



# Large-Scale VQE Simulations with Tensor Networks

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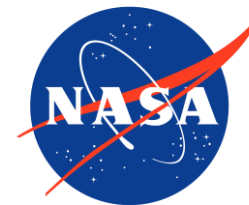
<sup>2</sup> USRA Research Institute for Advanced Computer Science

<sup>3</sup> NASA Ames Research Center

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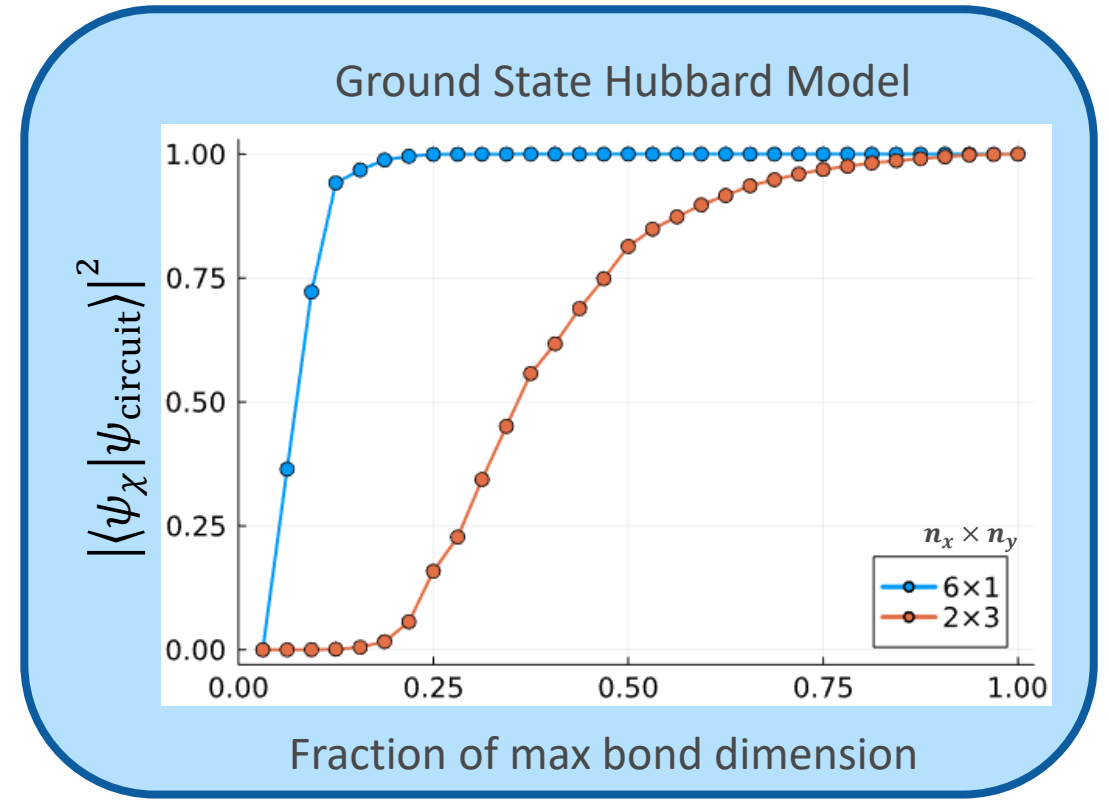
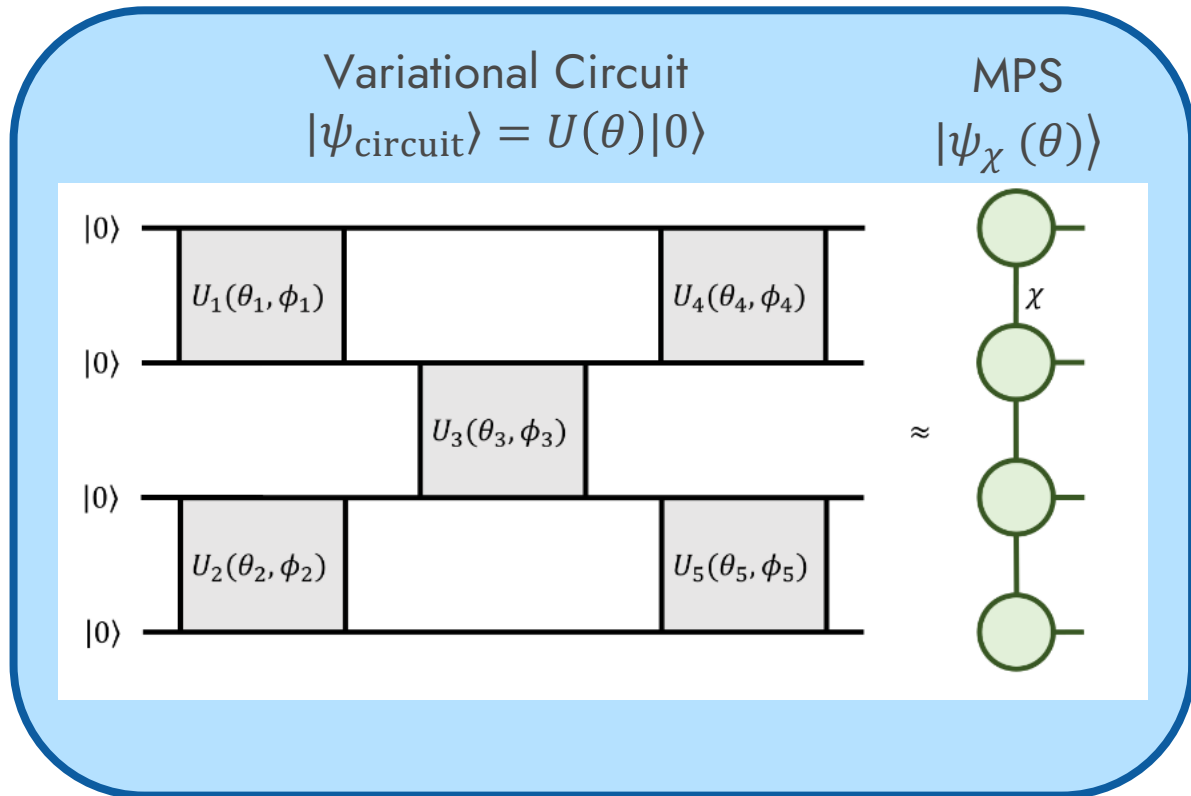
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# Tensor Networks and VQE

