### Lanczos Algorithm for Qiskit Dynamics

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# $|\psi(t)\rangle = e^{-iHt}|\psi(0)\rangle$



#### Why Lanczos

• Local Hamiltonians describing qubit systems are sparse

$$
H = \begin{pmatrix} 0 & 0 & \alpha & \beta \\ 0 & d_0 & 0 & 0 \\ \alpha^* & 0 & 0 & 0 \\ \beta^* & 0 & 0 & d_1 \end{pmatrix}
$$

- Calculating time evolution requires exponentiating the Hamiltonian  $|\psi(t)\rangle = e^{-iHt}|\psi_0\rangle$
- This can be done by diagonalizing  $H$

 $e^{-iHt} = S^{\dagger}e^{-iDt}S$ 









Time evolution requires only sparse matrix - vector multiplication  $\overline{|\psi(t)\rangle} = e^{-iHt}|\psi_{0}\rangle$ 

$$
(\mathcal{Y})^{\prime} = \sum_{n} \frac{(-it)^n}{n!} H^n |\psi_0\rangle
$$

$$
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$$

 $|u_0\rangle = |\psi\rangle$   $|u_1\rangle = H|u_0\rangle$   $|u_2\rangle = H|u_1\rangle$  ...

### Why Lanczos (Krylov Subspace)



- $K_r = \{ |\psi\rangle, A|\psi\rangle, A^2 |\psi\rangle, A^3 |\psi\rangle ... A^{k-1} |\psi\rangle \}$  Is the krylov subspace for a given matrix A and vector  $|\psi\rangle$  of order k
- One can construct a basis  $\{|\phi_i\rangle\}$  for this subspace using Gram-Schmidt

$$
|\widetilde{\phi}_{k-1}\rangle = |u_{k-1}\rangle - \sum_{i} \langle \phi_i | u_{k-1} \rangle | \phi_i \rangle
$$

$$
|\phi_{k-1}\rangle = \frac{|\widetilde{\phi}_{k-1}\rangle}{\langle \widetilde{\phi}_{k-1}|\widetilde{\phi}_{k-1}\rangle}
$$

## Why Lanczos (Krylov Subspace)



- One can construct an orthogonal matrix  $Q_{n,k}$  With  $|\phi_i\rangle$  as the columns such that  $T_{k,k} = Q_{k,n}^{\dagger} H_{n,n} Q_{n,k}$
- Where  $T$  is a Tridiagonal matrix
- Diagonalizing this Tridiagonal matrix is a lot faster since typically,  $k \ll n$
- The Eigen-vectors of T is then an approximation of the lowest k Eigen vectors of H
- Therefore, we have  $V_n = Q_{n,k} V_k$  where  $V_{n,k}$  are the eigenvectors of T

#### Lanczos vs NumPy

#### (ground state calculation)

#### (RunTime)<sub>100</sub> vs Array Dimension



# Lanczos Time-evolution



• Once we have the basis vectors and the Tridiagonal projection, we have the equation

$$
H_{n,n} = Q_{n,k} T_{k,k} Q_{k,n}^{\dagger}
$$

- Thus, the time evolution unitary becomes
- $e^{-iHt} = e^{-iQTQ^{\dagger}t} = Qe^{-iTt}Q^{\dagger} = QVe^{-i\text{diag}(T)t}V^{\dagger}Q^{\dagger}$
- If we had chosen the initial vector of the Lanczos iteration to be same as the initial state which we want to evolve, then the rows of  $Q$  (other than first) are orthogonal to  $|\psi_i\rangle$

# Lanczos Time-evolution



- If we had chosen the initial vector of the Lanczos iteration to be same as the initial state which we want to evolve, then the rows of  $Q$  (other than first) are orthogonal to  $|\psi_i\rangle$
- $e^{-iHt} = QV_k e^{-i\textbf{diag}(T)t} V_k^{\dagger} Q^{\dagger} |\psi_i\rangle$

$$
=QV_k e^{-i \textbf{diag}(T)t} V_k^{\dagger} \delta_{0,k}
$$

$$
=QV_{k}e^{-i\text{diag}(T)t}V_{0}^{\dagger}=QV_{k}\exp(-iE_{T}t)|\psi_{i}\rangle
$$

This increases the accuracy of the simulation since we aren't affected by loss of orthogonality.

### The PR



 $\odot \cdots$ 

#### Implementing Lanczos algorithm as a new solver method #109



Q Conversation 1

 $\overline{\bullet}$  Commits 27  $\overline{\bullet}$  Checks 0

 $\boxed{\pm}$  Files changed 5



rupeshknn commented 3 days ago · edited •

#### **Summary**

Lanczos algorithm is an approximate diagonalisation method. It is implemented as an LMDE method and is a considerable speedup compared to scipy.expm

#### Details and comments

This PR adds a new fixed-step solver method lanczos diag. This method only works with hermitian generators and works best in sparse evaluation mode. A follow up with a Jax implementation of the same is in progress.

### Thank You



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GitHub: github.com/rupeshknn/lanczos-QD