

// Quantum Science Generation – ECT*

Early Fault-Tolerant Quantum Algorithms in Practice: Application to Ground-State Energy Estimation

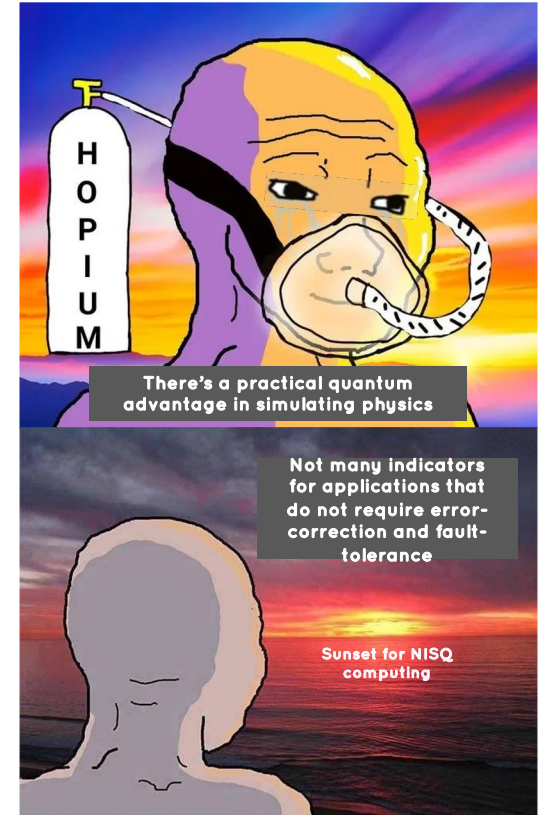
Oriel Kiss,
May 8th, 2024

// Motivation

We want to solve a **useful** problem on a quantum computer with an advantage.

Our bet is on **quantum spin models**, but scaling looks a bit scary in terms of precision. Childs, et al. PNAS 115.38 (2018): 9456-9461.

NISQ	Early-FTQC	Fully-FTQC
Variational Algorithms (VQE, QAOA)	ISQ Algorithms (Lin-Tong, Somma, ...)	Phase Estimation (QPE, QETU, filters, ...)
No theoretical arguments for quantum advantage	Optimized Runtime, samples-depth tradeoff, number of qubits	T-gate counts, High-depth Coherent computation
Scaling - ϵ^{-2}	Scaling - $\sim < \epsilon^{-2}$ Prefactor matters!	Scaling - ϵ^{-1}



// Overview

1. Early fault-tolerant algorithms for GSEE: review of the Lin&Tong algorithm.
2. Applying the Lin&Tong algorithm in practice.
3. Numerical simulations.
4. Bonus: application on NISQ devices.

// Early-FTQC Algorithms for GSEE

Find the lowest eigenvalue of a **Hamiltonian H** describing a system with some error ϵ ,

$$\tau H = \sum_k \tau_k |E_k\rangle\langle E_k| \text{ with } \|\tau H\| < \pi/2$$

An Early-FTQC algorithm solving this problems should have following **nice** features:

1. It uses limited (ideally one) or constant ancilla qubits.
2. The circuit depth is $O(\epsilon^{-1})$, i.e., it enjoys Heisenberg scaling.
3. Some degree of robustness against algorithmic errors.

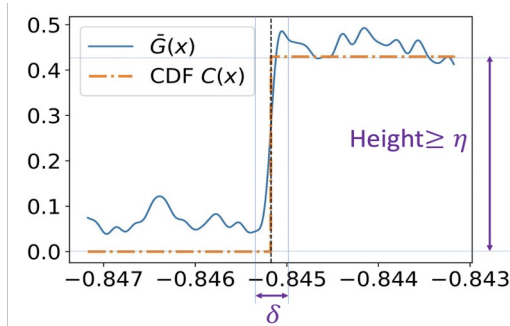
// The Lin-Tong Algorithm

Problem 1. Given a precision $\delta > 0$ and lower bound on the overlap parameter $\eta > 0$, we seek to decide if

$$\text{Tr}[\rho \Pi_{\leq x-\delta}] < \eta \quad \text{or} \quad \text{Tr}[\rho \Pi_{\leq x+\delta}] > 0. \quad (3)$$

