



# Quantum Machine Learning

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# Overview



| Artificial Neural Networks for Deep Learning

| Introduction to Quantum Computing

| Quantum Programming

| **Quantum Machine Learning**

# Quantum Machine Learning (QML)



**|Goals are identical to classical machine learning.**

**|For supervised learning:**

- Input data.
- Define model.
- Learn parameters to best fit data.
- Predict future (unknown) inputs.

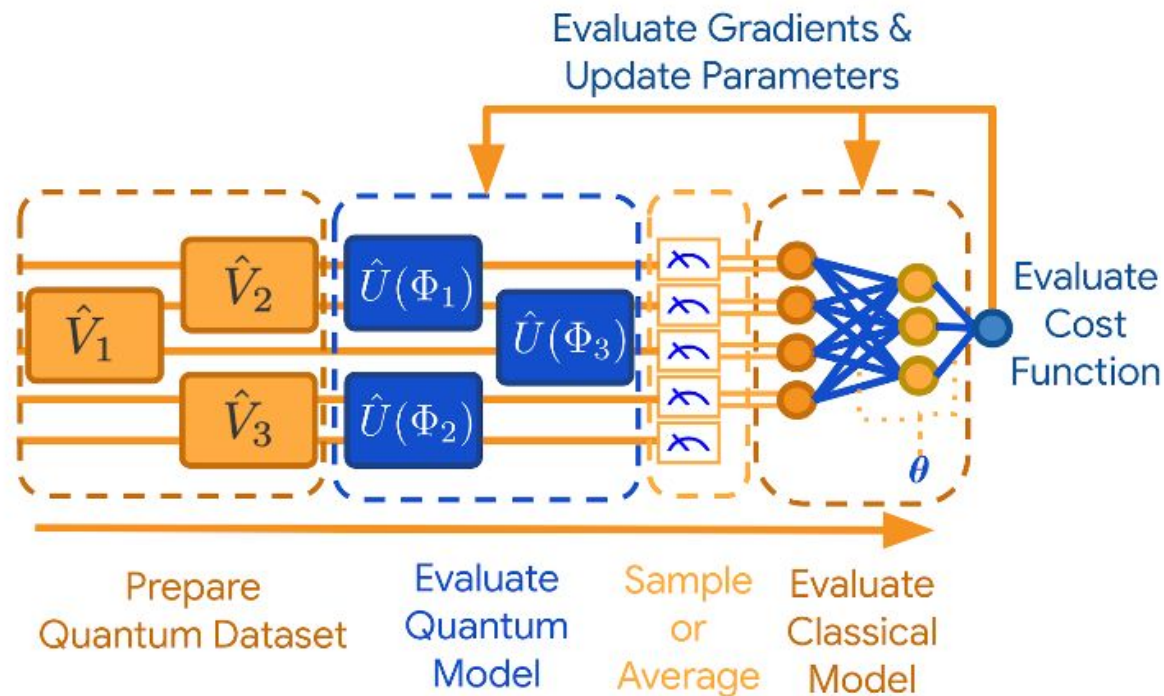
**|For unsupervised learning:**

- Categorize data using some distance measure.
- Predict category of future data.

# Quantum Machine Learning (QML) (cont'd)

## Key Components:

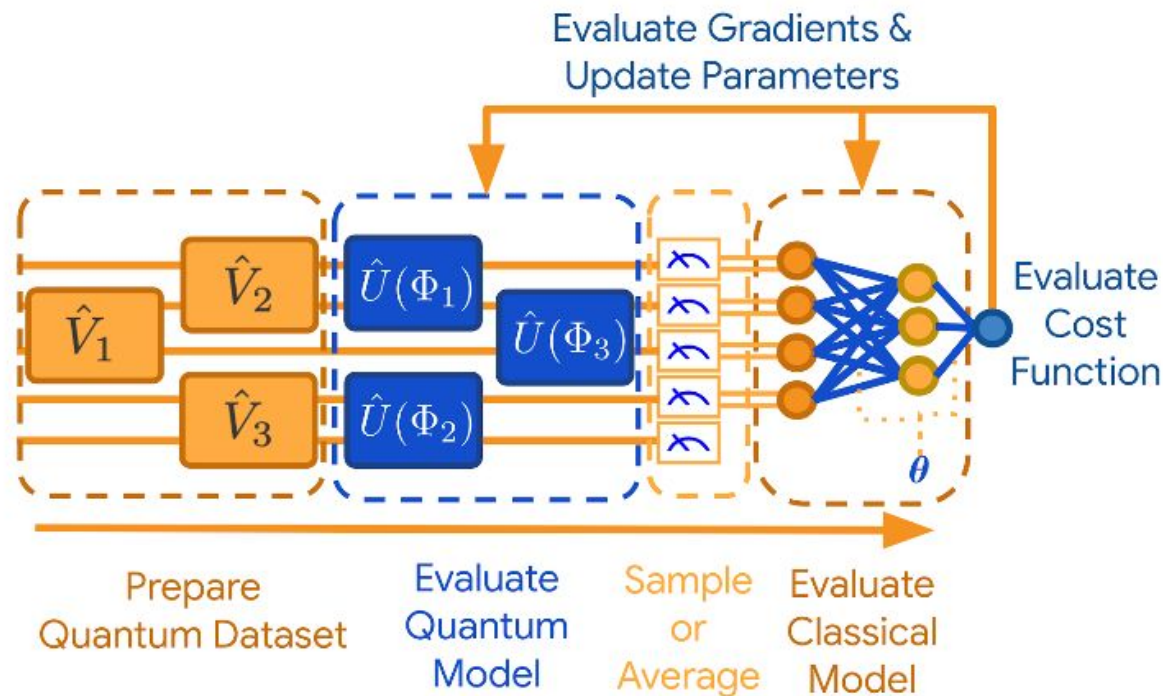
- Data embedding (store classical data in quantum states)
- Ansatz definition (similar to structure of neural network, including number and type of layers)
- Measurement (quasiprobability distribution or expectation values)



