

# Introduction to Quantum Machine Learning & Parameterized Quantum Circuit

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# Introduction to

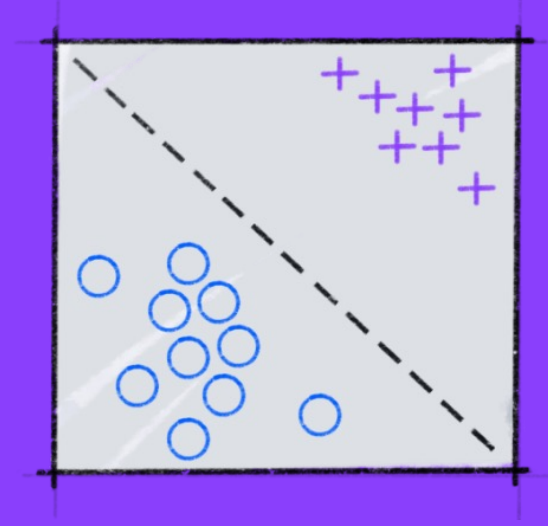


# Quantum Machine Learning

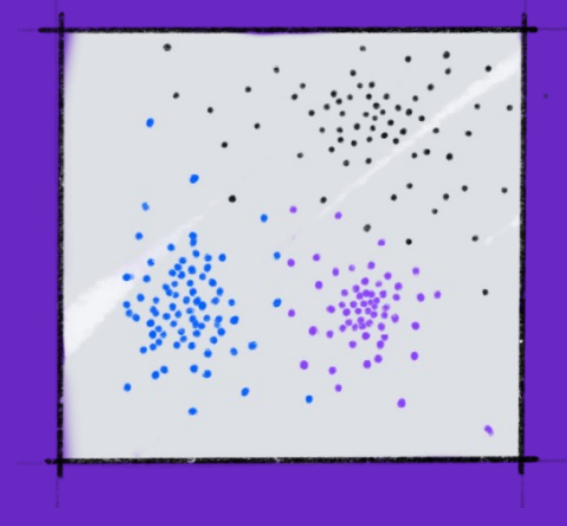
# Recap: Machine Learning

- Supervised learning
  - $\mathcal{D} = \{(x_i, y_i)\}_{i=1}^M, p(\hat{y}|\hat{x}, \mathcal{D})$
  - Data classification, regression, etc.
- Unsupervised learning
  - $\mathcal{D} = \{x_i\}_{i=1}^M, p(\hat{x}|\mathcal{D})$
  - Data clustering, generative models, etc.
- Reinforcement learning (RL)
  - Agent, Environment
  - Robotic control, game play, etc.

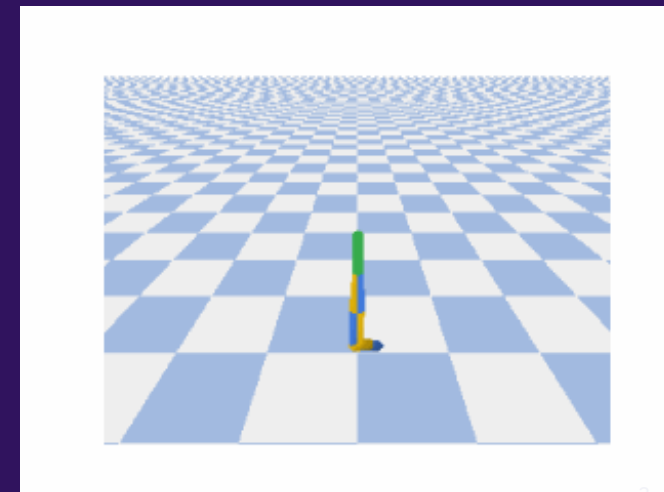
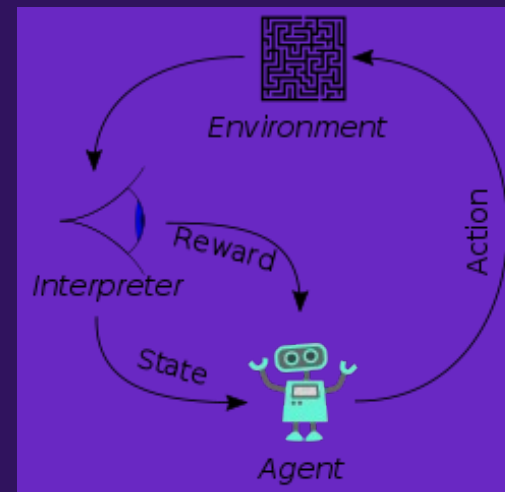
Supervised learning