The LT algorithm does this by approximating the CDF of the spectral measure -

$$p(x) = \sum_k p_k \tilde{\delta}(x - \lambda_k) = \sum_k \text{Tr}[\rho \Pi_k] \tilde{\delta}(x - \lambda_k),$$



Construction:

$$C(x) = p(x) * \Theta(x) \quad \text{convolution}$$

$$\begin{aligned} \tilde{C}(x) &= \int_{-\pi/2}^{\pi/2} p(y) F(x-y) dy \\ &= \sum_{|k| \leq D} F_k e^{ikx} \underbrace{\langle \Psi | e^{-i\tau k \mathcal{H}} | \Psi \rangle}_{\text{Fourier moments}} \end{aligned}$$

Fourier moments computed on the QC with a Hadamard test

$$G(x) = \frac{1}{2} + \frac{2\mathcal{F}}{M} \sum_{i=1}^M \left[\text{Re}[g_{k_i}(\tau)] \sin(k_i x) + \text{Im}[g_{k_i}(\tau)] \cos(k_i x) \right].$$



Prefactor ?

THIS IS
FINE

Theorist

Hamiltonian
Simulation

Unknown
overlap

Initial State
Preparation

// Overview

1. Early fault-tolerant algorithms for GSEE: review of the Lin&Tong algorithm.
2. **Applying the Lin&Tong algorithm in practice.**
3. Numerical simulations.
4. Bonus: application on NISQ devices.

// Extinguishing the fire

- † **Initial State Preparation:** **improve** from the best you can do classically: DMRG, Coupled Cluster, ...
- † **Unknown Overlap:** finding the **inflection** point and looking at the accumulation over the low energy sector.

Accumulation:

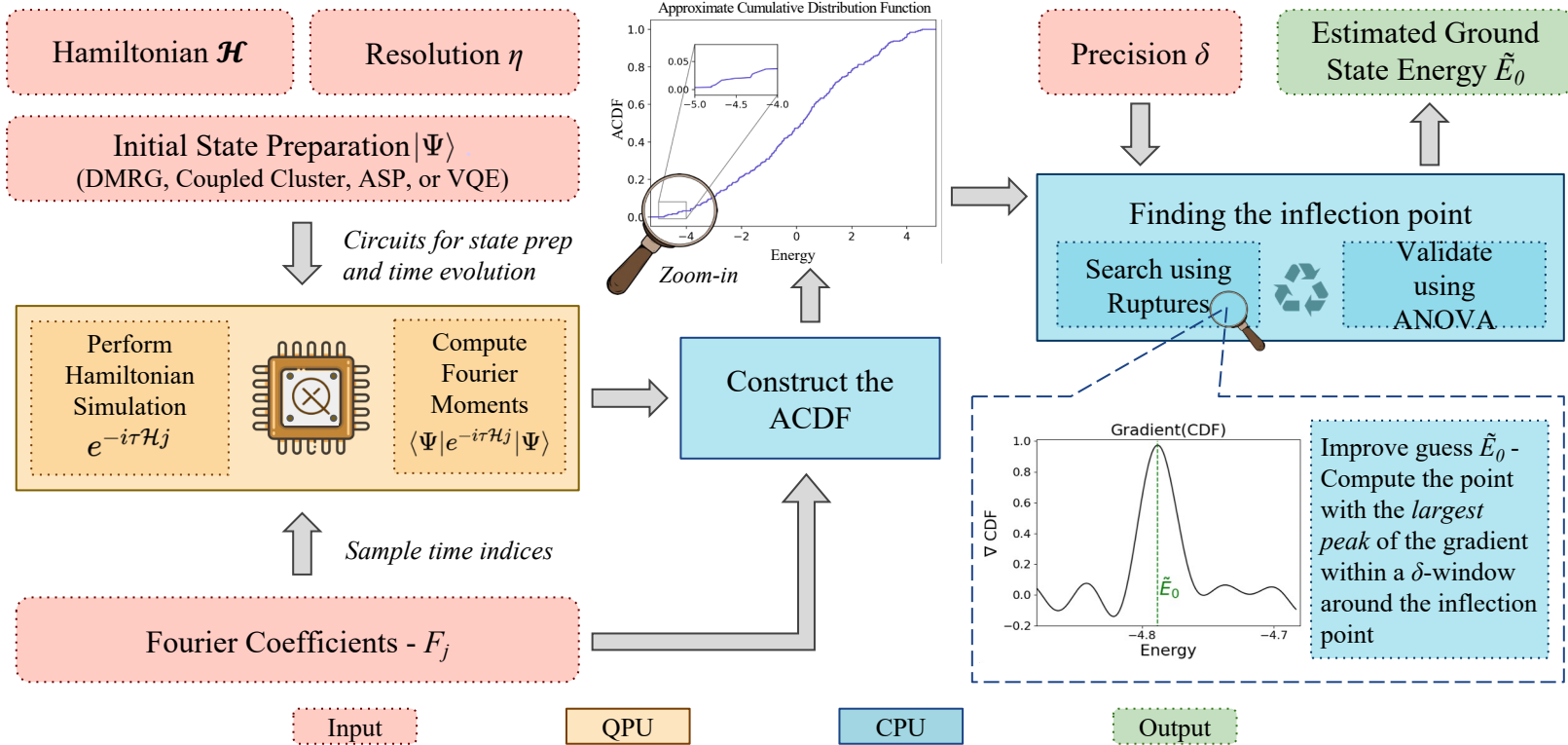
$$\eta = \sum_{i < m} |\langle E_i | \Psi \rangle|^2,$$