# Quantum Machine Learning (QML) (cont'd)

## Key Components:

- Classical postprocessing (optional step, can use classical neural network)
- Optimization (cost function evaluation, gradient evaluation, and parameter updates)



# Applications of QML



| Supervised and unsupervised learning (similar to standard neural network)

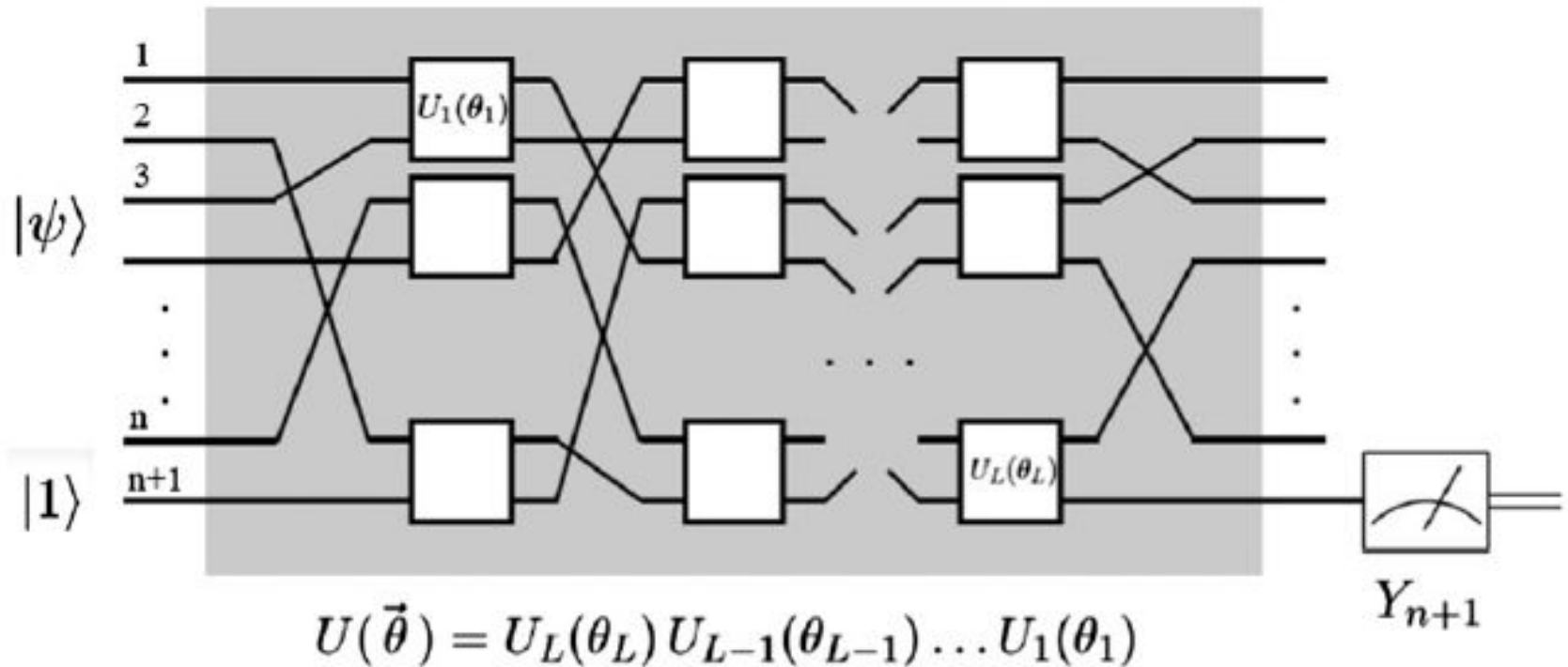
| Natural Language Processing (similar to RNN)

| Image recognition (similar to CNN)

| Classical or quantum data processing

- Can process quantum data directly without embedding.