Unsupervised learning

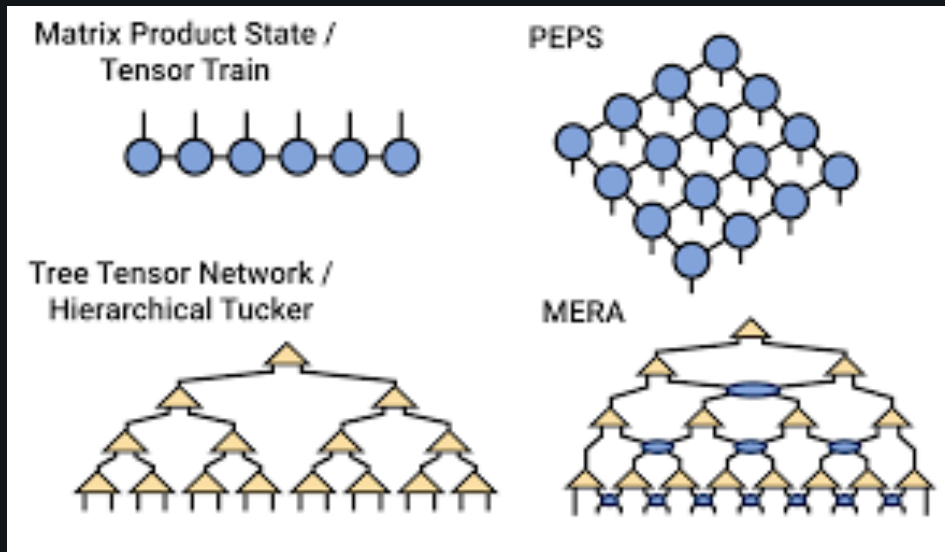


Reinforcement learning



# Categories of QML

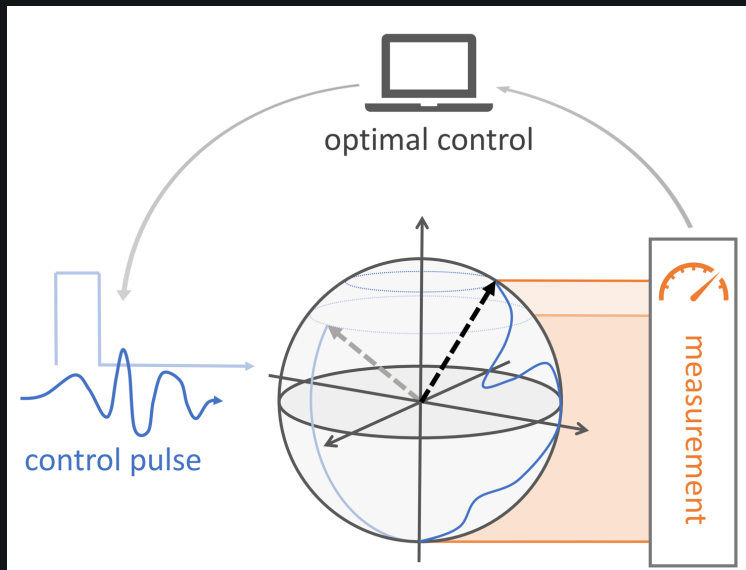
- Process classical data with classical computers
- “Inspired” by quantum systems
- Simulated Annealing, Quantum Monte Carlo, Tensor networks, etc.