- † **Hamiltonian Simulation:** **system-specific**; we use (asymptotical sub-optimal) product formula because:

1. no ancilla overhead
2. take advantage of locality
3. often much better than what is guaranteed
4. can scale better than qubitization in some regime

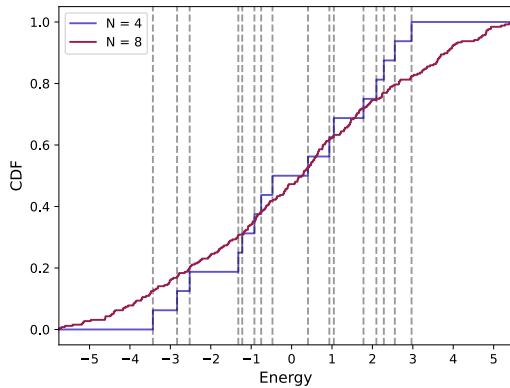


// Our Workflow

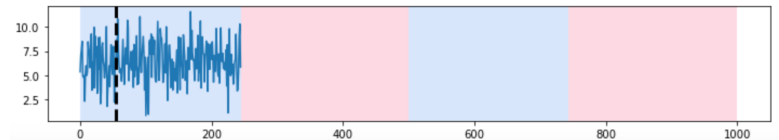
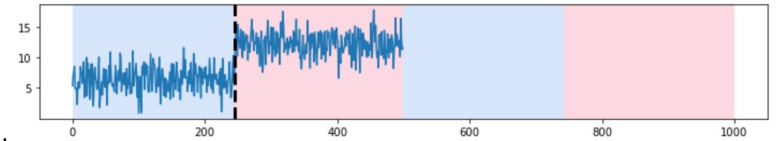
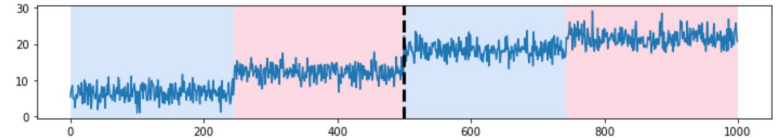
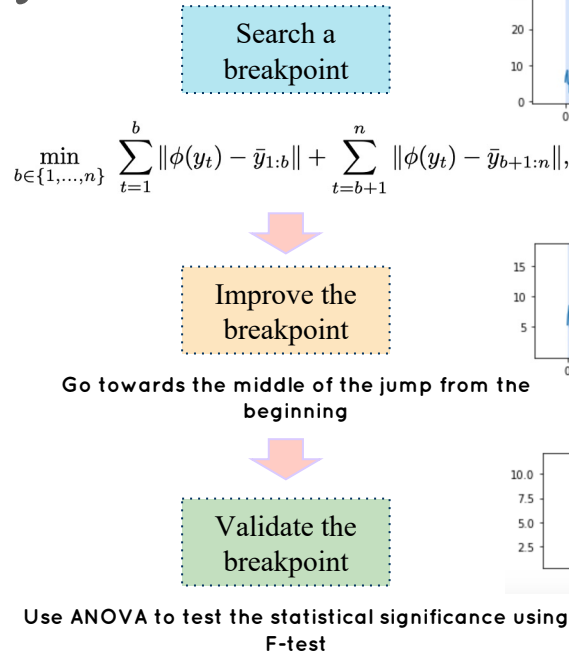


// Key Insights I: Find Inflection

CDF smoothes out with system size
(random initial state)



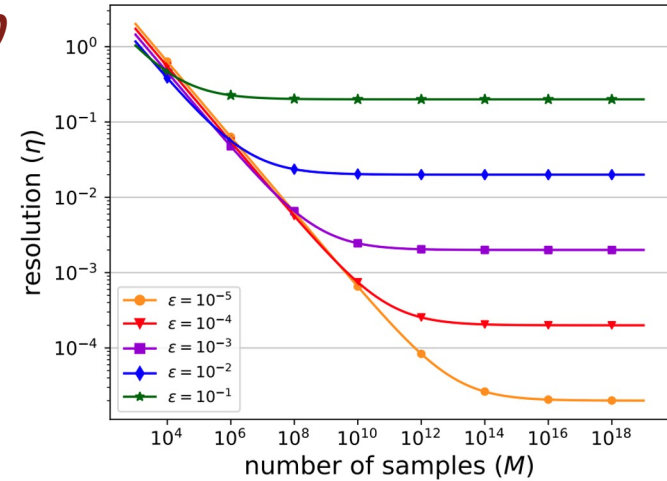
XXZ model



// Key Insights II: Quantitative Resources

For a given **maximal runtime D** $O(\delta^{-1} \log \delta^{-1} \eta^{-1})$ and accuracy ϵ , the number of samples M required to guarantee the correct result with probability $1 - \vartheta$ is

$$M = \left[2 \cdot \left[\frac{2.07\pi^{-1}(\log 4D + 1) + 1}{\eta - 2\epsilon} \right]^2 \cdot \left[\log \log \left(\frac{1}{\tau\epsilon} \right) + \log(\vartheta^{-1}) \right] \right],$$



To obtain high accuracy, we require more depth (and not only samples).

// Overview

1. Early fault-tolerant algorithms for GSEE: review of the Lin&Tong algorithm.
2. Applying the Lin&Tong algorithm in practice.
3. **Numerical simulations.**
4. Bonus: application on NISQ devices.

