

A snapshot of Quantum Computing

from quantum circuits to machine learning

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Introduction

Introduction

From a practical point of view, we are moving towards new technologies, in particular **hardware accelerators**:

CPU



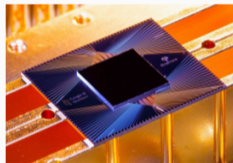
GPU



FPGA/ASIC



Quantum chip

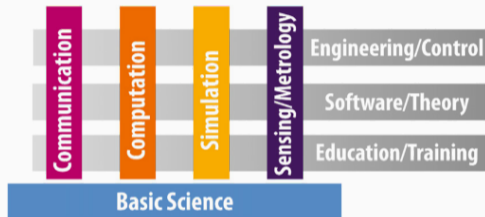


Moving from **general purpose devices** \Rightarrow **application specific**

For example, in HEP we are transitioning from **CPU** to **GPU**.

Quantum research

Structure of research field in **quantum technologies**:

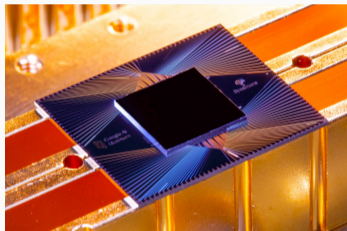
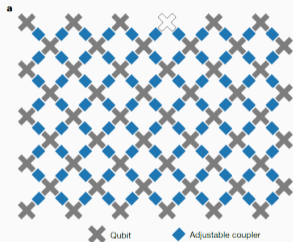


Quantum computing is a paradigm that exploits quantum mechanical properties of matter in order to perform calculations.

⇒ Unitary operators, entanglement, superposition, interference, etc.

Quantum advantage

First quantum computation that can not be reproduced on a classical supercomputer from Google, [Nature 574, 505-510\(2019\)](#):



53 qubits (86 qubit-couplers) → Task of sampling the output of a pseudo-random quantum circuit (extract probability distribution).

Classically the probability distribution is **exponentially more difficult**.

Quantum landscape

Software & Consultants



Quantum Computers



Enabling Technologies



New Funding Strategies



Representative list of players. A very active ecosystem!

Qubits

What is a qubit?

Let us consider a two-dimensional **Hilbert space**, we define the computational basis:

$$|0\rangle \rightarrow \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \quad |1\rangle \rightarrow \begin{pmatrix} 0 \\ 1 \end{pmatrix}.$$

A **quantum bit (qubit)** is the basic unit of quantum information and it written as:

$$|\psi\rangle = \alpha |0\rangle + \beta |1\rangle \rightarrow \begin{pmatrix} \alpha \\ \beta \end{pmatrix},$$

where $\alpha, \beta \in \mathbb{C}$ and the **state is normalized**, i.e. $|\alpha|^2 + |\beta|^2 = 1$.

All quantum mechanics rules are preserved: state measurement is probabilistic, wave-function collapse after measurement, no-cloning theorem, etc.

The Bloch sphere

Qubit states can be graphically represented in the [Bloch sphere](#), by defining ϕ and θ angles and associating to the state coefficients:

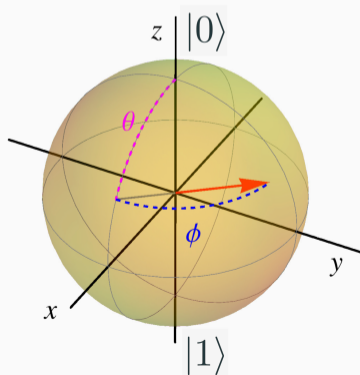
$$\alpha = \cos \frac{\theta}{2}, \quad \text{and} \quad \beta = e^{i\phi} \sin \frac{\theta}{2}, \quad \text{with} \quad \theta \in [0, \pi], \phi \in [0, 2\pi].$$

We can use a 3D vector representation as:

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} \sin \theta \cos \phi \\ \sin \theta \sin \phi \\ \cos \theta \end{pmatrix}$$

In particular:

- $|0\rangle = (0, 0, 1)$, $|1\rangle = (0, 0, -1)$
- $(|0\rangle + i|1\rangle)/\sqrt{2}$ equator of the sphere



Two qubits states

A system with 2 qubits is represented by the basis:

$$|0\rangle \otimes |0\rangle \equiv |00\rangle, \quad |0\rangle \otimes |1\rangle \equiv |01\rangle$$

$$|1\rangle \otimes |0\rangle \equiv |10\rangle, \quad |1\rangle \otimes |1\rangle \equiv |11\rangle$$

which lives in 2^2 -dimensional Hilbert space:

$$|00\rangle = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \quad |01\rangle = \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \quad |10\rangle = \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \quad |11\rangle = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

Therefore a generic 2 qubits state is defined as:

$$|\psi_2\rangle = \alpha_{00} |00\rangle + \alpha_{01} |01\rangle + \alpha_{10} |10\rangle + \alpha_{11} |11\rangle \quad \text{with} \quad \sum_{i,j=0}^1 |\alpha_{ij}|^2 = 1$$

