

Variational Quantum Eigensolver

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Quantum Computing in the NISQ era and beyond

John Preskill

Quantum **2**, 79 (2018)

arXiv: 1801.00862

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Noisy Intermediate-Scale Quantum (NISQ) technology will be available in the near future. Quantum computers with 50-100 qubits may be able to perform tasks which surpass the capabilities of today's classical digital computers, but noise in quantum gates will limit the size of quantum circuits that can be executed reliably. NISQ devices will be useful tools for exploring many-body quantum physics, and may have other useful applications, but the 100-qubit quantum computer will not change the world right away — we should regard it as a significant step toward the more powerful quantum technologies of the future. Quantum technologists should continue to strive for more accurate quantum gates and, eventually, fully fault-tolerant quantum computing.

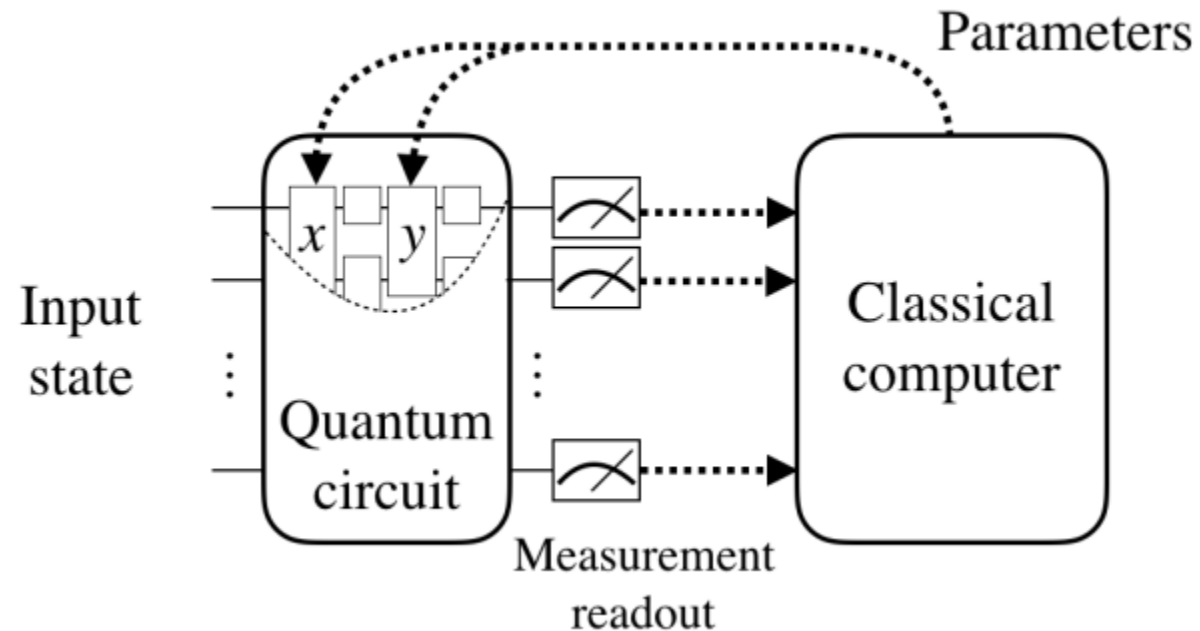
Hybrid Quantum-Classical Algorithm

- NISQ (**N**oisy **I**ntermediate **S**cale **Q**uantum) Devices:
 - Imperfect gates, noise measurements and decoherence of the qubits
(No error correction)
- NISQ algorithm solution: **small number of qubits and low circuit depth**
- Hybrid Quantum-Classical (HQC) Algorithm: **leverage strengths of quantum and classical computation**
- Variational quantum circuit algorithm is the most popular Hybrid Quantum-Classical algorithm.