Type of algorithm		Type of data
CC	CQ	
QC	QQ	

# Categories of QML

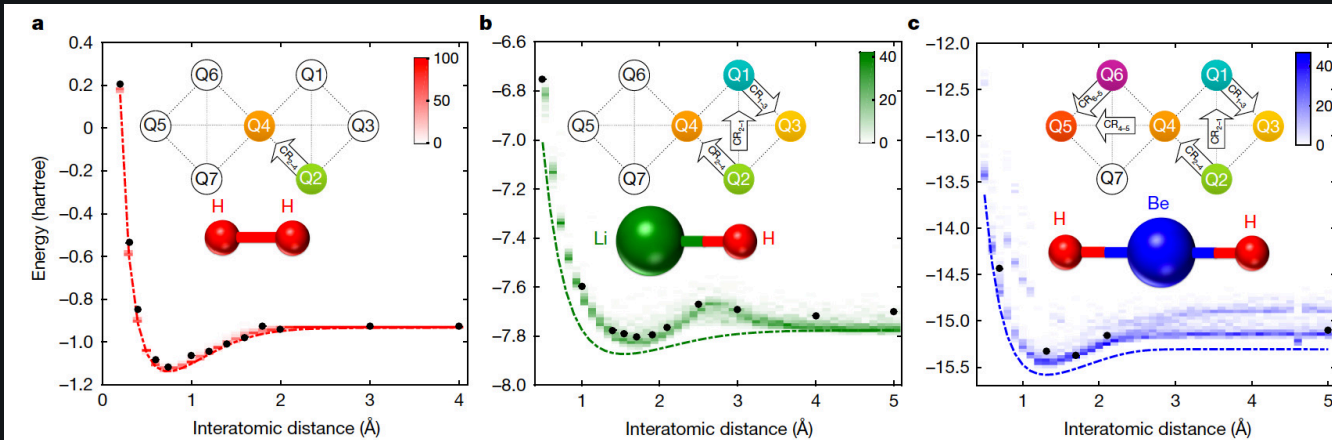
- Process quantum data with classical computers
- Inspect quantum systems with ML
- Qubit characterization, quantum optimal control, readout error mitigation, etc.



Type of algorithm		Type of data
CC	CQ	
QC	QQ	

# Categories of QML

- Process quantum data with quantum computers
- Measure quantum system  $\rightarrow$  quantum ML
- Or full quantum state as input
- Quantum many-body systems, Quantum Simulation, Quantum Chemistry



Type of algorithm	
Type of data	CC
	CQ
Type of data	QC
	QQ

# Categories of QML

- Process classical data with quantum computers
- Main focus!!
- 99% of all quantum-related ML (maybe?)

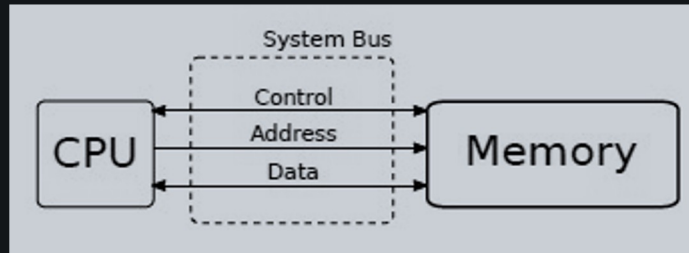
- A. With Power of **QRAM** (Skeptical)
- B. Without **QRAM** (Near-term)