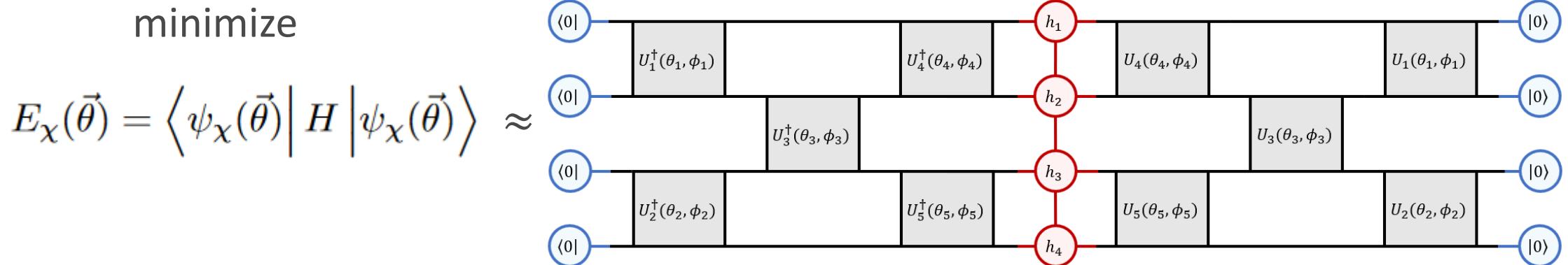
- It's hard to do large-scale circuit optimization
- Use Tensor Networks to approximate parameterized Quantum Circuits





# The Goal

- Perform VQE simulations with many qubits
  - Approximate parameterized quantum circuit as an MPS
  - Observe the performance as we tune the MPS bond dimension