Hybrid Quantum-Classical Algorithm

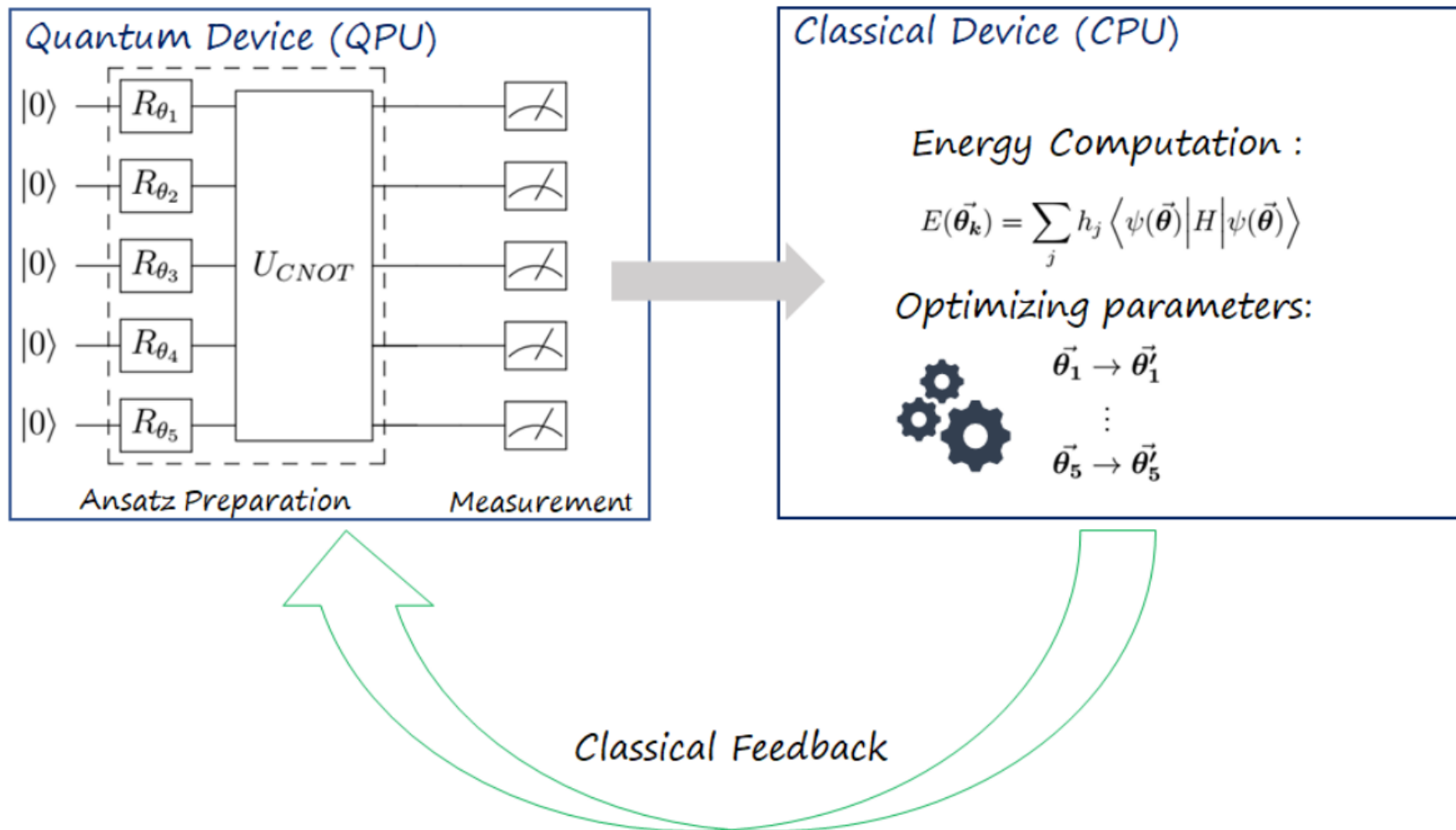
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- Hybrid Quantum-Classical (HQC) Algorithm: **leverage strengths of quantum and classical computation**
- Variational quantum eigensolver (VQE) is the most famous Hybrid Quantum-Classical algorithm.

Scheme of HQC Algorithm



- **QPU:** Input and output states
- **CPU:** Classical Optimization

Variational Quantum Eigensolver (VQE)



Variational principle

$$\mathcal{H}|\psi_\alpha\rangle = \varepsilon_\alpha|\psi_\alpha\rangle, \quad \alpha = 0, 1, \dots$$

where

$$\varepsilon_0 \leq \varepsilon_1 \leq \varepsilon_2 \leq \dots \leq \varepsilon_\alpha \leq \dots, \quad \langle\psi_\alpha|\psi_\beta\rangle = \delta_{\alpha\beta}$$

Theorem - the variational principle

Given *any normalized function* $\tilde{\psi}$ (that satisfies the appropriate boundary conditions), then the expectation value of the Hamiltonian represents an upper bound to the *exact* ground state energy

$$\langle\tilde{\psi}|\mathcal{H}|\tilde{\psi}\rangle \geq \varepsilon_0.$$

What if $\tilde{\psi}$ is a ground state w.f.?

$$\langle\tilde{\psi}|\mathcal{H}|\tilde{\psi}\rangle = \varepsilon_0$$

Proof

$\tilde{\psi}$ are normalized $\Rightarrow \langle \tilde{\psi} | \tilde{\psi} \rangle = 1$

On the other hand, (unknown) ψ_α form a complete set $\Rightarrow |\tilde{\psi}\rangle = \sum_\alpha c_\alpha |\psi_\alpha\rangle$

So,

$$\langle \tilde{\psi} | \tilde{\psi} \rangle = \left\langle \sum_\beta c_\beta \psi_\beta \left| \sum_\alpha c_\alpha \psi_\alpha \right. \right\rangle = \sum_{\alpha \beta} c_\beta^* c_\alpha \underbrace{\langle \psi_\beta | \psi_\alpha \rangle}_{\delta_{\alpha\beta}} = \sum_\alpha |c_\alpha|^2 = 1$$

Now

$$\langle \tilde{\psi} | \mathcal{H} | \tilde{\psi} \rangle = \left\langle \sum_\beta c_\beta \psi_\beta \left| \underbrace{\mathcal{H} \left| \sum_\alpha c_\alpha \psi_\alpha \right.}_{\sum_\alpha c_\alpha \underbrace{\mathcal{H} |\psi_\alpha\rangle}_{\mathcal{E}_\alpha |\psi_\alpha\rangle}} \right. \right\rangle = \sum_{\alpha \beta} c_\beta^* c_\alpha \mathcal{E}_\alpha \underbrace{\langle \psi_\beta | \psi_\alpha \rangle}_{\delta_{\alpha\beta}} = \sum_\alpha \mathcal{E}_\alpha |c_\alpha|^2$$