# Quantum Neural Network Example



# Data Embedding



**| Three straightforward strategies can be used to perform data embedding, though many advanced strategies also exist.**

## **| Basis Encoding:**

- Treat qubits as binary bits.
- Requires  $N * \tau$  qubits, where  $\tau$  is the classical data's precision (e.g., 32-bit).
- Embedding time is  $O(N * \tau)$ .



# Data Embedding (cont'd)



## |Amplitude Encoding:

- Directly embed features as the normalized amplitudes of qubits.
- Requires  $\log(N)$  qubits.
- Embedding time is  $O(2^n/n)$ , with  $n$  = number of qubits.

## |Rotation Embedding:

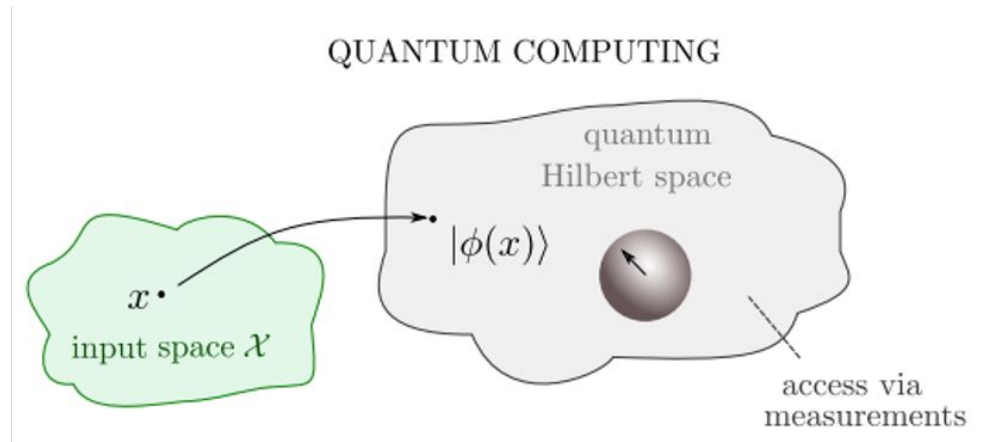
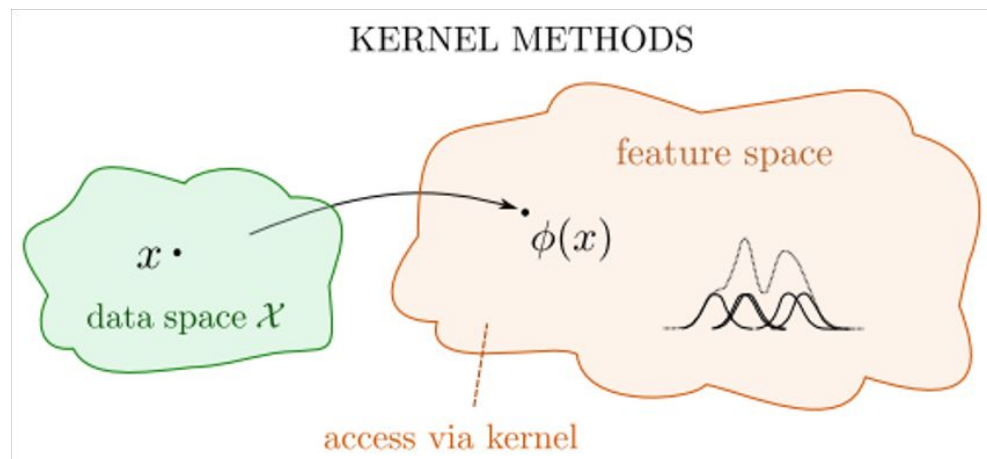
- Use qubit's spherical representation to embed data using rotations. (This approach is very easy for programmers.)
- Requires  $N$  qubits.
- Embedding time is  $O(N)$ .

# Data Embedding (cont'd)

| Data embedding can be considered as a type of kernel method.

| The formula for the kernel can be directly derived from the embedding strategy.

| Changing embedding strategies can increase the model's expressiveness.



