

Exact and efficient Lanczos* method on a quantum computer

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Goal: estimate ground state energy of quantum Hamiltonian.

Abstract perspective: approximate lowest eigenvalue of Hermitian matrix.

Example applications:

- Quantum chemistry.
- Condensed matter physics.
- High-energy physics.
- Nuclear physics.

Goal is classically challenging due to exponential Hilbert space dimension.

Lanczos* method

Lanczos method = classical method for approximating lowest eigenvalues:

- 1 Initial guess $|\psi_0\rangle \Rightarrow H|\psi_0\rangle \Rightarrow H^2|\psi_0\rangle \dots \Rightarrow H^{D-1}|\psi_0\rangle$.
- 2 $(\mathbf{H}, \mathbf{S}) = \text{project } H \text{ onto span}\{|\psi_0\rangle, H|\psi_0\rangle, H^2|\psi_0\rangle, \dots, H^{D-1}|\psi_0\rangle\}$.
- 3 lowest eigenvalue of $(\mathbf{H}, \mathbf{S}) \sim$ lowest eigenvalue of H .

Advantage: converges exponentially with D (in ∞ precision arithmetic).

Disadvantage: classically, requires storing entire statevectors $H^i|\psi_0\rangle \Rightarrow$ exponential overhead.

Can we construct a quantum version that mitigates the overhead to represent statevectors while keeping fast convergence?¹

¹Klymko *et al.* (2022), Epperly *et al.* (2022), and many more.

Block encoding


Idea: for H on n qubits (s.t. $\|H\| \leq 1$), find U on $m + n$ qubits s.t.²


$$U = \begin{pmatrix} \begin{matrix} \color{green} & & \\ & H & \\ & & \cdot \end{matrix} & \begin{matrix} \color{red} & & \\ & & \cdot \end{matrix} \\ \begin{matrix} \color{red} & & \\ & & \cdot \end{matrix} & \begin{matrix} \color{red} & & \\ & & \cdot \end{matrix} \end{pmatrix}, \quad R = \begin{pmatrix} \begin{matrix} \color{green} & & \\ & 1 & \\ & \ddots & \\ & & 1 \end{matrix} & \begin{matrix} \color{red} & & \\ & & 0 \end{matrix} \\ \begin{matrix} \color{red} & & \\ & & 0 \end{matrix} & \begin{matrix} \color{red} & & \\ & -1 & \\ & & \ddots \\ & & & -1 \end{matrix} \end{pmatrix}$$

$$U^2 = 1 \quad \rightarrow \quad (RU)^j = \begin{pmatrix} \begin{matrix} \color{green} & & \\ & T_j(H) & \\ & & \cdot \end{matrix} & \begin{matrix} \color{red} & & \\ & & \cdot \end{matrix} \\ \begin{matrix} \color{red} & & \\ & & \cdot \end{matrix} & \begin{matrix} \color{red} & & \\ & & \cdot \end{matrix} \end{pmatrix}$$

²Low and Chuang, Quantum 3 (2019).

Quantum Lanczos method


$$(RU)^j = \begin{pmatrix} T_j(H) & \cdot \\ \cdot & \cdot \end{pmatrix}$$



can use block encoding to exactly construct $T_j(H)|\psi_0\rangle$.

Recall: Lanczos method \sim project H onto

$$\begin{aligned} & \text{span}\{|\psi_0\rangle, H|\psi_0\rangle, H^2|\psi_0\rangle, \dots, H^{D-1}|\psi_0\rangle\} \\ & = \text{span}\{|\psi_0\rangle, T_1(H)|\psi_0\rangle, T_2(H)|\psi_0\rangle, \dots, T_{D-1}(H)|\psi_0\rangle\}. \end{aligned}$$

Quantum Lanczos method

To diagonalize H projected onto subspace, estimate

$$\mathbf{H}_{ij} := \langle \psi_0 | T_i(H) H T_j(H) | \psi_0 \rangle, \quad \mathbf{S}_{ij} := \langle \psi_0 | T_i(H) T_j(H) | \psi_0 \rangle$$

on quantum computer for $i, j = 0, 1, 2, \dots, D - 1$, then solve

$$\mathbf{H}\vec{v} = \lambda\mathbf{S}\vec{v}.$$

Required quantities

So the quantities we need are

$$\begin{aligned}\langle \psi_0 | T_i(H) H T_j(H) | \psi_0 \rangle &= \frac{1}{4} \left(\langle T_{i+j+1}(H) \rangle_0 + \langle T_{|i+j-1|}(H) \rangle_0 \right. \\ &\quad \left. + \langle T_{|i-j+1|}(H) \rangle_0 + \langle T_{|i-j-1|}(H) \rangle_0 \right), \\ \langle \psi_0 | T_i(H) T_j(H) | \psi_0 \rangle &= \frac{1}{2} \left(\langle T_{i+j}(H) \rangle_0 + \langle T_{|i-j|}(H) \rangle_0 \right)\end{aligned}$$