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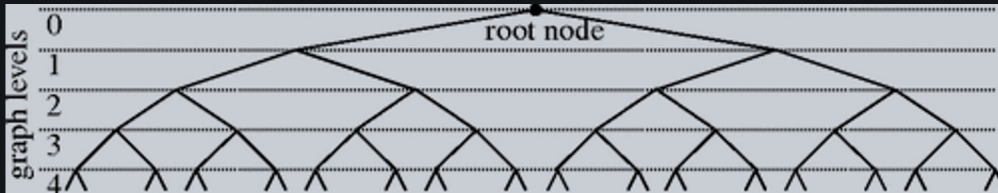
# Quantum RAM

## Random Access Memory

- Bus: [(ctrl)][address bits= $i$ ][data bits=0]
- $\rightarrow$  [(ctrl)][address bits= $i$ ][data bits= $data(i)$ ]



- Very fast  $\mathcal{O}(\log M)$ , access on at a time

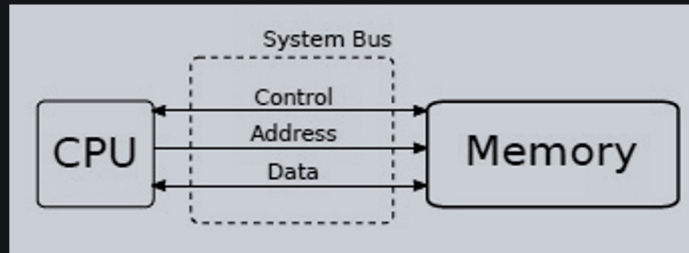




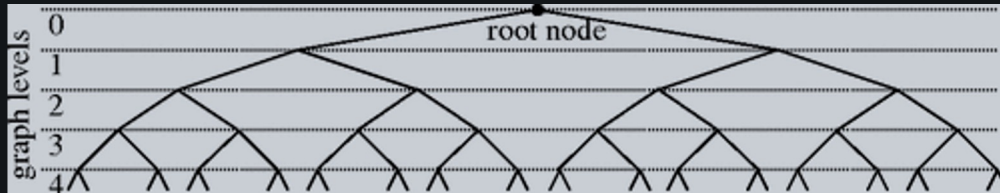
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## Quantum Random Access Memory

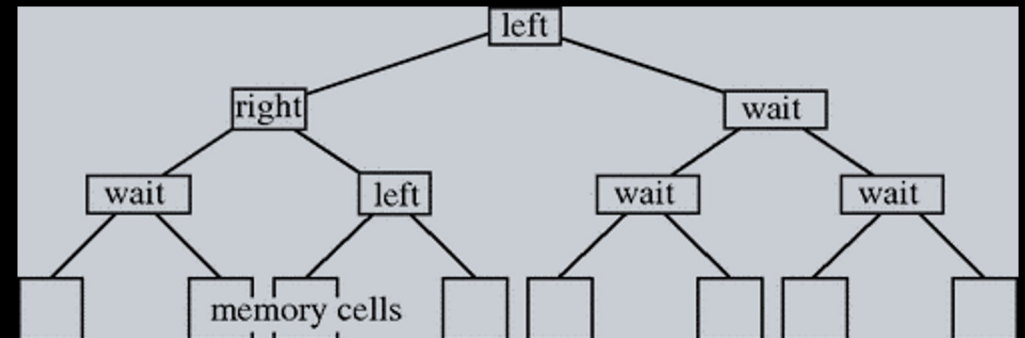
- Load classical data to quantum state:

$$QRAM|i\rangle|0\rangle = |i\rangle|x_i\rangle$$

- Also, load **all data simultaneously**;

$$QRAM\left(\frac{1}{\sqrt{M}}\sum_{i=1}^M|i\rangle\right)|0\rangle = \frac{1}{\sqrt{M}}\sum_{i=1}^M|i\rangle|x_i\rangle$$

- Very fast  $\mathcal{O}(\log M) \rightarrow$  **Exponential Speed-up**



$$|\psi_{QRAM}\rangle = \frac{1}{\sqrt{M}} \sum_{i=1}^M |i\rangle |x_i\rangle$$

Index register  $|i\rangle$ :

$\log M$  qubits,

Data register

$$|x_i\rangle = \frac{1}{\sqrt{\|x_i\|}} \sum_{j=1}^N x_j^{(i)} |j\rangle:$$

$\log N$  qubits

System size

$$\mathcal{O}(MN) \rightarrow \mathcal{O}(\log MN)$$

Really?