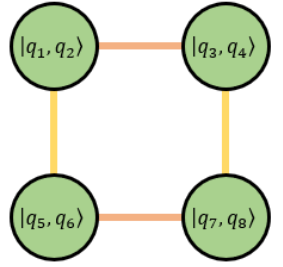


# Model and Ansatz

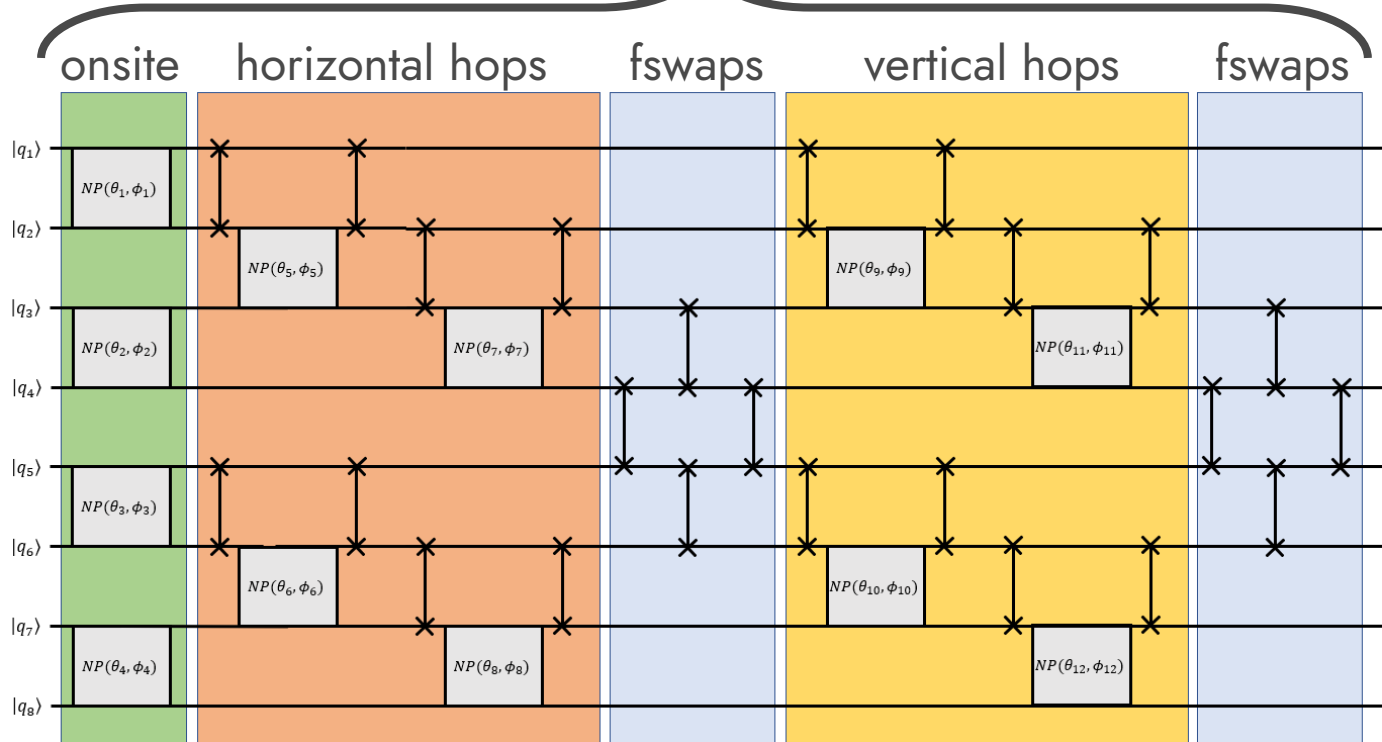
Fermi-Hubbard model with no Boundary Conditions at half filling

$$H = -t \sum_{\langle i,j \rangle, \sigma \in \{\uparrow, \downarrow\}} (a_{i\sigma}^\dagger a_{j\sigma} + a_{j\sigma}^\dagger a_{i\sigma}) + U \sum_i n_{i\uparrow} n_{i\downarrow} \quad (t, U) = (1, 2)$$

$$n_q = 2n_x n_y$$



NP ansatz layer



$$NP(\theta, \phi) = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos \theta & i \sin \theta & 0 \\ 0 & i \sin \theta & \cos \theta & 0 \\ 0 & 0 & 0 & e^{i\phi} \end{pmatrix}$$

$$\begin{matrix} \times \\ | \\ \times \end{matrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & -1 \end{pmatrix}$$

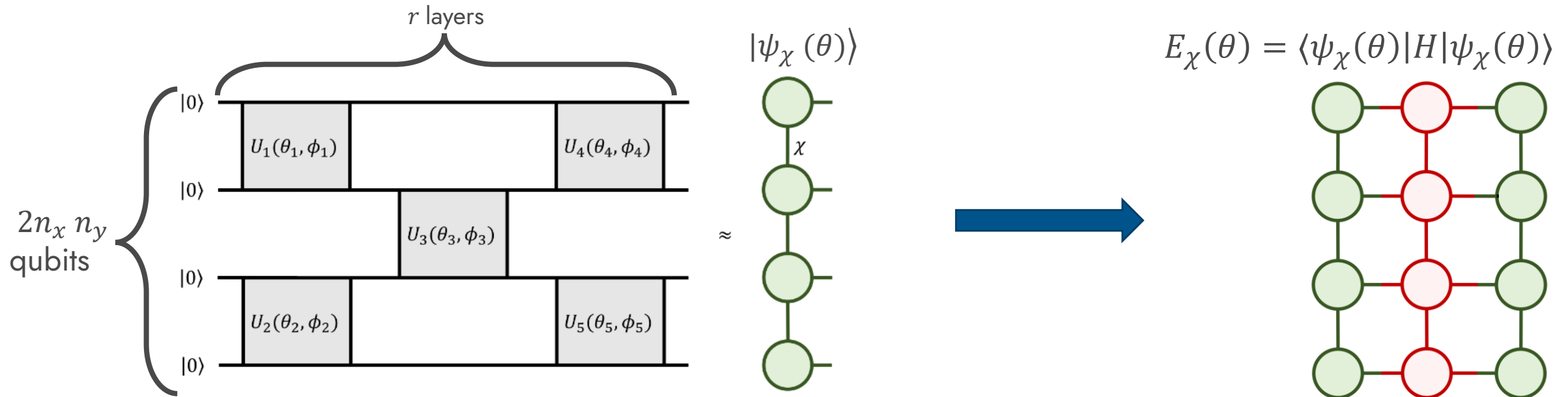
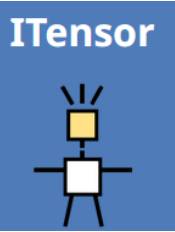
For an  $n_x \times n_y$  grid, one layer of the NP ansatz requires  $10n_x n_y - 4n_x - 4n_y$  parameters