# Ansatz Definition



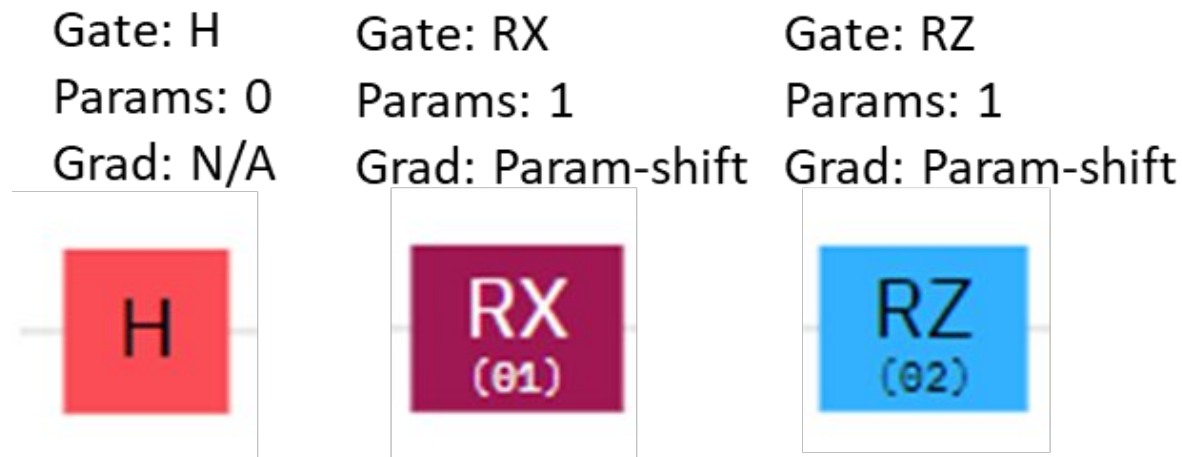
|The ansatz definition is open to significant experimentation.

|Many "standard" ansatzes are available for testing.

|The ansatz should typically incorporate entanglement (e.g., CNOT) and parameterized gates (e.g., rotation gates).

# Measurement and Optimization

- | QML faces the same issues as classical ML, such as the barren plateau problem.
- | QML optimization runs on a classical computer using classical optimization methods (e.g., AdamOptimizer).
- | The method used for gradient computation can be experimented with, though most platforms can do it automatically (e.g., parameter-shift rules).

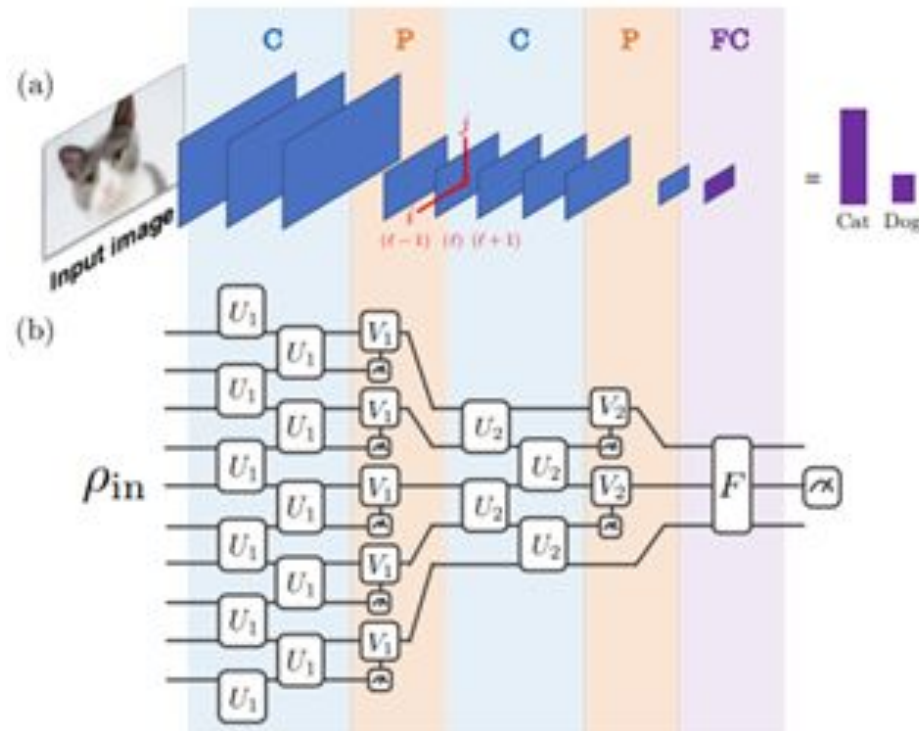


# QML Application Example

It is possible to develop quantum versions of existing classical methodologies (e.g., a quantum CNN – QCNN for image processing).

Classical  
CNN

QCNN



# Summary



## | Artificial Neural Networks for Deep Learning

- Neurons → Qubits

## | Introduction to Quantum Computing

- Bits vs. Qubits

## | Quantum Programming

- Various Quantum programming libraries and environments

## | Quantum Machine Learning

- CNN vs. QCNN