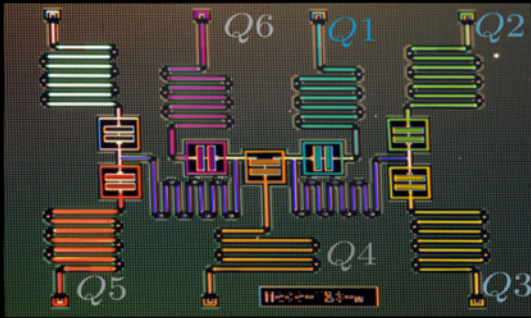
But $\mathcal{E}_\alpha \geq \mathcal{E}_0$, $\forall \alpha$, hence

$$\langle \tilde{\psi} | \mathcal{H} | \tilde{\psi} \rangle \geq \sum_\alpha \mathcal{E}_0 |c_\alpha|^2 = \mathcal{E}_0 \sum_\alpha |c_\alpha|^2 = \mathcal{E}_0$$

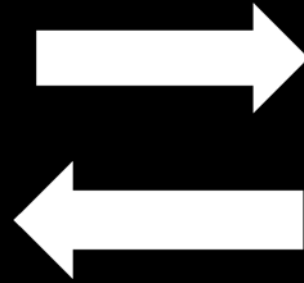
Hybrid Algorithms for NISQ Devices - VQE

IBM Q

A simple hybrid quantum-classical algorithm can be used to solve problems where the goal is to **minimize** the **energy** of a system



Prepare a trial state $|\psi(\theta)\rangle$
and compute its energy $E(\theta)$



Use classical optimizer to choose
a new value of θ to try

$$\langle \psi(\theta) | H | \psi(\theta) \rangle \geq \langle \psi_0 | H | \psi_0 \rangle$$

Variational Principle

Variational Quantum Eigensolver (VQE)

IBM Q

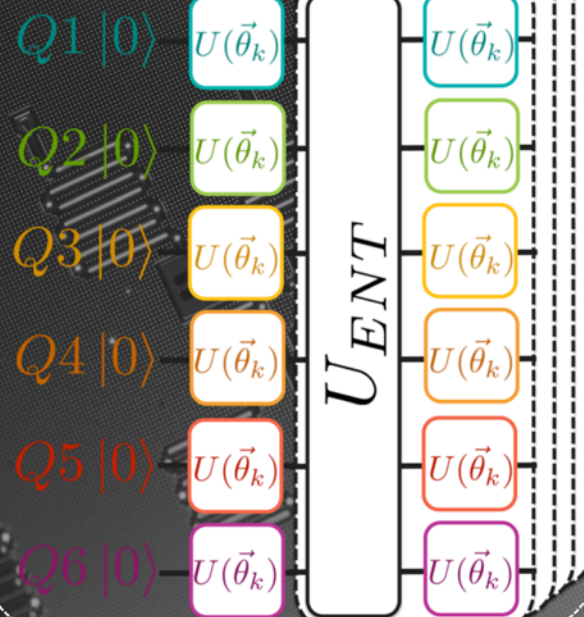
Map fermionic
Hamiltonian to
qubits

$$\mathbf{H} = \\ -(0.206)IIIIII \\ -(0.002)XZXXZZ \\ +(0.028)XXXIIX \\ \dots$$

Generate controls
($\theta, \phi, \lambda, \dots$)



Trial state preparation



Pauli
Sampling



Estimate trial
state energy

$$E(\theta, \phi, \lambda, \dots) = \\ -0.206 \langle IIIIII \rangle \\ -0.002 \langle XZXXZZ \rangle \\ +0.028 \langle XXXIIX \rangle \\ \dots$$

Optimize: Simultaneous perturbation
stochastic approximation

Focus areas:

- Reduce number of qubits (qubit tapering)
- Trial states (the ansatz)
- Fast, robust classical optimizers

Major steps in VQE

- **Variational Principle:**

$$\text{Minimize } \langle \Psi | H | \Psi \rangle$$

- **Transformation of \mathcal{H} into:**

$$\mathcal{H} = h_{\alpha}^i \sigma_{\alpha}^i + h_{\alpha\beta}^{ij} \sigma_{\alpha}^i \sigma_{\beta}^j + h_{\alpha\beta\gamma}^{ijk} \sigma_{\alpha}^i \sigma_{\beta}^j \sigma_{\gamma}^k + \dots$$

- **Linearity:**

$$\langle \psi | \mathcal{H} | \psi \rangle \equiv \langle \mathcal{H} \rangle = \mathcal{H} = h_{\alpha}^i \langle \sigma_{\alpha}^i \rangle + h_{\alpha\beta}^{ij} \langle \sigma_{\alpha}^i \sigma_{\beta}^j \rangle + h_{\alpha\beta\gamma}^{ijk} \langle \sigma_{\alpha}^i \sigma_{\beta}^j \sigma_{\gamma}^k \rangle + \dots$$