# Optimization

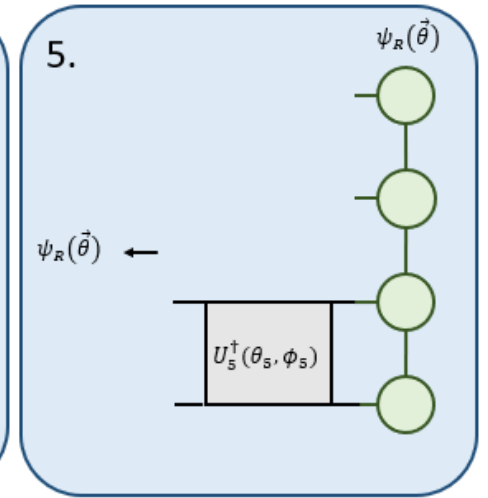
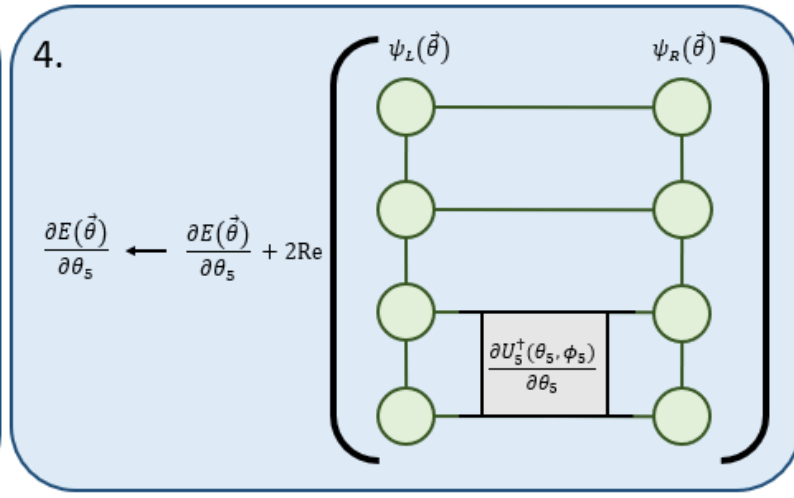
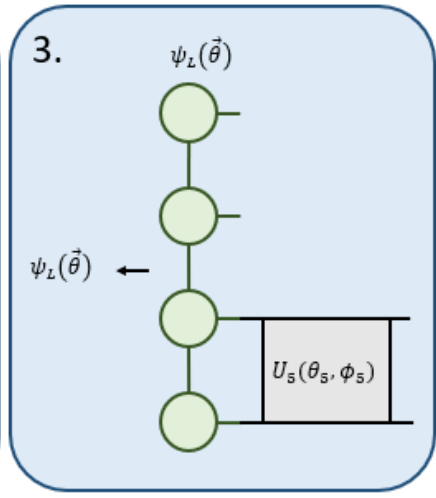
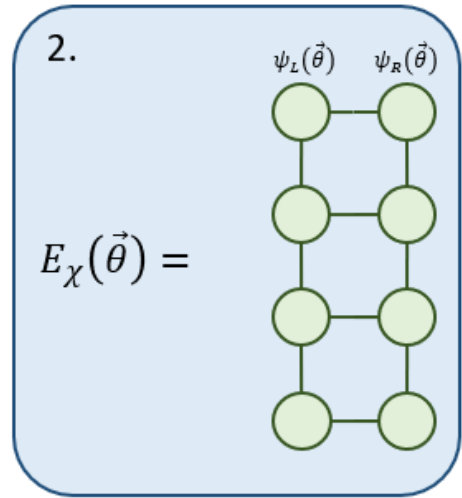
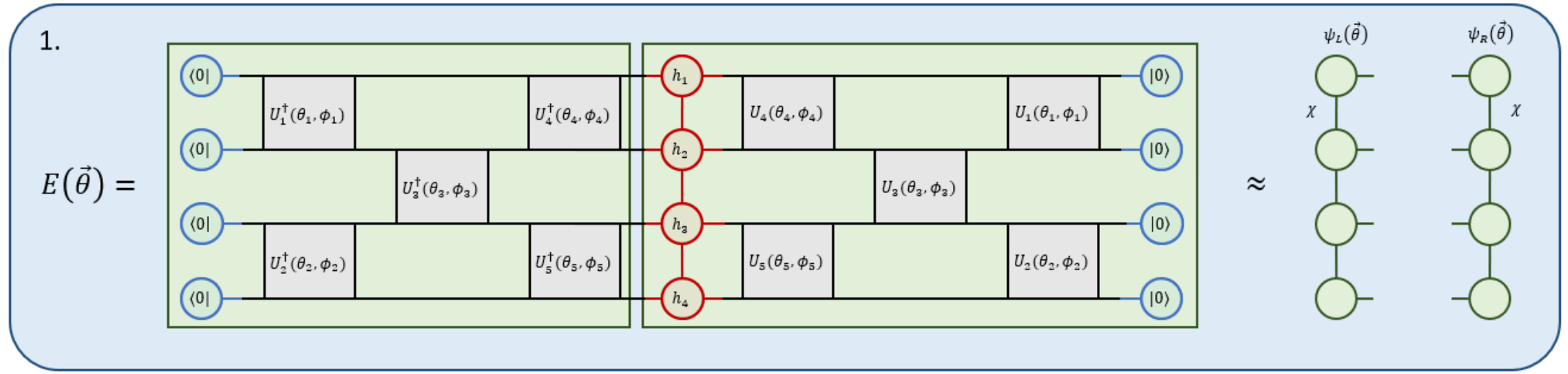
Given a system size  $n_x \times n_y$  and  $r$  layers of the NP ansatz:

