

Subgroup Meeting #3

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DMRG for the SIS Model

- In ordinary quantum mechanics, the operators for the observables live in $\mathcal{L}(\mathcal{H}, \mathcal{H})$, where \mathcal{H} is a \mathbb{C} -**Hilbert space**.
- In the SIS model, the linear operators also live in $\mathcal{L}(\mathcal{H}, \mathcal{H})$, but here \mathcal{H} is a \mathbb{R} -**Hilbert space**.
- The linear operators in the SIS model are **not hermitian**.
- Why does the DMRG work?

Is the Hamiltonian diagonalizable?

What is the spectrum of the Hamiltonian?

How to get the NESS from the Hamiltonian?



DMRG for the SIS Model

- Diagonalizability:

Master equation: $\partial_t |P(t)\rangle = \hat{W} |P(t)\rangle$

The state: $|P(t)\rangle = e^{t\hat{W}} |P(0)\rangle$

Write \hat{W} in its Jordan form

→ \hat{W} is **diagonalizable** or the state will diverge.

- The spectrum:

\hat{W} is diagonalizable

→ The eigenvalues are **negative** or the state will diverge.

- NESS:

Entries of \hat{W} are $e^{\lambda_i t}$ goes to zero as $t \rightarrow \infty$ if $\lambda_i < 0$

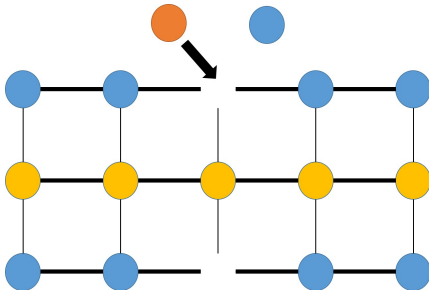
→ the only important entry is $e^{0t} = 1$

→ the NESS is the **eigenvector of \hat{W} with $\lambda = 0$**



DMRG for the SIS Model

- Variational problem: $\min_A (A^\dagger \hat{W}_{\text{eff}} A - \lambda A^\dagger \hat{N} A)$
 - Consider the gradient: $\nabla_{A^\dagger} (A^\dagger \hat{W}_{\text{eff}} A - \lambda A^\dagger \hat{N} A) = 0$
 - Generalized eigenvalue problem: $\hat{W}_{\text{eff}} A = \lambda \hat{N} A$



From NRG to DMRG

- Consider a one-dimensional quantum chain:

Hilbert space: $\mathcal{H} = \bigotimes_{i=1}^N \mathcal{H}_i$

Dimension: $\dim \mathcal{H} = \prod_{i=1}^N \dim \mathcal{H}_i$, growing rapidly as N growing.

- The Numerical Renormalization Group (NRG) is a kind of numerical algorithm that can find the **groundstate** of a system.
- The key idea of NRG:
Truncation \rightarrow **Add a new site** \rightarrow **Diagonalization** \rightarrow **Truncation**
 $\rightarrow \dots \rightarrow$ **Until Convergence**



From NRG to DMRG

- NRG works well in some impurity models but fails in **strongly correlated systems**. **Whole**= \sum **Parts**+**?**.
- The **entanglement effect**, related to many interesting phenomena, is not considered in NRG.
- In the density matrix renormalization group (DMRG), we use the **density matrix** to measure the **entanglement entropy**:



From NRG to DMRG

- Algorithms

- Infinite-size DMRG

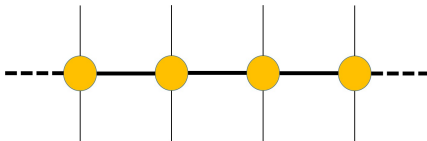
- Finite-size DMRG

<https://kikiyenhaoyang.github.io/kikiyen/Web/TN.html>



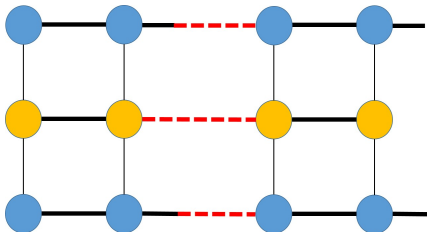
DMRG as Renormalization Group

- We can obtain the ground state by diagonalizing the infinite tensor:



This is impossible.

- We can consider the truncation to reduce the dimension of the MPO:



Variational Perspective of Finite-Size DMRG

- We want to find the ground state of a quantum state:

Ground state energy: $\inf_{|\Psi\rangle} \frac{\langle\Psi|\hat{H}|\Psi\rangle}{\langle\Psi|\Psi\rangle}$.

Optimization problem: $\min_{|\Psi\rangle} (\langle\Psi|\hat{H}|\Psi\rangle - \lambda \langle\Psi|\Psi\rangle)$.

- The quantum state $|\Psi\rangle$ can be rewritten in **MPS**.
- In principle, we can solve it by implementing the variational method concerning **all the tensors in MPS**
→ **impossible in the computers** → **implement tensor by tensor**.
- Solve the variational problem

$$\min_A (\langle\Psi|\hat{H}|\Psi\rangle - \lambda \langle\Psi|\Psi\rangle) = \min_A (A^\dagger \hat{H}_{\text{eff}} A - \lambda A^\dagger \hat{N} A)$$

A : variational parameter

\hat{H}_{eff} : effective Hamiltonian, $\langle\Psi|\hat{H}|\Psi\rangle$ without A, A^\dagger .

\hat{N} : normalization matrix, $\langle\Psi|\Psi\rangle$ without A, A^\dagger .



Variational Perspective of Finite-Size DMRG

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Conjugate gradient method

Gradient Descent

Tangent space method

