

Quantum Simulation of Molecular Ground States

GitHub Repository Documentation

Authors:

Dasari Ravi Teja (R210225)
Thirumalasetty Lavanya (R211003)

Institution:

Department of Electronics and Communication Engineering
Rajiv Gandhi University of Knowledge Technologies
RK Valley, Kadapa, Andhra Pradesh - 516330

Supervisor:

Dr. Priyanka Sahu, PhD
Assistant Professor, Department of ECE

Date: February 13, 2026

Table of Contents

1. Project Overview
2. Problem Statement
3. Solution Approach
4. Technology Stack
5. Installation Guide
6. Usage Examples
7. Results and Analysis
8. Advantages
9. Challenges Faced
10. Future Scope
11. References

1. Project Overview

Drug discovery and molecular analysis are complex and resource-intensive processes, often requiring extensive computational power and time to evaluate molecular properties. Classical computational methods face limitations in accurately modeling quantum systems due to the exponential growth of the Hilbert space with system size.

This project explores the application of Variational Quantum Eigensolver (VQE) algorithms on quantum computing platforms such as Qiskit to estimate the ground-state energies of molecular systems. Using parameterized quantum circuits like EfficientSU2 and qubit-mapped molecular Hamiltonians, simulations were performed to compute minimum energy configurations, which are crucial for understanding molecular structure, chemical reactivity, and potential drug interactions.

Key Features

- Implementation of VQE for molecular ground-state energy estimation
- Comparison with classical Hartree-Fock calculations
- Support for various molecular systems using SMILES/PDB formats
- Visualization of quantum circuits and energy convergence
- Modular architecture for easy experimentation

2. Problem Statement

Accurately determining the ground-state energy of molecules is one of the most fundamental and computationally expensive problems in quantum chemistry. The exact solution of the electronic Schrödinger equation becomes intractable as the number of electrons increases, due to the exponential growth of the Hilbert space.

Classical methods face several challenges:

- **Exponential scaling:** Full Configuration Interaction (FCI) and Coupled Cluster methods scale exponentially with system size
- **Correlation effects:** Hartree-Fock neglects electron correlation, limiting accuracy
- **Computational cost:** Post-HF methods (CCSD, CCSDT) are impractical for larger molecules
- **Strongly correlated systems:** Traditional approximations break down for complex electron interactions

Research Question: Can hybrid quantum-classical algorithms (specifically VQE) provide an efficient and scalable approach for estimating molecular ground-state energies using currently available quantum simulation platforms?

3. Solution Approach

Workflow

1. **Molecular Input:** Define molecule using .xyz, .pdb, or SMILES format
2. **Hamiltonian Construction:** Apply second quantization and qubit mapping (JW/BK/Parity)
3. **Ansatz Selection:** Choose parameterized circuit (EfficientSU2/UCCSD)
4. **VQE Optimization:** Quantum state preparation and classical parameter optimization
5. **Energy Output:** Compare results with HF/FCI and analyze convergence

Mathematical Foundation

The electronic Hamiltonian in second quantization is expressed as a sum of one-electron and two-electron integrals with fermionic creation and annihilation operators. This Hamiltonian is then mapped to qubit operators using transformations such as Jordan-Wigner or Bravyi-Kitaev mappings.

4. Technology Stack

Component	Technology	Purpose
Quantum Framework	Qiskit	Quantum circuit construction and simulation
Quantum Chemistry	PySCF	Molecular integrals and Hartree-Fock calculations
Optimization	SciPy	Classical optimization (SLSQP, COBYLA)
Simulation Backend	Qiskit Aer	High-performance quantum circuit simulation
Programming	Python 3.8+	Implementation language

5. Installation Guide

Prerequisites

- Python 3.8 or higher
- pip package manager
- Virtual environment (recommended)