Easy for a Quantum Computer:

$$\langle \sigma_{\alpha}^i \sigma_{\beta}^j \sigma_{\gamma}^k \dots \rangle$$

Easy for a Classical Computer:

$$+, \times \rightarrow \langle H \rangle$$

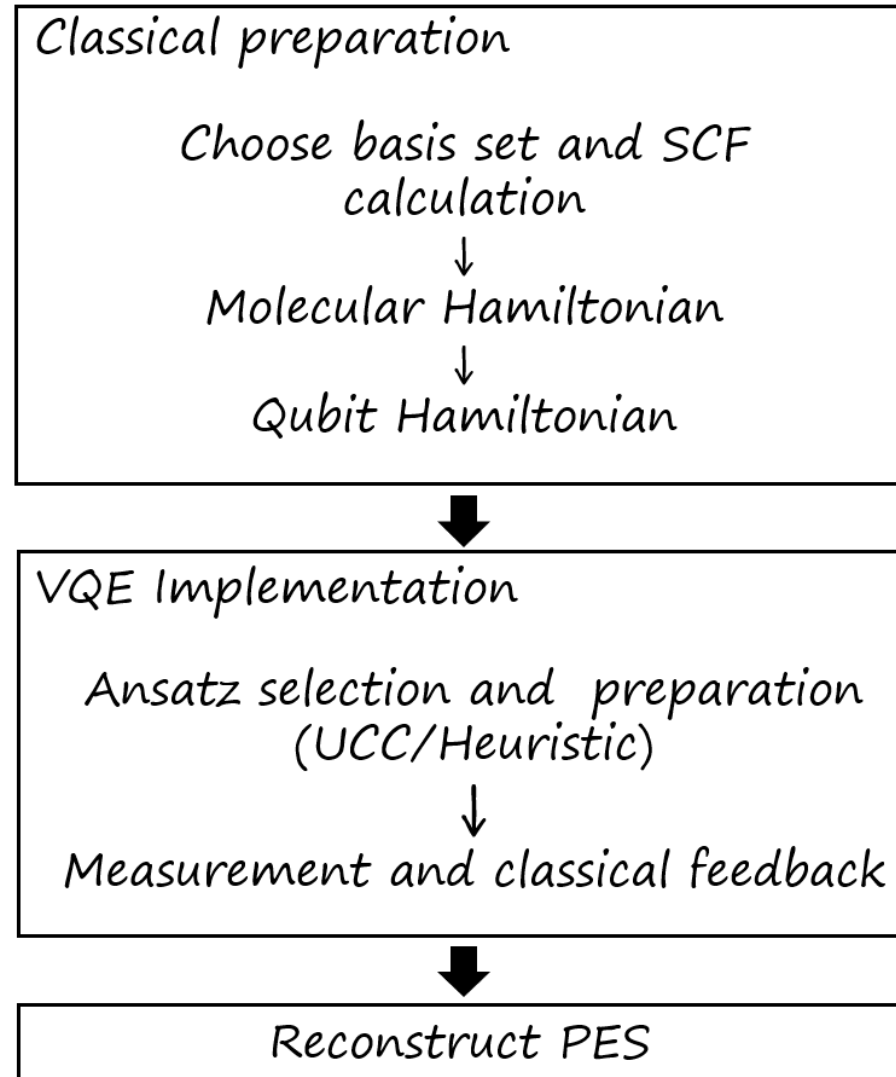
Outline of the VQE Algorithm

- **Goal:** Find the lowest eigenvalue of a given Hamiltonian

$$E(\theta) \equiv \langle H \rangle = \langle \psi(\theta) | H | \psi(\theta) \rangle \geq E_0 = \langle \psi_0 | H | \psi_0 \rangle.$$

- **Method:**
 - **Quantum:** Preparation of trial wavefunction ansatz with variational parameters θ , measurements of the expectation values
 - **Classical:** Sum the individual expectation value result and optimize the parameters
 - Iterate this procedure until it converges
- **Applications:** Quantum Chemistry, Optimization, Quantum Machine Learning...