// Numerical Simulations

We consider a fully-connected **Heisenberg model** with random couplings over N spins

$$\mathcal{H} = \frac{1}{N} \sum_{i < j} \sum_{a \in \{x, y, z\}} J_a^{ij} \cdot \sigma_a^i \sigma_a^j, \text{ where } J_a^{ij} \sim \mathcal{N}(0, 1).$$

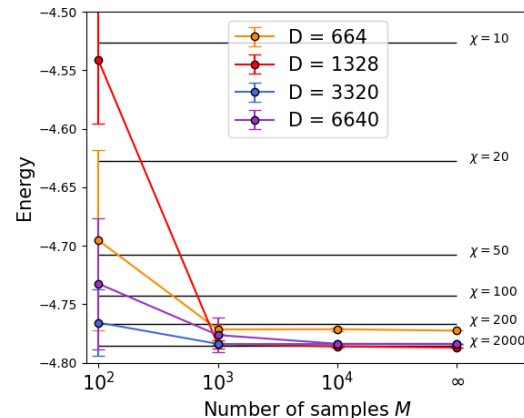
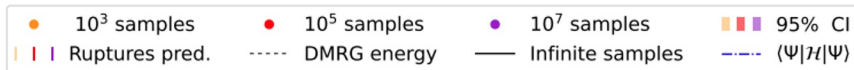
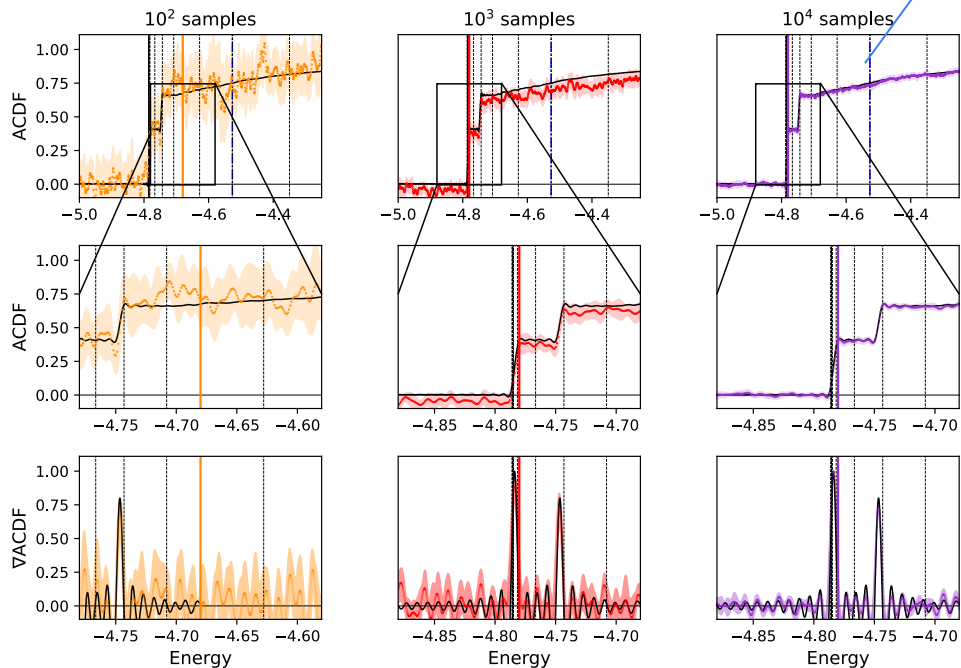
Why? : universal, gapless, challenging for DMRG.

For dynamics we use a **second-order Trotter-Suzuki** with time step $\Delta t = \tau/8$.
The circuit construction is based on SWAP-networks.

We look at **$N=26$** spins,
and use initial states prepared via DMRG (with low bond dimension).