Multiple qubits states

A system with n qubits lives in 2^n -dimensional Hilbert space, defining the basis:

$$|0\rangle_n = |00 \dots 00\rangle, |1\rangle_n = |00 \dots 01\rangle, |2\rangle_n = |00 \dots 10\rangle, \dots, |2^n - 1\rangle_n = |11 \dots 1\rangle$$

therefore a generic n qubits state is defined as

$$|\psi_n\rangle = \sum_{i=0}^{2^n-1} \alpha_i |i\rangle_n \quad \text{with} \quad \sum_{i=0}^{2^n-1} |\alpha_i|^2 = 1$$

i.e. a superposition state vector in 2^n dimensional Hilbert space.

Quantum operators

As any other quantum state defined in Hilbert space, qubits are subject to:

- **time evolution** via Schrödinger equation: $H(t) |\psi(t)\rangle = i\hbar\partial_t |\psi(t)\rangle$
- **quantum operators/gates**, in particular unitary operators (reversible computing):

$$UU^\dagger = U^\dagger U = I$$

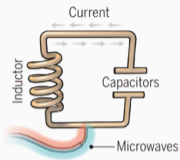
- **entanglement state**, e.g. supposing $|\psi_A\rangle |\phi_B\rangle$, e.g. Bell's states:

$$|\psi^+\rangle = \frac{|0_A\rangle |0_B\rangle + |1_A\rangle |1_B\rangle}{\sqrt{2}}, \quad |\psi^-\rangle = \frac{|0_A\rangle |0_B\rangle - |1_A\rangle |1_B\rangle}{\sqrt{2}}$$

$$|\phi^+\rangle = \frac{|1_A\rangle |0_B\rangle + |0_A\rangle |1_B\rangle}{\sqrt{2}}, \quad |\phi^-\rangle = \frac{|1_A\rangle |0_B\rangle - |0_A\rangle |1_B\rangle}{\sqrt{2}}$$

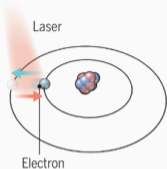
Quantum technologies

Some popular quantum technologies available today



Superconducting loops

A resistance-free current oscillates back and forth around a circuit loop. An injected microwave signal excites the current into superposition states.



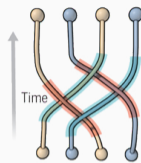
Trapped ions

Electrically charged atoms, or ions, have quantum energies that depend on the location of electrons. Tuned lasers cool and trap the ions, and put them in superposition states.



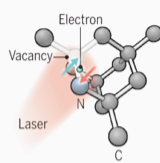
Silicon quantum dots

These "artificial atoms" are made by adding an electron to a small piece of pure silicon. Microwaves control the electron's quantum state.



Topological qubits

Quasiparticles can be seen in the behavior of electrons channeled through semiconductor structures. Their braided paths can encode quantum information.



Diamond vacancies

A nitrogen atom and a vacancy add an electron to a diamond lattice. Its quantum spin state, along with those of nearby carbon nuclei, can be controlled with light.

Number entangled

9

14

2

N/A

6

Company support

Google, IBM, Quantum Circuits

ionQ

Intel

Microsoft, Bell Labs

Quantum Diamond Technologies

+ Pros

Fast working. Build on existing semiconductor industry.

Very stable. Highest achieved gate fidelities.

Stable. Build on existing semiconductor industry.

Greatly reduce errors.

Can operate at room temperature.

- Cons

Collapse easily and must be kept cold.

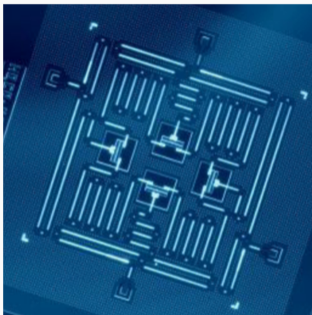
Slow operation. Many lasers are needed.

Only a few entangled. Must be kept cold.

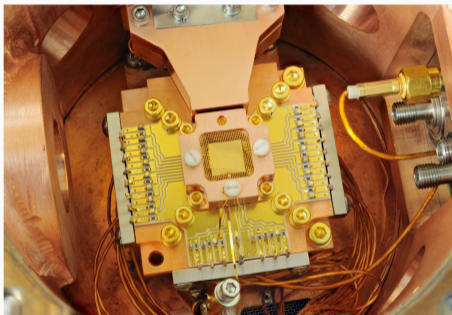
Existence not yet confirmed.

Difficult to entangle.

Quantum chips



(a) Superconducting device assembled by IBM



(b) Chip based on trapped ions technology

Superconducting labs



The current Quantum era

⇒ **We are in a Noisy Intermediate-Scale Quantum era** ⇐
(i.e. hardware with few noisy qubits)

How can we contribute?

