

Classical simulations of noisy variational quantum circuits

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Motivations for VQAs

It can be argued that there are three main motivations for VQAs:

1. Ease of implementation

No fault-tolerance required. Respects native gateset, topology.

2. Usefulness

Can solve interesting problems. Not trivially simulatable.

3. Noise resilience

Conjectured [1] and observed resilience to noise, coherent [2] and incoherent [3]. Amenable to quantum error mitigation (QEM) techniques.

[1] McClean, Jarrod R., et al. "The theory of variational hybrid quantum-classical algorithms." New Journal of Physics 18.2 (2016): 023023. [2] O'Malley, Peter JJ, et al. "Scalable quantum simulation of molecular energies." Physical Review X 6.3 (2016): 031007. [3] Sharma, Kunal, et al. "Noise resilience of variational quantum compiling." New Journal of Physics 22.4 (2020): 043006.

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Mixed evidence of partial resilience, at best.

- Basic numerical experiments found that there was some resilience, but only for very low noise rates [1].
- It also appeared that noise tends to accumulate -> noise-induced barren plateaus [2].
- For any QEM technique, overhead scales worst-case exponentially in both circuit depth and width [3].

Even with QEM, asymptotically we cannot get rid of the noise

[1] Fontana, Enrico, et al. "Evaluating the noise resilience of variational quantum algorithms." Physical Review A 104.2 (2021) [2, image] Wang, Samson, et al. "Noise-induced barren plateaus in variational quantum algorithms." Nature communications 12.1 (2021) [3] Takagi, Ryuji, et al. "Fundamental limits of quantum error mitigation." npj Quantum Information 8.1 (2022)

Does noise prevent quantum advantage?

- Heuristically: MPS with truncation \approx noisy QC [1].
- VQE with sufficient noise can be classically simulated via Gibbs states [2].
- This suggests that as noise accumulates, at some point a **transition** occurs and "quantumness" disappears.

Q: In VQAs, at what point is 'quantumness' lost?

[1] Zhou, Yiqing, E. Miles Stoudenmire, and Xavier Waintal. "What limits the simulation of quantum computers?." Physical Review X 10.4 (2020) [2] Stilck França, Daniel, and Raul Garcia-Patron. "Limitations of optimization algorithms on noisy quantum devices." Nature Physics 17.11 (2021)

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Fourier <

VQA cost functions are Fourier series

It is well-known [1,2,3,4] that VQA cost functions admit a Fourier series representation:

$$
\rho_0 - U(\theta) \longrightarrow Tr\{\rho(\theta)O\} = \sum_{\omega \in \Omega} a(\omega)e^{i\omega \cdot \theta} = \sum_{\omega_{\theta}, \omega_{\theta} \dots} \underbrace{a(\omega_1)}_{a(\omega_0)} \cdot \underbrace{\bigvee}_{a(\omega_0)} \cdot \underbrace{\
$$

- The frequencies ω live in a discrete, finite (but large) spectrum Ω .
- Spectrum is set of pairwise differences of eigenvalues of rotations:

$$
R_z(\theta) = e^{-i\sigma_z \theta/2} \to \lambda(\sigma_z/2) = {\pm 1/2} \to \Omega = {-1, 0, +1}
$$

[1] Schuld, Maria, Ryan Sweke, and Johannes Jakob Meyer. "Effect of data encoding on the expressive power of variational quantum-machine-learning models." Physical Review A 103.3 (2021): 032430.

[2] Gil Vidal, Francisco Javier, and Dirk Oliver Theis. "Input redundancy for parameterized quantum circuits." Frontiers in Physics 8 (2020): 297.

[3] Nakanishi, Ken M., Keisuke Fujii, and Synge Todo. "Sequential minimal optimization for quantum-classical hybrid algorithms." Physical Review Research 2.4 (2020): 043158.

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Q: What happens when noise is present?

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Channel mode decomposition

• To include noise, move to channels.

$$
\mathbf{Rz}(\theta) = \mathbf{C}_0 + e^{i\theta} \mathbf{C}_1 + e^{-i\theta} \mathbf{C}_{-1}
$$

- Now go from Fourier to **trigonometric decomposition** for single-Pauli rotation gate channel: $\mathbf{Rz}(\theta) = \mathbf{D}_0 + \cos(\theta)\mathbf{D}_1 + \sin(\theta)\mathbf{D}_{-1}$
- With noise, the channel modes $\{D_0, D_{\pm 1}\}$ will depend on the error parameters.
- Single qubit Pauli noise following rotation give PTM:

$$
\left(\frac{1}{\sqrt{2}}\right)^{1/2}
$$

$$
\mathbf{D}_0 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & q_Z \end{pmatrix}, \ \ \mathbf{D}_1 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & q_X & 0 & 0 \\ 0 & 0 & q_Y & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}, \ \ \mathbf{D}_{-1} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & -q_X & 0 \\ 0 & q_Y & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}
$$

Decomposition gives PTMs that map 1 Pauli into ≤1 Paulis

Noisy Heisenberg-picture simulation

• Evolving a Pauli measurement in the Heisenberg picture, we obtain a trigonometric decomposition of the cost function that includes noise factors.