// Key Insights III: DMRG(bd = 10) Initial State

5% improvement

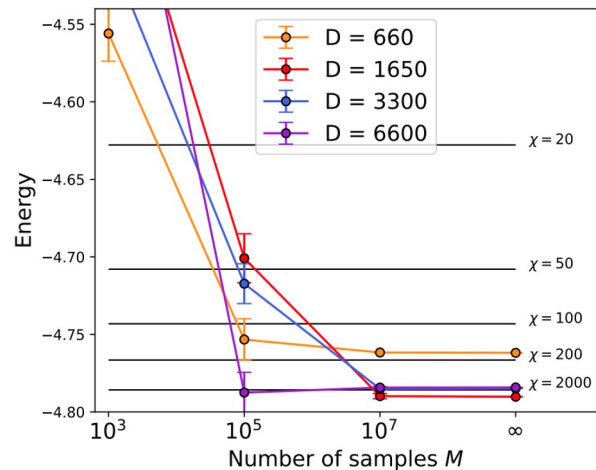
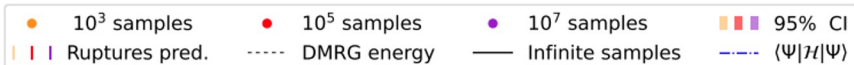
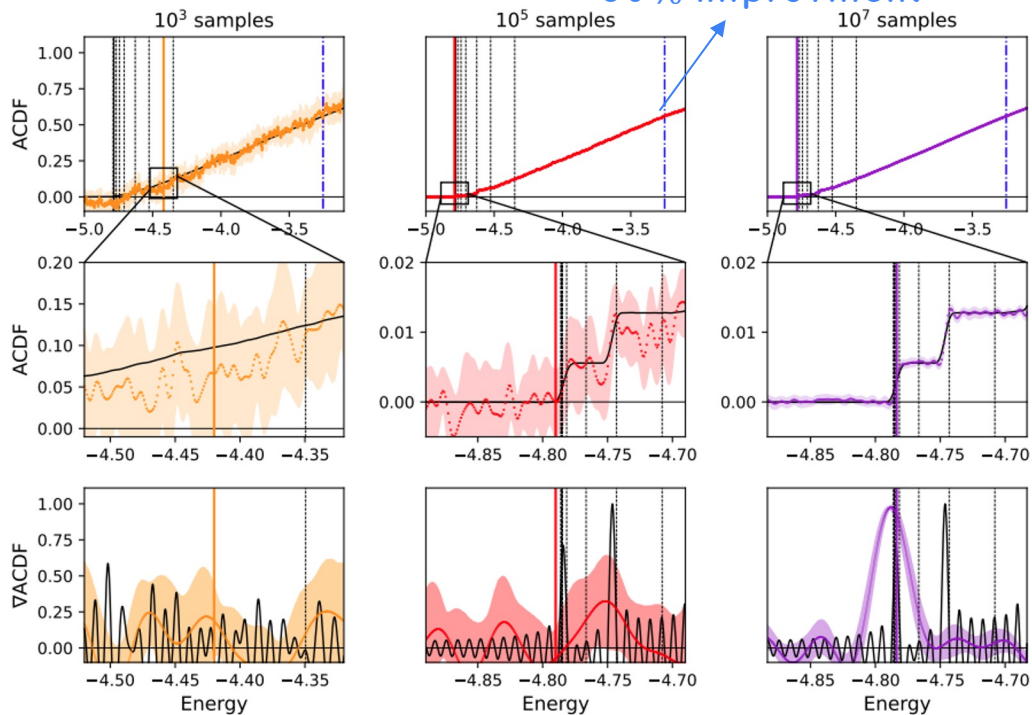


Depth matters most!

$p_0 = 7 \times 10^{-6}$ (overlap)
 $M = 10^{13}$ samples

// Key Insights III: Sparsified DMRG

30% improvement



Fidelity with
initial state
 $p_0 = 7 \times 10^{-3}$

// To Conclude

1. LT algorithm (and friends) **bridges** the gap between the NISQ and FTQC eras.
2. Instead of aiming at the true ground-state energy, one can concentrate on a finding the **inflection point** of the spectral CDF.
3. Product formulas are solid candidates for this task.
4. LT algorithms are able to **improve on classical solutions**, using only limited quantum resources ($\sim 10^5$ samples).
5. If arbitrary precision is required, use quantum phase estimation. However, if a **good approximation is enough**, LT is a robust algorithm, which can be run **in practice**.



// Overview

1. Early fault-tolerant algorithms for GSEE: review of the Lin&Tong algorithm.
2. Applying the Lin&Tong algorithm in practice.
3. Numerical simulations.
4. **Bonus: application on NISQ devices.**

// Bonus: what can we do on NISQ?