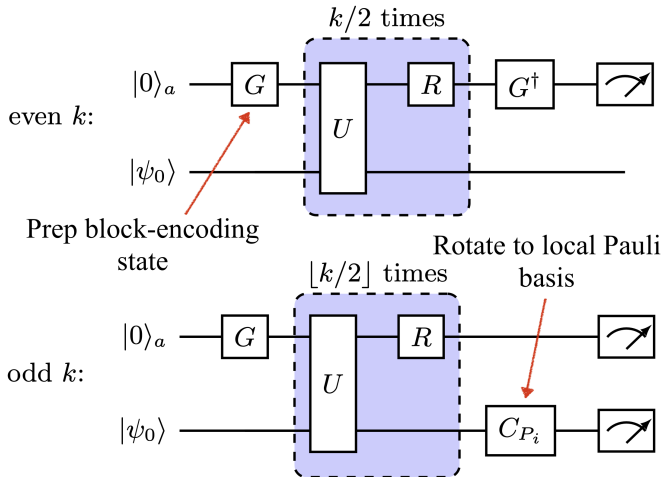
for $i, j = 0, 1, 2, \dots, D - 1$. In other words, need to estimate

$$\langle T_k(H) \rangle_0 := \langle \psi_0 | T_k(H) | \psi_0 \rangle$$

for each $k = 0, 1, 2, \dots, 2D - 1$.

Required circuits

Hence, required circuits are: for each $k = 0, 1, 2, \dots, 2D - 1$,



where $G|0\rangle_a$ identifies the block in the block encoding.

“In practice” error analysis⁴

Can bound energy error \mathcal{E} as function of noise rate, Krylov space dimension, problem parameters.

Assuming noise only from finite shots, to reach energy error \mathcal{E} we require

$$D = \Theta \left[\left(\log \frac{1}{|\gamma_0|} + \log \frac{1}{\mathcal{E}} \right) \min \left(\frac{1}{\mathcal{E}}, \frac{1}{\Delta} \right) \right]$$

and

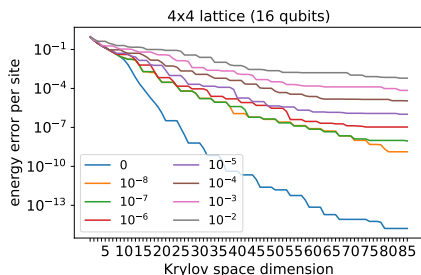
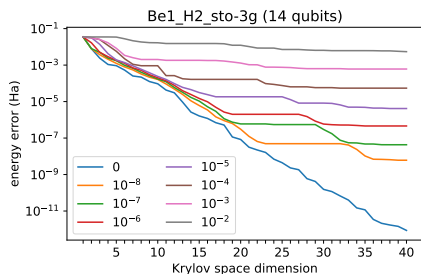
$$M = \Theta \left[D \left(\frac{1}{\mathcal{E}^2} + \frac{1}{\mathcal{E}|\gamma_0|^4} \right) \right].$$

measurements in total, for initial state overlap γ_0 and spectral gap Δ .³

³Uses analysis based on Epperly *et al.*, 2022.

⁴“In practice” because theoretical bound only guarantees first term in M is \mathcal{E}^{-p} for $p \in [2, 3]$; $p = 2$ is based on numerics.

Numerical examples



Left plot shows energy error versus Krylov space dimension for BeH_2 in STO-3G, equilibrium configuration (bondlength 1.3264 Å), with different noise rates per matrix element. Right plot shows energy error per site versus Krylov space dimension for J1-J2 model on 4×4 square lattice with couplings $J_1 = 1$ and $J_2 = 0.5$.

Thank you!

Preprint: <https://arxiv.org/abs/2208.00567>