Variation-Based Quantum Chemistry



Hamiltonian of the quantum chemistry problem

- Molecule Hamiltonian consist M nuclei and N electron

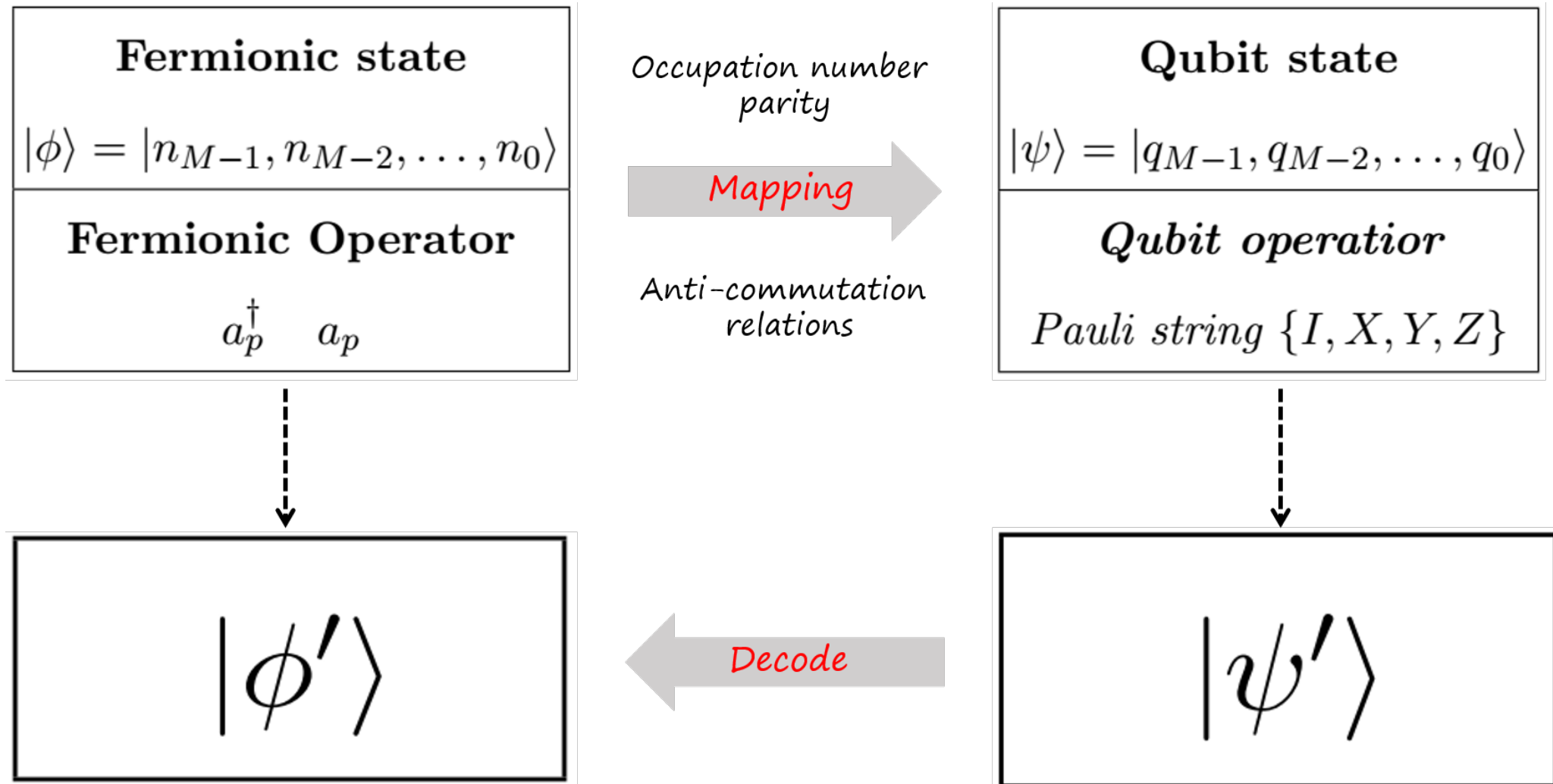
$$H = -\sum_A \frac{\nabla_A^2}{2M_A} - \sum_i \frac{\nabla_i^2}{2} - \sum_{i,A} \frac{Z_A}{|r_i - R_A|} + \frac{1}{2} \sum_{i \neq j} \frac{1}{|r_i - r_j|} + \frac{1}{2} \sum_{A \neq B} \frac{Z_A Z_B}{|R_A - R_B|}.$$

- Electronic Hamiltonian in second-quantization representation

$$H_{\text{elec}} = \sum_{pq} h_{pq} a_p^\dagger a_q + \frac{1}{2} \sum_{pqrs} h_{pqrs} a_p^\dagger a_q^\dagger a_r a_s.$$

$$h_{pq} = \int d\sigma \phi_p^*(\sigma) \left(\frac{\nabla_r^2}{2} - \sum_i \frac{Z_i}{|R_i - r|} \right) \phi_q(\sigma),$$
$$h_{pqrs} = \int d\sigma_1 d\sigma_2 \frac{\phi_p^*(\sigma_1) \phi_q^*(\sigma_2) \phi_s(\sigma_1) \phi_r(\sigma_2)}{|r_1 - r_2|}.$$