- Blunt et al: variational dynamics + ZNE on a H_3 molecule (6 qubits).
- We focus on extracting Fourier moments of a nuclear EFT (4 qubits).

$$\langle \Psi_0 | \hat{O}(\vec{q})^\dagger e^{-iHj\tau} \hat{O}(\vec{q}) | \Psi_0 \rangle$$

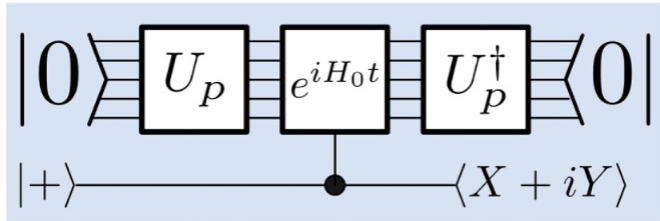
- We use **purified echo verification** + various error suppression techniques (twirling, dynamical decoupling, pulse efficient calibration, readout calibration).



// Purified echo verification

Verification

Unprepare the state and verify that it is the zero state



Verification passed \Rightarrow state contributes +/- 1 to the expectation value.

Otherwise \Rightarrow garbage

Purification of the ancilla

$$\rho = \lambda |\psi\rangle\langle\psi| + (1 - \lambda) \underbrace{\sum_{k=2}^{2^N} p_k |\psi_k\rangle\langle\psi_k|}_{\text{Noisy components}}.$$

Pure main component

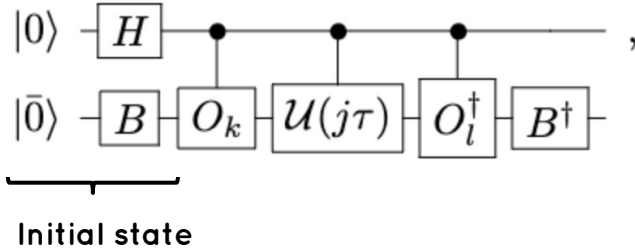
Noisy components

Without noise: the ancilla is pure after post-selection.

With noise: It is not. Extract the closest pure state from measurements.

// Purified echo verification (details)

Hadamard test



$$|\Phi\rangle = \frac{1}{\sqrt{2}} \left(|\bar{0}\rangle \otimes |0\rangle + B^\dagger O_k^\dagger \mathcal{U}(j\tau) O_l B |\bar{0}\rangle \otimes |1\rangle \right)$$

$$\equiv \frac{1}{\sqrt{2}} (|\bar{0}\rangle \otimes |0\rangle + |\phi\rangle \otimes |1\rangle),$$

$$|\phi\rangle = \alpha |\bar{0}\rangle + \beta |\bar{0}^\perp\rangle.$$

1. We measure the 3 single-qubit Pauli expectation (X,Y and Z) values of the ancilla.

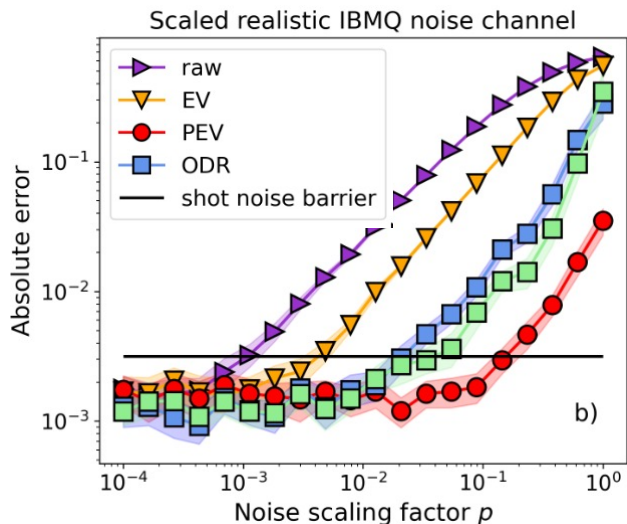
We only care about α , so we can disregard any orthogonal states!