1. Choose a bond dimension  $\chi$
2. Minimize  $E_\chi(\theta) = \langle \psi_\chi(\theta) | H | \psi_\chi(\theta) \rangle$  with tensor network backend
  - Gradients can be computed using automatic differentiation
  - Optimized with BFGS
  - Optimization terminates when  $E_\chi(\theta^{(t+1)}) - E_\chi(\theta^{(t)}) < 10^{-7}$
3. Obtain the optimized parameters  $\theta_\chi^*$





# Computing Gradients

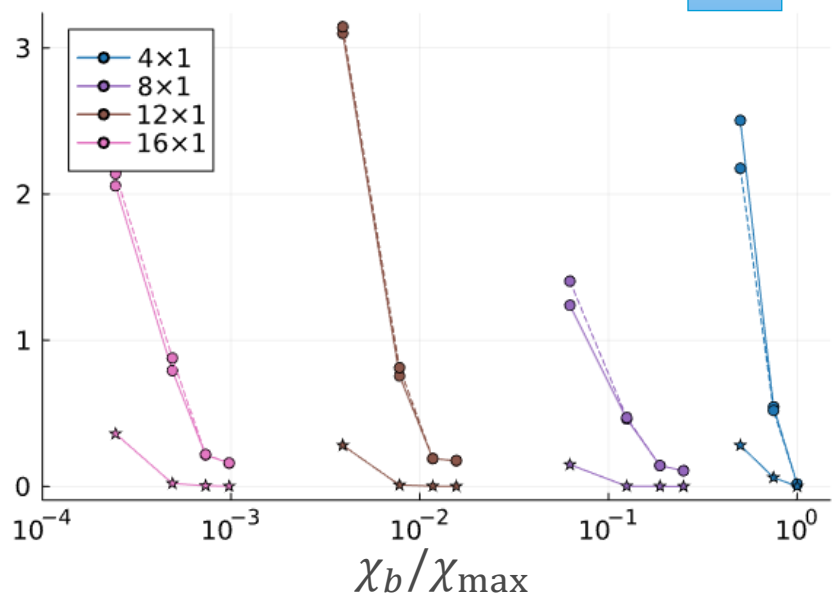




# Results

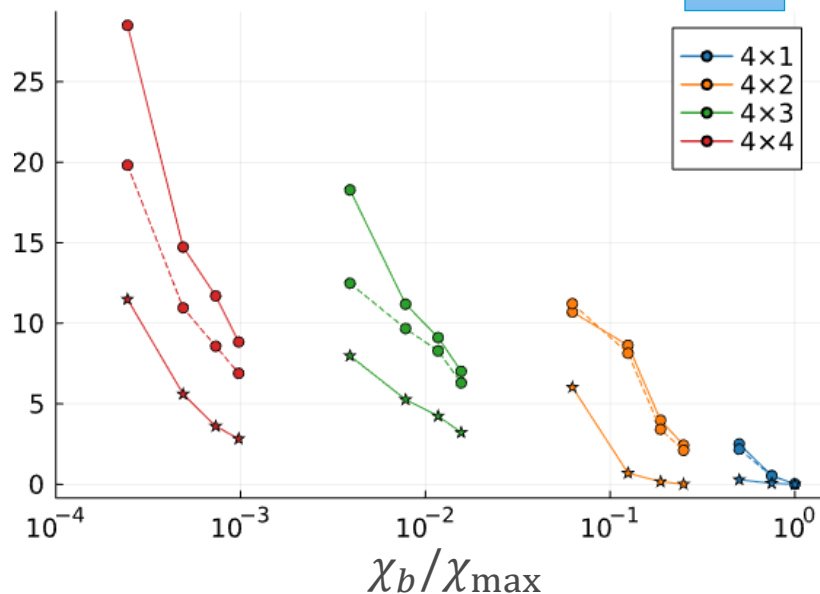
Relative Energy Error (%)