Mapping the problem into Pauli's matrices



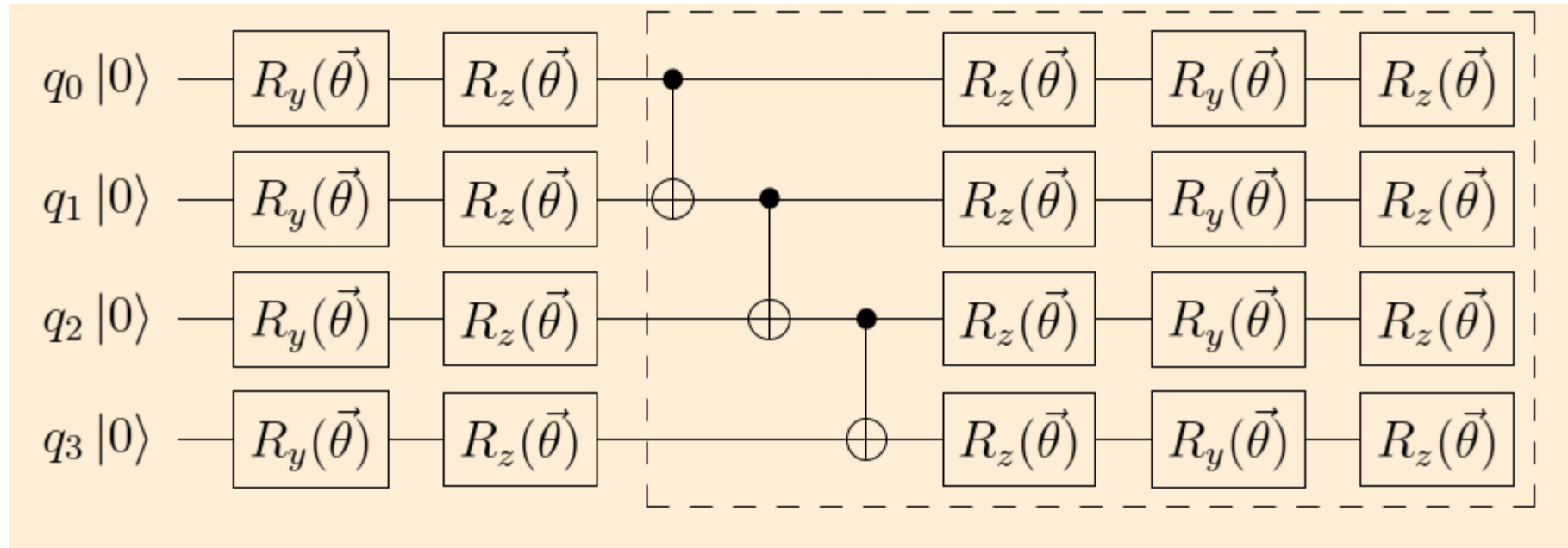
Hamiltonian in Pauli's matrix representation

- Qubit Hamiltonian: String of Pauli matrices

```
Representation: paulis, qubits: 4, size: 15
IIII      (-0.8121706073608677+0j)
IIIZ      (0.17141282639402383+0j)
IIZI      (-0.22343153674664057+0j)
IZII      (0.1714128263940239+0j)
ZIII      (-0.22343153674664057+0j)
IIZZ      (0.12062523481381844+0j)
IZIZ      (0.1686889816869329+0j)
XXYY      (0.04530261550868937+0j)
YYYY      (0.04530261550868937+0j)
XXXX      (0.04530261550868937+0j)
YYXX      (0.04530261550868937+0j)
ZIIZ      (0.16592785032250779+0j)
IZZI      (0.16592785032250779+0j)
ZIZI      (0.17441287610651643+0j)
ZZII      (0.12062523481381844+0j)
```

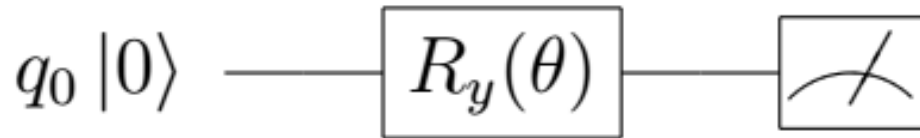

Ansatz Preparation and trial wavefunction

- General rotation ansatz

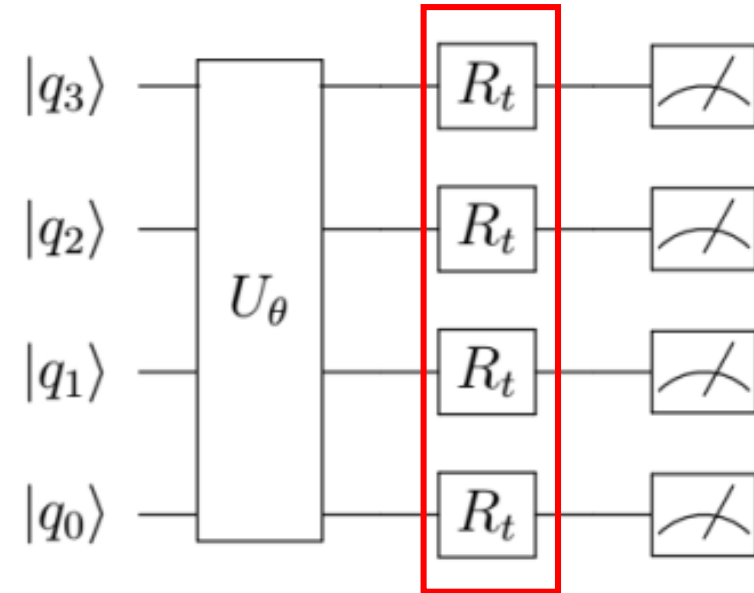


Measurements

- Expectation value is easy to evaluate on a quantum computer
- How to measure the expectation value on a quantum computer ?
(a practice problem)



probability difference $P(+1) - P(-1)$.



