Design of Quantum-classical Computing Hybrid Algorithms for Materials Simulation Noah Berthusen^{1,2}, Anirban Mukherjee¹, Yongxin Yao^{1,3}, Peter P. Orth^{1,3}

Abstract

Accurately predicting properties of strongly correlated quantum materials is one of the grand challenges in many-body physics research. Classical methods for correlated materials simulation are computationally costly, in particular for complex materials with multi-orbitals and strong spin-orbit coupling. This complexity is intrinsic to the manyelectron problem and arises from the exponential growth of the complexity of the electronic many-body wave function. To overcome these limitations, this project aims to leverage existing noisy intermediatescale quantum (NISQ) computer technology to achieve highly accurate total energy calculations and to simulate quantum dynamics in correlated multi-orbital quantum materials. Specifically, we are determining the ground state solution quantum embedding Hamiltonians by implementing a variational quantum eigensolver (VQE). This is feasible on NISQ devices, as it takes advantage of relatively shallow quantum circuits for error mitigation.

Background

- Finds eigenvalues of a (often large) Hamiltonian matrix
- Based on the Variational principle

$$E_0 \leq \langle \Psi | H | \Psi \rangle = \sum_{i=1}^{N} \lambda_i | \langle \Psi_i | \Psi \rangle |^2$$

- ► Takes advantage of NISQ-era devices as well as classical computing technologies.
- Quantum subroutine is run inside of a classical optimization loop
- \triangleright Prepare and measure expectation value of the **ansatz**, $|\Psi(U(\boldsymbol{\theta})\rangle)$
- \triangleright Optimize θ until parameters converge and E_0 is reached
- Variational methods have also been developed to target excited states $E_i, i > 0$ or time-evolved states $e^{-iHt} |\Psi_0\rangle$ [2]
- ► Models of interest include Hubbard and Heisenberg models.



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Hamiltonian Variational Ansatz

- Different heuristic variational wavefunction ansätze have been proposed, often with physical insight such as keeping relevant excitations above a mean-field ground state (unitary coupled cluster ansatz UCC) or considering symmetry [3]
- ► The Hamiltonian Variational Ansatz (HVA) improves on unitary coupled cluster ansatz, which is a common ansatz used in quantum chemistry, by using rotations by terms in the Hamiltonian to reduce circuit depth [4]

$$|\Psi_T\rangle = \prod_{b=1}^{S} \left(U_U \left(\frac{\theta_U^b}{2}\right) U_h \left(\theta_h^b\right) U_v \left(\theta_v^b\right) U_U \left(\frac{\theta_U^b}{2}\right) \right)$$

- $\blacktriangleright U_x(\theta_x^b)$ approximates $e^{i\theta_x^b H_x}$ for $x \in \{U, h, v\}$, the interaction, horizontal, and vertical hopping terms
- Proposes two different optimization methods
- \triangleright **Global variational** optimizes all parameters for a fixed S \triangleright **Annealed variational** targets $H_{b/S}$ and optimizes the *b* ansatz step each iteration from 1, ..., S

qubit-ADAPT

- ► Ansatz still might contain "useless" exponential terms. Can we alleviate that?
- Dynamically build ansatz by selecting most important multi-qubit operators from an **operator pool**, $\{A_1, ..., A_j\}$ [5]
- ▷ A new way of constructing the operator pool by exploiting the noncommutativity between different parts of the Hamiltonian.
- $\triangleright N_Y(O_{ij})$ is the number of Y gates in O_{ij}

pool = {
$$O_{ij} = [A_i, A_j]$$
 if $N_Y(O_{ij}) = 2k$ -





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Results



- Larger systems can be solved by solving two decoupled smaller systems and "stitching" them together with another VQE cycle
- Feasibility of running on NISQ devices is still in question due to the many measurements needed as well as the number of CNOT gates in the ansatz

References

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 $||\Psi_I\rangle$

 $+1\}$

Repeat until radient vector has converged Select operator with largest gradient $|\Psi^{(n+1)}\rangle = e^{\theta_j A_j} |\Psi^{(n)}\rangle$ Grow ansatz