Access time

$$\mathcal{O}(MN) \rightarrow \mathcal{O}(\log MN)$$

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$$\begin{aligned} \rho_{QRAM} &= \frac{1}{M} \sum_{i,j=1}^M |i\rangle\langle j| \otimes |x_i\rangle\langle x_j| \end{aligned}$$

Let's trace out index register

$$\rho_{data} = \frac{1}{M} \sum_{i=1}^M |x_i\rangle\langle x_i|$$

Let's trace out data register

$$\begin{aligned} \rho_{index} &= \frac{1}{M} \sum_{i,j=1}^M \langle x_j | x_i \rangle |i\rangle\langle j| \\ &= \frac{K}{\text{tr}(K)} \end{aligned}$$

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Playing with Density matrix

- Matrix Exponentiation  
 $\rho \rightarrow e^{-i\rho\Delta t}$
- Matrix Inversion  
 $|x\rangle = \rho^{-1}|b\rangle$

This techniques achieves exponential quantum speed-up  
 $\mathcal{O}(M^2N) \rightarrow \mathcal{O}(\log MN)$

&

Very useful

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Applications

- qPCA  
 $\mathcal{O}(\min\{M^2N, N^2M\})$   
 $\rightarrow \mathcal{O}(\epsilon^{-2}R \log \min\{M, N\})$
- qSVM  
 $\mathcal{O}(M^2N)$   
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- qClustering  
 $\mathcal{O}(M^2N)$   
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# Issue with QRAM-based algorithms

- Complex quantum subroutines → Requires Fault-tolerant QC
- Low-rank Approximation (Read the fine print!!)
- Not good for non-linear algebra
- No viable hardware candidate for realizing qRAM (not that I know of...)
- QML without efficient data loading assumption!!