Post-rotations to Z-basis

Variational Quantum Eigensolver (VQE)

IBM Q

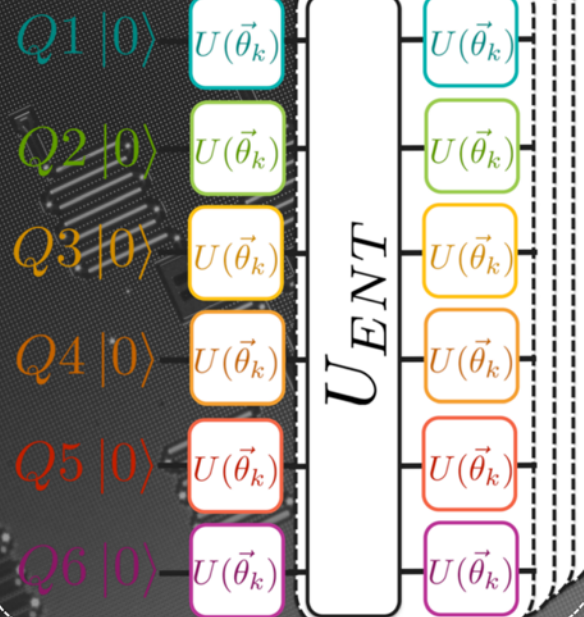
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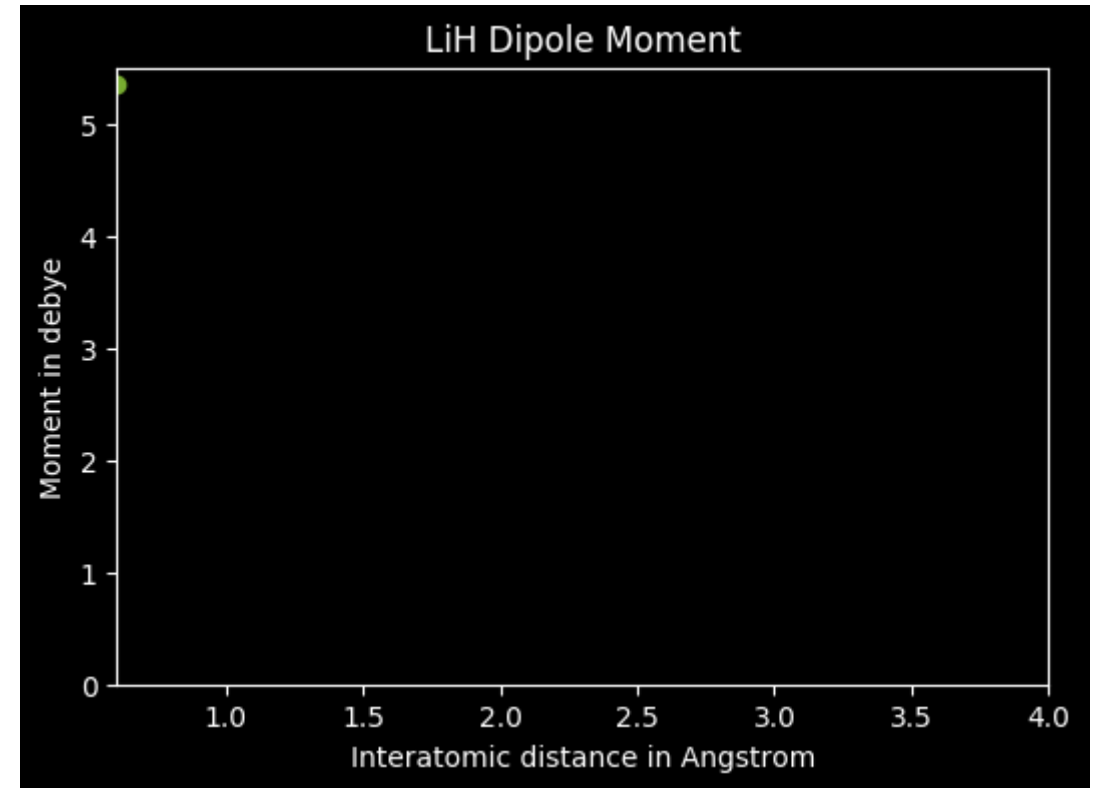
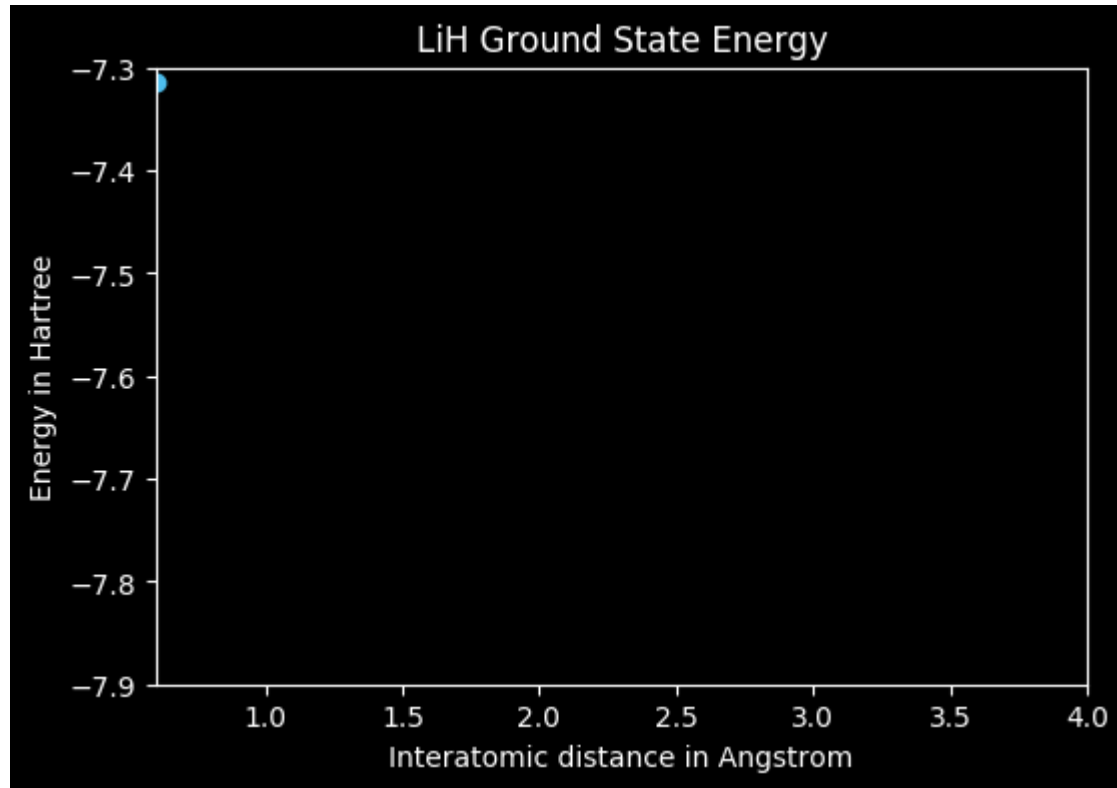
$$\mathbf{E}(\theta, \phi, \lambda, \dots) = \\ -0.206 \langle \text{IIIIII} \rangle \\ -0.002 \langle \text{XZXXZZ} \rangle \\ +0.028 \langle \text{XXXIIX} \rangle \\ \dots$$