1D

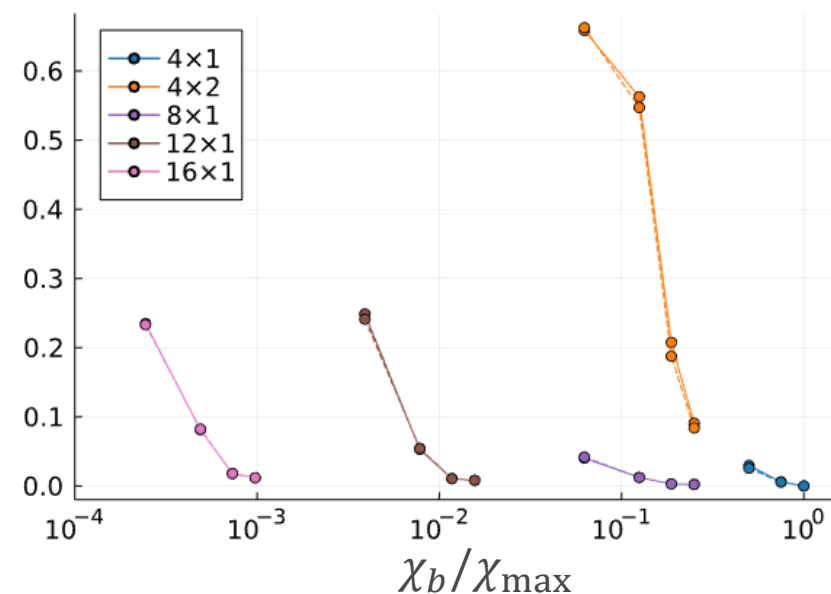


Relative Energy Error (%)

2D



$1 - |\langle \psi(\theta_{\chi_b}^*) | \psi_{\text{exact}} \rangle|^2$



We perform VQE optimization with  $\chi_b \in \{16, 32, 48, 64\}$

$$\theta_{\chi_b}^* = \operatorname{argmin}_{\theta} \langle \psi_{\chi_b}(\theta) | H | \psi_{\chi_b}(\theta) \rangle$$

$$E_{\chi_a}(\theta_{\chi_b}^*) = \langle \psi_{\chi_a}(\theta_{\chi_b}^*) | H | \psi_{\chi_a}(\theta_{\chi_b}^*) \rangle$$