Installation Steps

```
# Clone the repository git clone
https://github.com/yourusername/quantum-molecular-simulation.git cd
quantum-molecular-simulation # Create virtual environment python -m venv venv source
venv/bin/activate # On Windows: venv\Scripts\activate # Install dependencies pip
install -r requirements.txt
```

Required Packages

- qiskit >= 1.0.0
- qiskit-aer >= 0.13.0
- qiskit-nature >= 0.7.0
- pyscf >= 2.3.0
- scipy >= 1.10.0
- numpy >= 1.24.0
- matplotlib >= 3.7.0

6. Usage Examples

Basic H₂ Molecule Simulation

```
from qiskit_nature.units import DistanceUnit from qiskit_nature.second_q.drivers import
PySCFDriver from qiskit_nature.second_q.mappers import JordanWignerMapper from
qiskit.algorithms.minimum_eigensolvers import VQE from qiskit.algorithms.optimizers
import SLSQP from qiskit.primitives import Estimator from qiskit.circuit.library import
EfficientSU2 # Define molecule driver = PySCFDriver( atom="H 0 0 0; H 0 0 0.735",
basis="sto3g", charge=0, spin=0, unit=DistanceUnit.ANGSTROM ) # Get molecular problem
problem = driver.run() # Map to qubits mapper = JordanWignerMapper() qubit_op =
mapper.map(problem.second_q_ops()[0]) # Setup VQE ansatz =
EfficientSU2(num_qubits=qubit_op.num_qubits) optimizer = SLSQP(maxiter=100) estimator =
Estimator() vqe = VQE(estimator, ansatz, optimizer) # Run calculation result =
vqe.compute_minimum_eigenvalue(qubit_op) print(f"VQE Energy: {result.eigenvalue}")
print(f"HF Energy: {problem.reference_energy}")
```


7. Results and Analysis

Example: H₂ Molecule Simulation

Method	Energy (Hartree)	Computation Time
Hartree-Fock	-1.1168	< 1s
VQE (EfficientSU2)	-1.0614	~30s
Correlation Energy	-1.0614	-

Note: The energy difference captures electron correlation beyond mean-field approximation. VQE provides a more accurate representation of the molecular ground state compared to Hartree-Fock.

8. Advantages

Computational Benefits

- **Scalability:** Better scaling than classical FCI/CCSD(T) methods
- **Correlation:** Naturally captures electron correlation effects
- **Flexibility:** Modular architecture allows easy customization
- **Future-ready:** Compatible with emerging quantum hardware

Real-World Impact

- **Faster drug discovery** - Reducing time to market for new medicines
- **Better batteries** - Longer life and faster charging for electric vehicles
- **Cleaner manufacturing** - Efficient catalysts and sustainable processes
- **Improved agriculture** - Better fertilizers for food production
- **Advanced materials** - Stronger, lighter materials for various applications
- **Energy efficiency** - More efficient solar cells and energy storage
- **Environmental benefits** - Reduced pollution through better chemical processes

9. Challenges Faced

Implementation Challenges

- **NISQ limitations:** Current quantum hardware is noisy and limited in qubit count
- **Software instability:** Rapid evolution of quantum frameworks causes compatibility issues
- **Learning curve:** Requires deep understanding of quantum mechanics, chemistry, and programming
- **Resource intensive:** Large molecules require significant computational resources
- **Documentation gaps:** Rapidly changing APIs make many examples outdated

Lessons Learned

Throughout this project, we encountered significant challenges due to the highly technical nature of quantum computing. Every concept required extensive research, browsing multiple websites and research papers. Many open-source repositories were outdated or incompatible with current frameworks. Even AI tools frequently produced errors due to the rapid evolution of the quantum ecosystem. This experience highlighted the importance of building strong fundamentals from scratch rather than relying solely on existing examples.

10. Future Scope

Market Opportunities

The global quantum computing market is growing exponentially with multi-billion dollar projections. Quantum chemistry and materials discovery represent key high-impact domains. Industries including pharmaceuticals, energy, aerospace, materials science, cybersecurity, and finance are actively exploring quantum algorithms.