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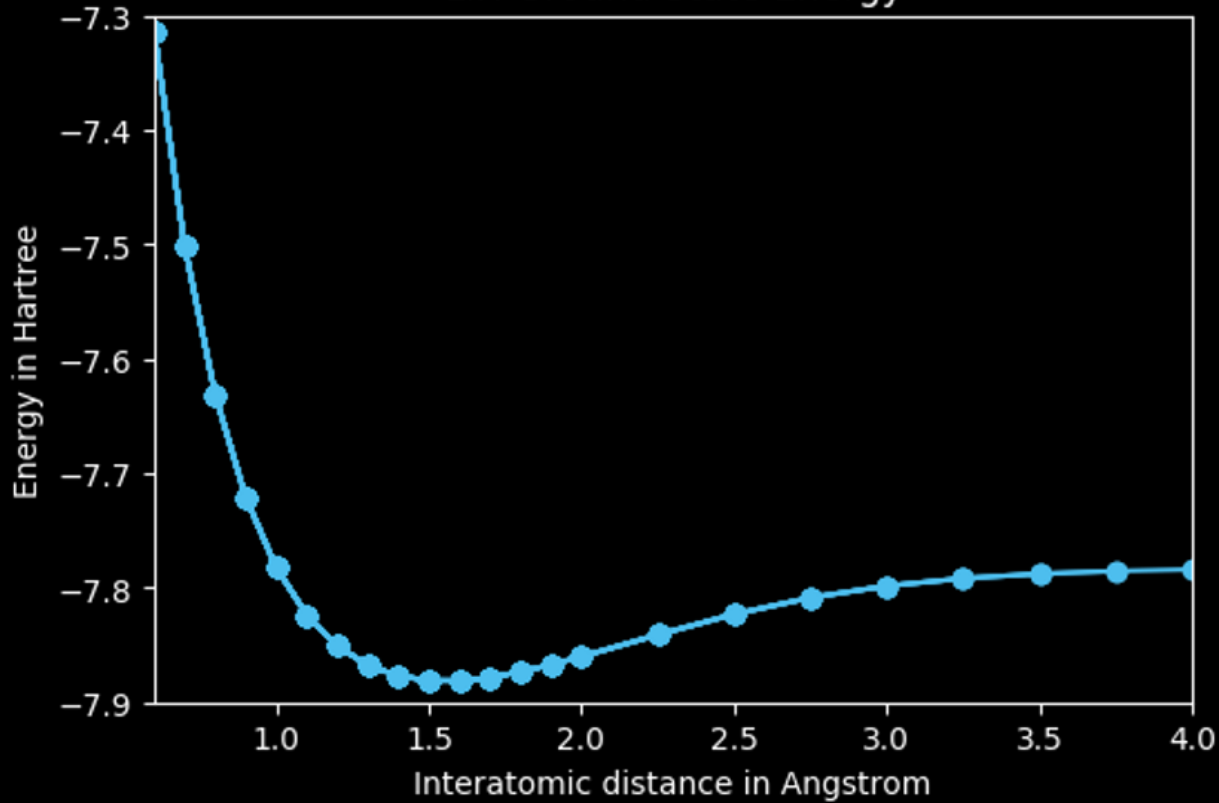
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Simulation results for LiH



Simulation results for LiH

LiH Ground State Energy



LiH Dipole Moment

