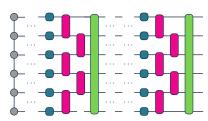
# How Can We Use Quantum Systems To Learn?

#### Matthew Duschenes

University of Waterloo, Institute for Quantum Computing, Perimeter Institute & Vector Institute

June 27, 2024

PSI Start Seminar











Engineering Physics





Engineering Physics Theoretical Physics







Engineering Physics

Theoretical Physics Applied Physics







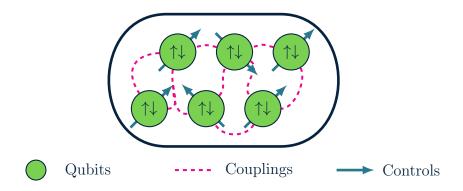


Engineering Physics

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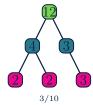
Quantum Physics

#### Quantum Systems

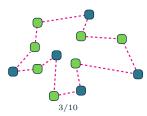


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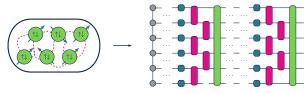
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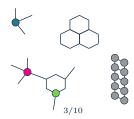
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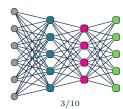
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- What makes quantum systems *potentially better* than classical systems?

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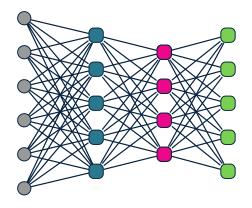
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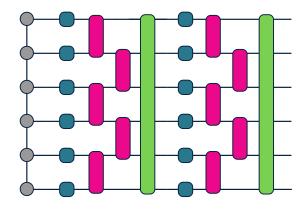
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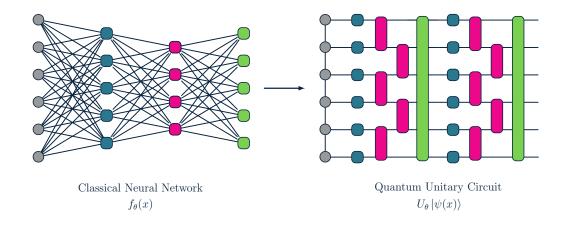
- Existing quantum devices are very noisy, error-prone, and difficult to scale
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- It remains up for debate on the quantum-classical complexity hierarchy