2. Construct the closest compatible pure state (purification + tomography).

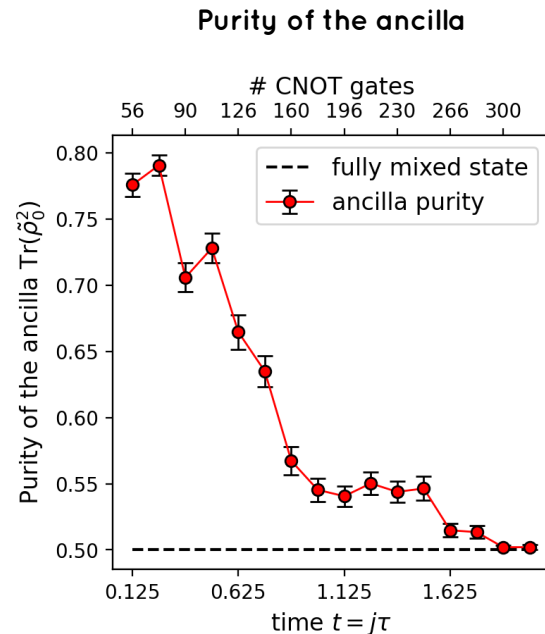
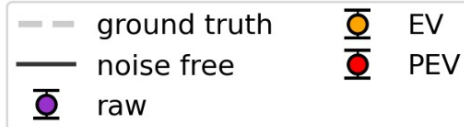
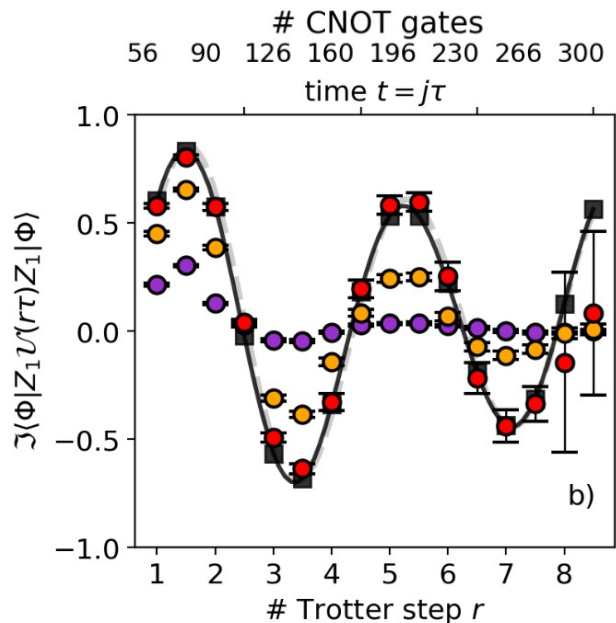


$$\text{Re}\{\alpha\} = \frac{\langle X_a \rangle_0}{1 + \langle Z_a \rangle_0}, \quad \text{Im}\{\alpha\} = \frac{\langle Y_a \rangle_0}{1 + \langle Z_a \rangle_0}.$$

// Results (IBMQ)



x100 error reduction



We know the
breakpoint

Thank You!

More Questions?

// Team



Utkarsh Azad



Borja Requena



David Wakeham



Michele
Grossi



Alessandro
Roggero



Juan Miguel
Arrazola

arXiv:
2405.03754



arXiv:
2401.13048



contact: oriel.kiss@cern.ch
twitter: [@oriel_kiss](https://twitter.com/oriel_kiss)

// Key Contributions

- **Relaxation:** finding inflection point and looking at the accumulation over the low-energy sector.

Accumulation:
$$\eta = \sum_{i < m} |\langle E_i | \Psi \rangle|^2,$$

- Gather **quantitative** resource estimates for the LT-algorithm.
- Looking at the **resilience** of the LT-algorithm against the quality of initial state preparation and time evolution.