● — =  $E_{\chi_b}(\theta_{\chi_b}^*)$

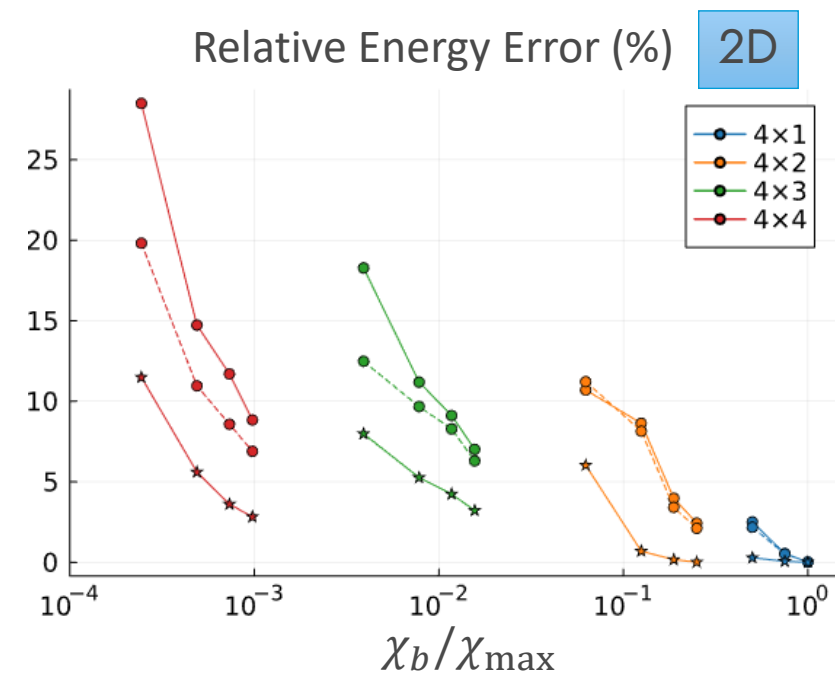
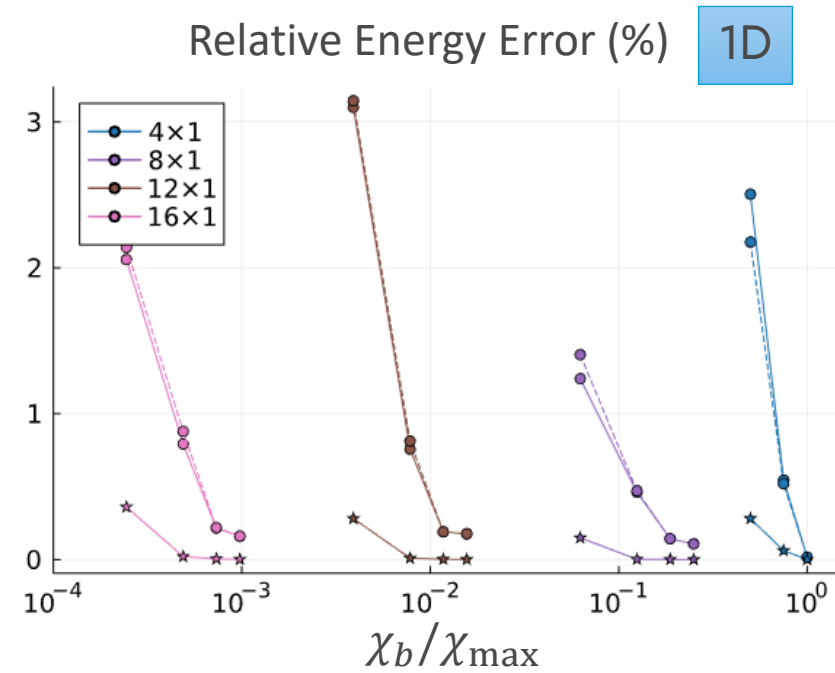
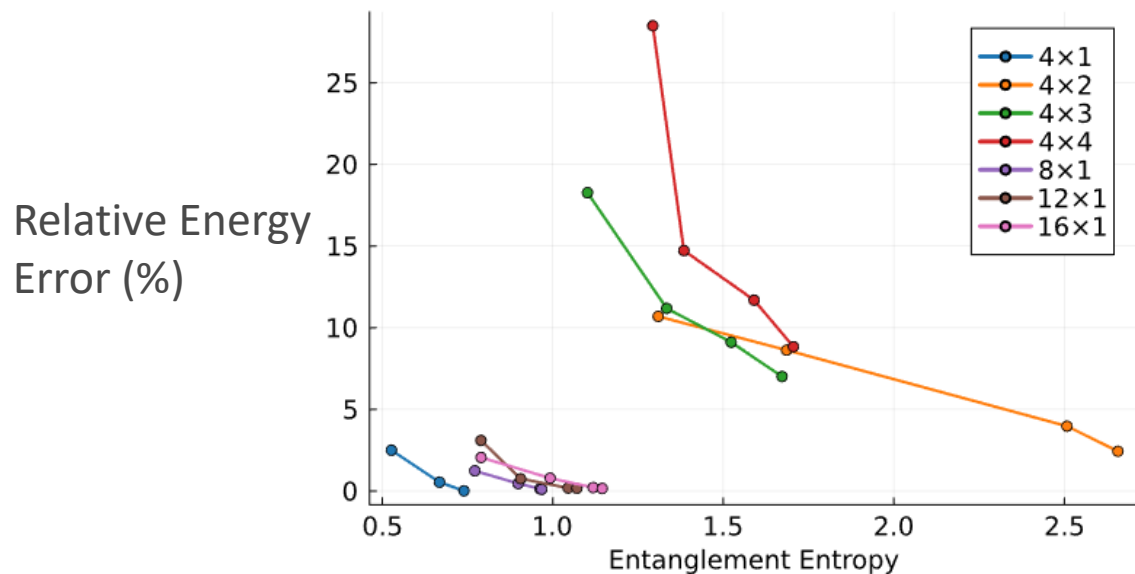
● - - - =  $E_{\min(512, \chi_{\max})}(\theta_{\chi_b}^*)$

★ — = DMRG



# Takeaways

- Using the MPS bond dimension, we can trade computational resources with ground state accuracy
- The complexity of the circuit is bounded by its entanglement entropy, which is bounded by the MPS bond dimension
- Generally, when  $\chi_1 < \chi_2$ ,  $E_{\chi_2}(\theta_{\chi_1}^*) < E_{\chi_1}(\theta_{\chi_1}^*)$ 
  - Optimizing at a low bond dimension yields circuits that yield more accurate target states when the circuit is evaluated exactly