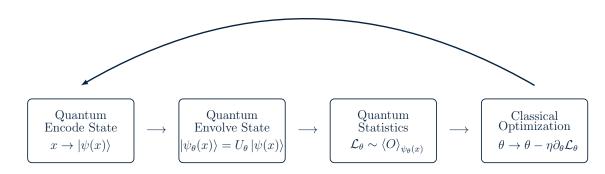


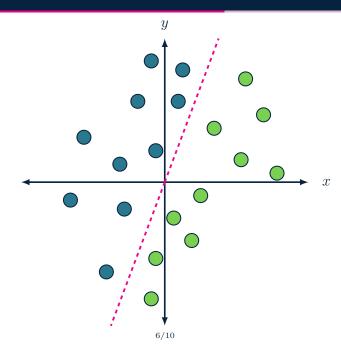
Classical Neural Network

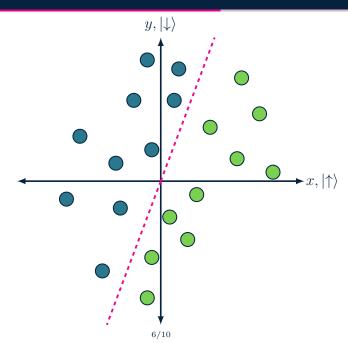


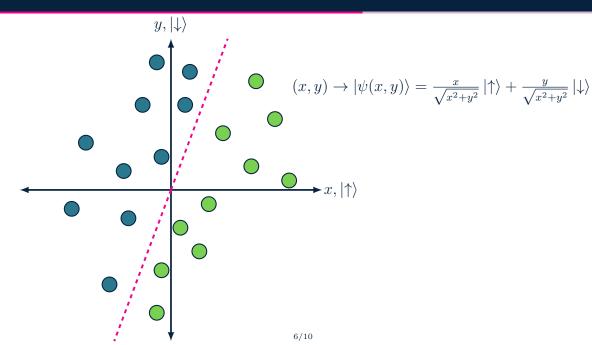
Quantum Unitary Circuit

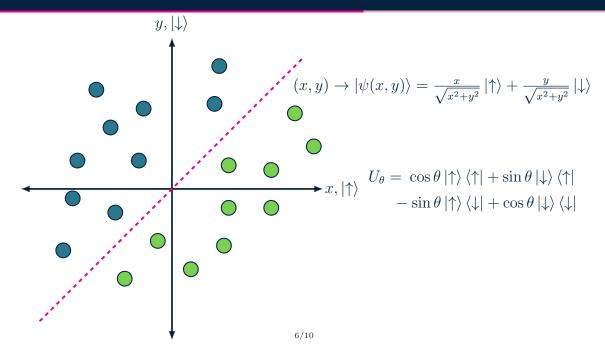


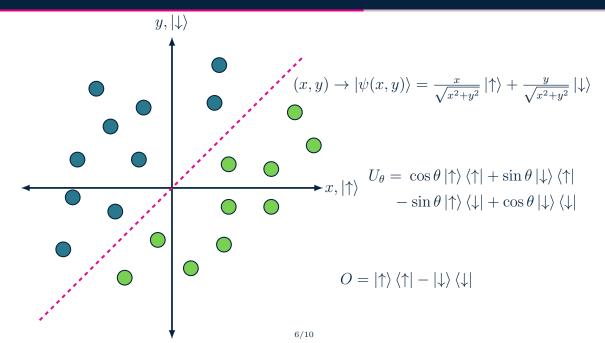






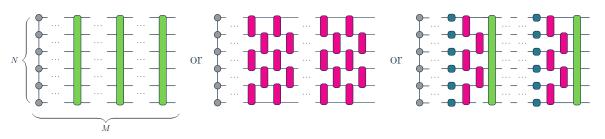






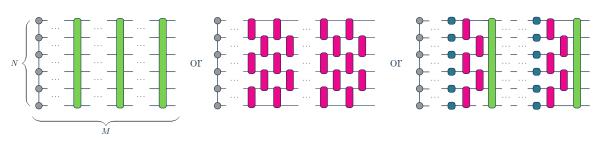
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• Basis: How do we choose our set of parameterized operators  $\{U_{\theta}^{(i)}\}_{i}^{M}$ ?



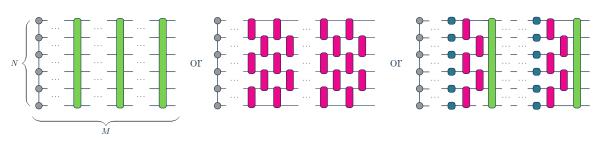
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- Depth: Many models are periodic layers, repeated M times
- Structure: How do we incorporate patterns in the data and objectives?



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$$U_{\theta} \approx U : \mathcal{L}_{\theta}^{U} \sim \text{Infidelity}(U, U_{\theta})$$
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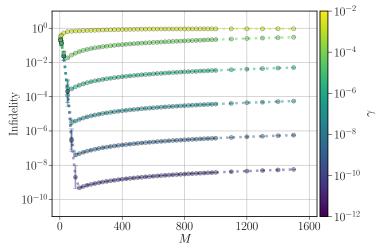
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- Develop quantum-inspired classical models

#### What About Noise?

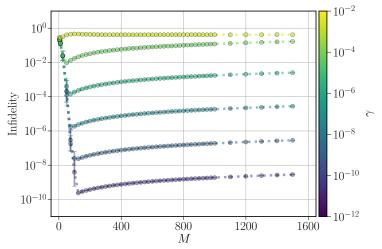
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**Figure 1:** Entropy-Increasing Noise: Unital Dephasing

#### What About Noise?

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**Figure 2:** Entropy-Decreasing Noise: Non-Unital Amplitude Damping

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- Useful Reviews:
  - 1. Cerezo, M. et al., Variational quantum algorithms. Nature Reviews Physics, 3(9), 625–644. (2021).
  - 2. Schuld, M.  $et\ al.$ , Is Quantum Advantage the Right Goal for Quantum Machine Learning? PRX Quantum, 3(3), 030101. (2022).
  - 3. Bharti, K.  $et\ al.$ , Noisy intermediate-scale quantum algorithms. Reviews of Modern Physics, 94(1), 015004. (2022).
  - 4. Duschenes, M. *et al.*, Characterization of overparametrization in the simulation of realistic quantum systems. Phys. Rev. A, 109, 062607. (2024).