & \dots & \bigcirc & \\ & | & | & | & & | & \\ 1 & & 2 & 3 & & & N \end{array}$$

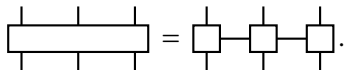
Efficient compression of state: $O(ND^2)$ degrees of freedom (compare $O(\exp N)$ for an arbitrary state).

Efficient evaluation of expectation values:

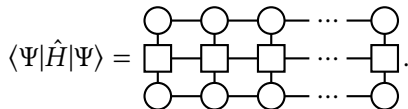
$$\langle \Psi | O_i | \Psi \rangle = \begin{array}{ccccccc} & & A_{i-1} & A_i & A_{i+1} & & \\ & \dots & \bigcirc & \bigcirc & \bigcirc & \dots & \bigcirc \\ & & | & | & | & & | \\ & \dots & \bigcirc & \bigcirc & \bigcirc & \dots & \bigcirc \\ & & & \square O & & & \\ & & & | & & & \\ & & & \bigcirc & & & \end{array} .$$

Matrix product operators

Decompose larger operators as **matrix product operators** (MPOs):



Efficiently evaluate global operators, e.g. Hamiltonian:



What can we do with MPSs?

- Cannot fully diagonalize Hamiltonian.
- Can find ground states (DMRG).
- Can perform time evolution (limited timescale: entanglement).
- Can look at some excited/thermal states.

Density matrix renormalization group (DMRG)

To find ground state: minimize energy.

Cannot update entire MPS at once: update tensors one at a time.

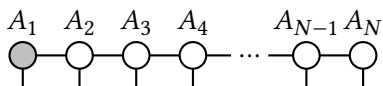
To optimize tensor A_i , want to minimize

$$\mathcal{H}_{\text{eff}}^i = \text{Diagram of MPS contraction} = \text{Diagram of effective Hamiltonian contraction}.$$

The diagram illustrates the contraction of the effective Hamiltonian $\mathcal{H}_{\text{eff}}^i$ for tensor A_i . On the left, a horizontal chain of tensors is shown. Each tensor is a square with two horizontal legs and two vertical legs. The vertical legs are connected to circles, representing indices. The chain is labeled with A_{i-1} and A_{i+1} above the corresponding tensors. The rightmost part of the chain is connected to a vertical rectangle, representing the effective Hamiltonian. The equation shows that this contraction is equivalent to a simplified diagram on the right, where the vertical rectangle is connected to a single square tensor, which is then connected to another vertical rectangle.

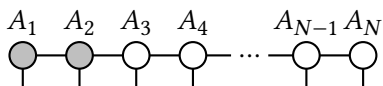
Density matrix renormalization group (DMRG)

Update tensors, sweeping from left to right:



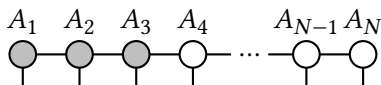
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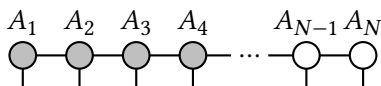
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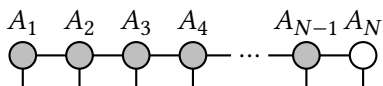
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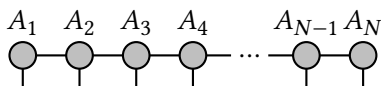
Density matrix renormalization group (DMRG)

Update tensors, sweeping from left to right:



Density matrix renormalization group (DMRG)

Update tensors, sweeping from left to right:



Sample Python DMRG code

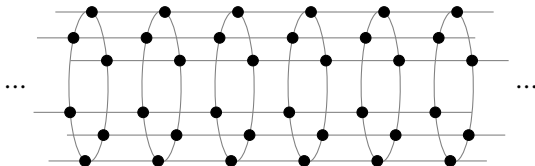
<https://mptoolkit.qusim.net/Tutorials/MPS-DMRG>

$$|\Psi(t + \Delta t)\rangle = e^{i\hat{H}\Delta t} |\Psi(t)\rangle$$

- Can replace energy optimization in DMRG with time propagation (time-dependent variational principle: TDVP).
- Can approximate the time-evolution operator (time-evolving block decimation: TEBD; MPO time evolution).

Outlook: Other directions

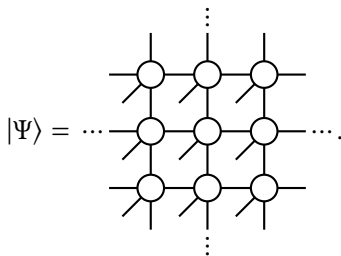
- Infinite states by explicitly enforcing translation invariance.
- Explicitly enforce global symmetries by writing tensors in quantum number blocks.
- Can look at two-dimensional systems by using cylinders:



Exponential growth in bond dimension with circumference.

Outlook: Other directions

Other types of *tensor networks* for higher spatial dimensions:



Downside: Less efficient algorithms compared with MPS.

U. Schollwöck, *The density-matrix renormalization group in the age of matrix product states*, Annals of Physics **326**, 96 (2011).

J. C. Bridgeman and C. T. Chubb, *Hand-waving and interpretive dance: an introductory course on tensor networks*, Journal of Physics A: Mathematical and Theoretical **50**, 223001 (2017).