Future Developments

- **AI Integration:** Hybrid quantum-AI drug design pipelines
- **Larger Biomolecules:** Advanced ansätze for protein modeling
- **Personalized Medicine:** Tailored drug discovery for genetic structures
- **Quantum Data Marketplaces:** Secure institutional data sharing
- **Open Collaboration:** University-industry quantum API development
- **Hardware advances:** Fault-tolerant quantum computers enabling previously impossible simulations

11. Conclusion

This project demonstrates how quantum computing—specifically the Variational Quantum Eigensolver (VQE)—can be used to estimate the ground-state energy of molecular systems with significantly improved scalability compared to traditional classical methods. While techniques like Hartree-Fock, MP2, CI, and CCSD(T) remain foundational in quantum chemistry, their computational cost increases exponentially with molecular size, making them impractical for large or highly correlated systems.

By contrast, hybrid quantum-classical algorithms allow us to approximate molecular energies using shallow quantum circuits that can be simulated on present-day hardware. Throughout this work, we successfully constructed the molecular Hamiltonian, implemented VQE using publicly available quantum platforms such as Qiskit, analyzed energy outputs, and compared the results against classical Hartree-Fock baselines.

As quantum hardware moves toward fault tolerance and algorithmic innovations accelerate, hybrid HPC-AI-Quantum workflows will become central to next-generation scientific discovery. This project serves as an early step toward that future—showcasing how emerging quantum technologies can reshape research, industry, and human progress.

12. References

1. Qiskit Nature Documentation - <https://qiskit-community.github.io/qiskit-nature/>
2. IBM Quantum Learning - <https://learning.quantum.ibm.com/>
3. PySCF Documentation - <https://pyscf.org/>
4. SciPy Optimization Reference - <https://docs.scipy.org/doc/scipy/reference/optimize.html>
5. Peruzzo, A., et al. (2014). 'A variational eigenvalue solver on a photonic quantum processor.' Nature Communications, 5, 4213.
6. Shor, P. W. (1994). 'Algorithms for quantum computation: discrete logarithms and factoring.' Proceedings 35th Annual Symposium on Foundations of Computer Science.
7. Schrödinger, E. (1926). 'An Undulatory Theory of the Mechanics of Atoms and Molecules.' Physical Review, 28(6), 1049-1070.
8. Hartree, D. R. (1928). 'The Wave Mechanics of an Atom with a Non-Coulomb Central Field.' Mathematical Proceedings of the Cambridge Philosophical Society, 24(1), 89-110.
9. Fock, V. (1930). 'Näherungsmethode zur Lösung des quantenmechanischen Mehrkörperproblems.' Zeitschrift für Physik, 61, 126-148.

Appendix A: Repository Structure

```
quantum-molecular-simulation/ ■■■■ src/ ■ ■■■■ hamiltonian/ # Hamiltonian construction
utilities ■ ■■■■ vqe/ # VQE implementation ■ ■■■■ ansatz/ # Quantum circuit ansätze ■
■■■■ utils/ # Helper functions ■■■■ examples/ ■ ■■■■ h2_molecule.ipynb # H2 molecule
example ■ ■■■■ lih_molecule.ipynb # LiH molecule example ■ ■■■■ water.ipynb # H2O
molecule example ■■■■ tests/ # Unit tests ■■■■ docs/ # Documentation ■■■■ results/ #
Simulation results ■■■■ requirements.txt # Python dependencies ■■■■ README.md # Main
documentation ■■■■ LICENSE # MIT License
```

Contact Information

Authors:

Dasari Ravi Teja (R210225)
Thirumalasetty Lavanya (R211003)

Department:

Electronics and Communication Engineering
Rajiv Gandhi University of Knowledge Technologies
RK Valley, Kadapa, Andhra Pradesh - 516330

Project Supervisor:

Dr. Priyanka Sahu, PhD
Assistant Professor, Department of ECE