$$
U(\boldsymbol{\theta}) = \prod_{i=1}^{m} (C_i \cdot R_z(\theta_i)) \cdot C_0
$$

$$
\mathbf{Rz}(\boldsymbol{\theta}) = \mathbf{D}_0 + \cos(\boldsymbol{\theta}) \mathbf{D}_1 + \sin(\boldsymbol{\theta}) \mathbf{D}_{-1}
$$

Noisy Rz

• The cost function (expectation value of Pauli operator) is

$$
f(\boldsymbol{\theta}) = \sum_{\boldsymbol{\omega}} \Phi_{\boldsymbol{\omega}}(\boldsymbol{\theta}) \langle \!\langle 0 | \mathbf{U}_{\boldsymbol{\omega}}^{\dagger} | O \rangle \!\rangle
$$

contains sin/cos and noise Constant (± 1 or 0)
Classically simultaneously

Low-weight efficient simulation algorithm (LOWESA)

Why this algorithm?

- Similar algorithms had been developed before [1, 2].
- Same algorithm developed in parallel by Nemkov et al. [3] and Begusic et al. [4] (different truncation scheme).
- However, these versions had no noise.

What is the effect of noise on the cost function?

$$
f(\boldsymbol{\theta}) = \sum_{\omega} d_{\omega} \Phi_{\omega}(\boldsymbol{\theta}) \xrightarrow{\text{Dephasing}} \tilde{f}(\boldsymbol{\theta}) = \sum_{\omega} (1 - p)^{|\omega|} d_{\omega} \Phi_{\omega}(\boldsymbol{\theta})
$$

\nWith noise, paths with large weights
\nare suppressed exponentially

[1] Cerf, N. J., and S. E. Koonin. "Monte Carlo simulation of quantum computation." Mathematics and Computers in Simulation 47.2-5 (1998): 143-152.

[2] Gao, Xun, and Luming Duan. "Efficient classical simulation of noisy quantum computation." 1810.03176

[3] Nemkov, Nikita, et al. "Fourier expansion in variational quantum algorithms." arXiv:2304.03787

[4] Begušić, Tomislav, Kasra Hejazi, and Garnet Kin Chan. "Simulating quantum circuit expectation values by Clifford perturbation theory." arXiv:2306.04797

Theoretical performance

- Truncation reduces the paths from exponential in the number of rotations (2^m) for m rotations) to constant (2^{ℓ} for cutoff ℓ).
- The overall runtime is now polynomial in the number of qubits and gates: $O(n^2m^2)$.
- Since discarded paths are the ones most affected by noise, the **average error** over landscape is bounded:

$$
\Delta\big(\tilde{f},\tilde{g}\big)\leq \big(1-2p-2p_Z\big)^{\ell+1}\leq e^{-2(p+p_Z)\ell}
$$

Given any nonzero, constant amount of noise, a VQA is approximately simulatable in polynomial time with guaranteed average error.

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Given any nonzero, constant amount of noise, a VQA is approximately simulatable in polynomial time with guaranteed average error.

> Within a small average error, there is **no exponential separation** between classical and quantum with noise

Experiments

Simulated large random noisy circuits (up to 50 qubits) in 2D topology, in under ~10s on laptop.

- With high ℓ , worst-case number of paths is huge.
- However, vast majority annihilate early and so simulation is still fast.
- Bounds on simulation error are loose by 1-3 orders of magnitude.

Caveats

In theory, there is a strong dependency of optimal cutoff on the noise rate: $\ell \approx p^{-1} \log \epsilon^{-1}$ With 1% phase noise, 1% error tolerance, worst-case runtime has a prefactor of $10^{100} = 1$ Google -> Large gap between empirical and worst-case performance.

Open questions

- Bound holds for **uncorrelated**, single qubit rotation gates only.
- Expectation values only.

- Can we improve the bounds?
- Can we extend to correlated parameter VQAs?
- Can we do sampling?

Outlook

- Presented LOWESA, a Fourier-based simulation algorithm that can handle noise.
- Under a reasonable noise model, uncorrelated parameter VQAs are classically simulatable in polynomial time.

Noise is a fundamental barrier to exponential quantum advantage for VQAs.

Is it a practical algorithm, though?

Outlook

- Presented LOWESA, a Fourier-based simulation algorithm that can handle noise.
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Noise is a fundamental barrier to exponential quantum advantage for VQAs.

Is it a practical algorithm, though?

- If hardware noise is large, LOWESA may be more accurate and efficient than a QC.
- There's indication that the bounds are potentially very loose for certain circuits.
- Appears that LOWESA has a role as a standalone (noiseless) simulation algorithm [1].

[1] Rudolph, Manuel S., et al. "Classical surrogate simulation of quantum systems with LOWESA." arXiv:2308.09109

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Bonus: Simulating 127 clean qubits

Transverse Field Ising Hamiltonian Trotter Time Evolution Circuit

$$
H = \sum_{\langle i,j \rangle} J^{(i,j)} Z_i Z_j + \sum_i h^{(i)} X_i
$$

$$
U(\pmb{\theta}) = \prod_{l=1}^L \left[\prod_{\langle i,j \rangle} \left(R_{ZZ}(\theta_J^{(i,j)}) \right) \prod_i \left(R_X(\theta_h^{(i)}) \right) \right]
$$

 $t = 0.00$

Rudolph, Manuel S., et al. "Classical surrogate simulation of quantum systems with LOWESA." arXiv:2308.09109

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