# Introduction to



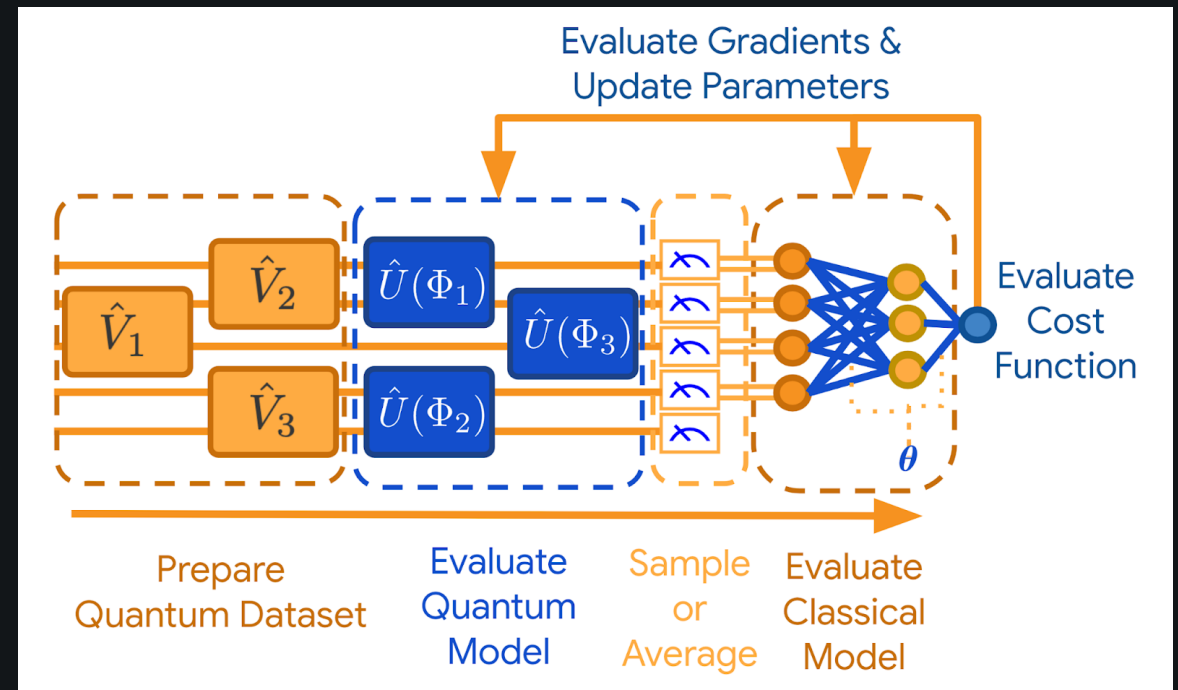
# Parameterized Quantum Circuit



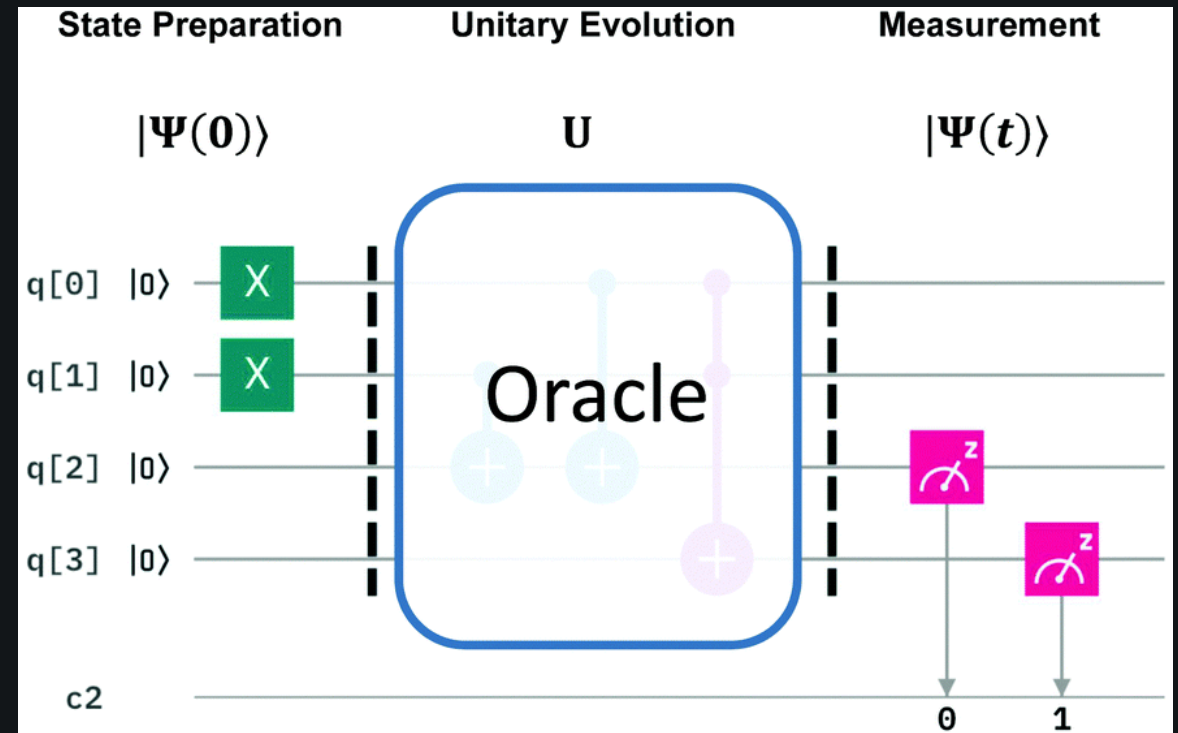
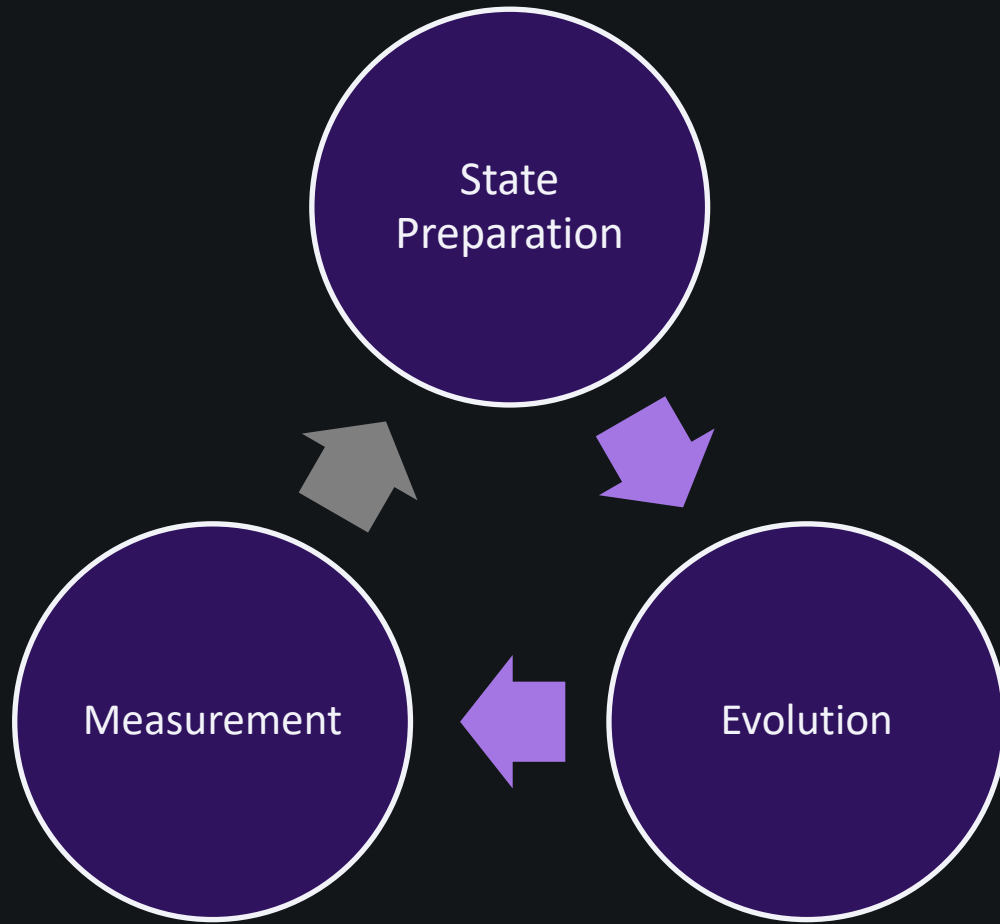
# Hybrid Quantum-Classical Algorithm



- Do not assume loading data is not trivial
- Do assume NISQ
- Q: How to perform QML in near-future?
  - How to encode classical data to QC?
  - How to tune circuit?
  - How to tune measurement basis
- A: Parameterize Quantum Circuit + Variational Quantum Eigensolver/Algorithm
  - $E_0 = \min_{\psi} \langle \psi | H | \psi \rangle$