# Further Research

1. Instead of minimizing

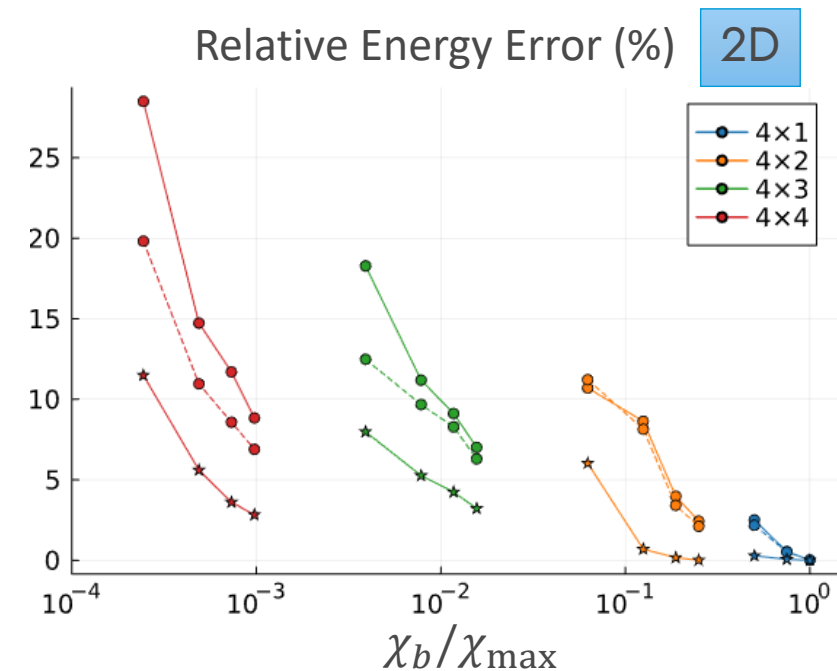
$$E_\chi(\theta) = \langle \psi_\chi(\theta) | H | \psi_\chi(\theta) \rangle,$$

what if we minimized

$$1 - F(\theta) = 1 - |\langle \psi_\chi(\theta) | \psi_{\text{DMRG}} \rangle|^2$$

2. How well do these optimized parameters perform as initialization for true VQE?

1. How many optimization steps
2. Does this get us out of barren plateaus





# Conclusion

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- Performed VQE simulations of up to 32 qubits
- MPS bond dimension tunes accuracy with resources
- Optimized parameters yield more accurate energies when circuits are evaluated exactly

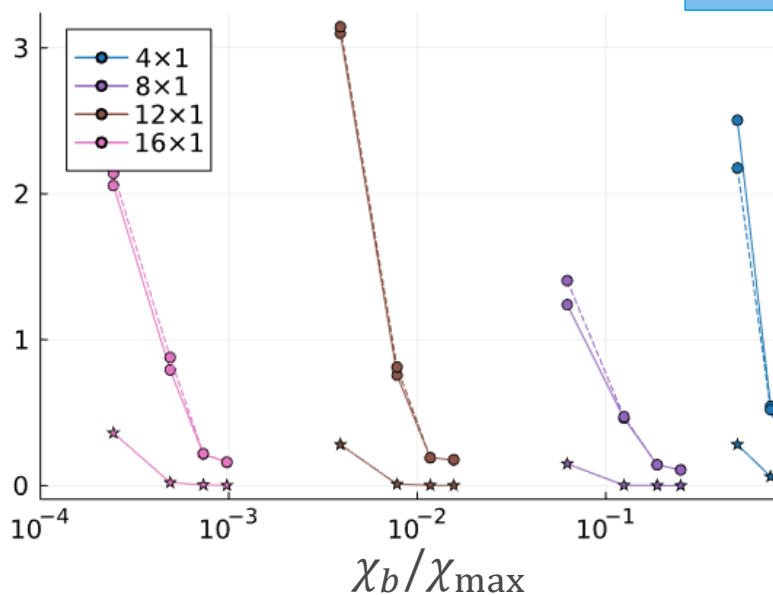


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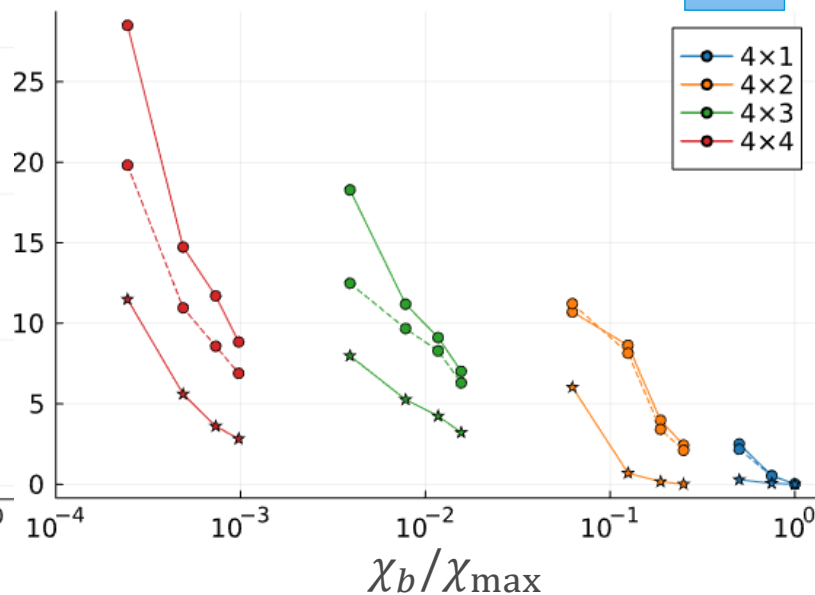
Relative Energy Error (%)

1D



Relative Energy Error (%)

2D



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