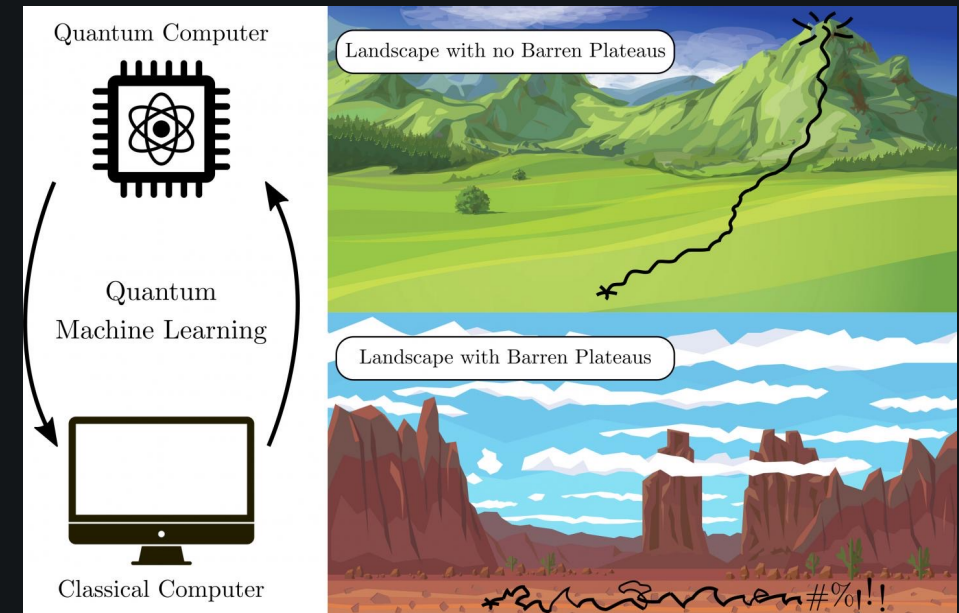
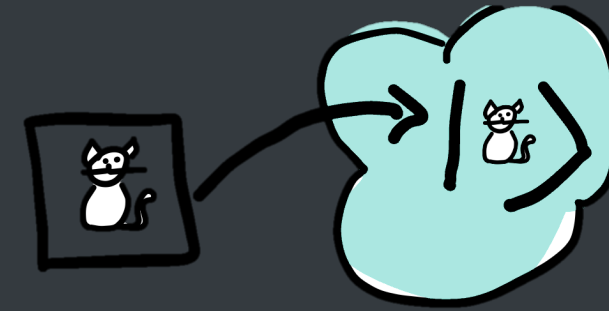
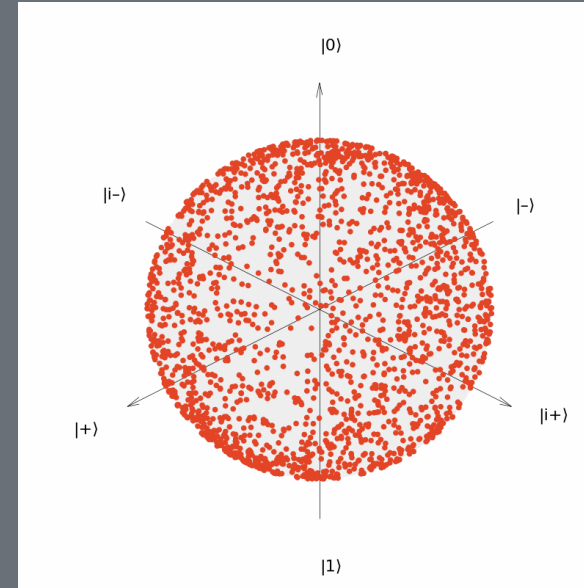
# Principle of QC



# Learning With Qiskit

# Next Lecture

- PQC Properties
  - Expressibility
  - Entanglement Capacity
  - Hardware Efficiency
- Data encoding
  - Amplitude encoding, etc.
- Trainability
  - Gradients of PQC
  - Barren plateaus



# Thank you

Park, Siheon

Qiskit Machine Learning <https://learn.qiskit.org/course/machine-learning/introduction>

Lloyd, Seth, Masoud Mohseni, and Patrick Rebentrost. "Quantum principal component analysis." *Nature Physics* 10.9 (2014): 631-633. <https://www.nature.com/articles/nphys3029>

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