- Develop **new algorithms**
 - ⇒ using classical simulation of quantum algorithms
- Adapt problems and strategies for **current hardware**
 - ⇒ hybrid classical-quantum computation

Quantum Algorithms

There are three families of algorithms:

Gate Circuits

- Search (Grover)
- QFT (Shor)
- Deutsch
- ...

Variational (AI inspired)

- Eigensolvers
- Autoencoders
- Classifiers
- ...

Annealing

- Direct Annealing
- Adiabatic Evolution
- QAOA
- ...

However, there are several challenges:

- simulate efficiently algorithms on classical hardware for QPU?
- control, send and retrieve results from the QPU?
- error mitigation, keep noise and decoherence under control?



Quantum computing with qubits

Quantum circuits

The **quantum circuit** model considers a sequence of unitary quantum gates:

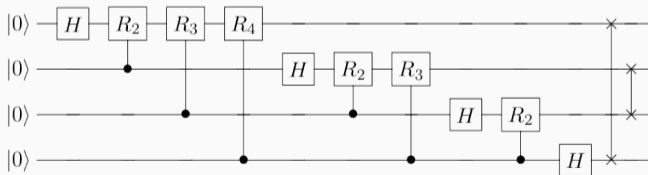
$$|\psi'\rangle = U_2 U_1 |\psi\rangle \quad \rightarrow \quad |\psi\rangle \text{ --- } \boxed{U_1} \text{ --- } \boxed{U_2} \text{ --- } |\psi'\rangle$$

Quantum circuits

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For example a Quantum Fourier Transform with 4 qubits is represented by



Quantum gates














- **Single-qubit gates**

- Pauli gates
- Hadamard gate
- Phase shift gate
- Rotation gates

- **Two-qubit gates**

- Conditional gates
- Swap gate
- fSim gate

- Special gates: Toffoli

Operator	Gate(s)	Matrix
Pauli-X (X)	 	$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$
Pauli-Y (Y)		$\begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix}$
Pauli-Z (Z)		$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$
Hadamard (H)		$\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$
Phase (S, P)		$\begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix}$
$\pi/8$ (T)		$\begin{bmatrix} 1 & 0 \\ 0 & e^{i\pi/4} \end{bmatrix}$
Controlled Not (CNOT, CX)		$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix}$
Controlled Z (CZ)	 	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}$
SWAP	 	$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$
Toffoli (CCNOT, CCX, TOFF)		$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$

Quantum circuit simulation

Classical simulation of quantum circuits uses dense complex state vectors $\psi(\sigma_1, \sigma_2, \dots, \sigma_N) \in \mathbb{C}$ in the computational basis where $\sigma_i \in \{0, 1\}$ and N is the total number of qubits in the circuit.

The final state of circuit evaluation is given by:

$$\psi'(\sigma) = \sum_{\sigma'} G(\sigma, \sigma') \psi(\sigma_1, \dots, \sigma'_{i_1}, \dots, \sigma'_{i_{N_{\text{targets}}}}, \dots, \sigma_N),$$

where the sum runs over qubits targeted by the gate.

- $G(\sigma, \sigma')$ is a gate matrix which acts on the state vector.
- $\psi(\sigma)$ from a simulation point of view is bounded by memory.

Pauli gates

X gate

The X gate acts like the classical NOT gate, it is represented by the σ_x matrix,

$$\sigma_x = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

therefore

$$\begin{array}{l} |0\rangle \text{ --- } \boxed{X} \text{ --- } |1\rangle \\ |1\rangle \text{ --- } \boxed{X} \text{ --- } |0\rangle \end{array}$$

Z gate

The Z gate flips the sign of $|1\rangle$, it is represented by the σ_z matrix,

$$\sigma_z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

therefore

$$\begin{array}{l} |0\rangle \text{ --- } \boxed{Z} \text{ --- } |0\rangle \\ |1\rangle \text{ --- } \boxed{Z} \text{ --- } -|1\rangle \end{array}$$

Hadamard gate

The Hadamard gate (H gate) is defined as

$$H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

Therefore it creates a superposition of states

$$|0\rangle \text{ --- } \boxed{H} \text{ --- } \frac{|0\rangle + |1\rangle}{\sqrt{2}} \equiv |+\rangle$$

$$|1\rangle \text{ --- } \boxed{H} \text{ --- } \frac{|0\rangle - |1\rangle}{\sqrt{2}} \equiv |-\rangle$$

The rotation gates

Rotations gates (Bloch sphere) are defined as

$$R_X(\theta) = e^{-i\frac{\theta}{2}\sigma_x} = \begin{pmatrix} \cos \frac{\theta}{2} & -i \sin \frac{\theta}{2} \\ -i \sin \frac{\theta}{2} & \cos \frac{\theta}{2} \end{pmatrix}, \quad R_Y(\theta) = e^{-i\frac{\theta}{2}\sigma_y} = \begin{pmatrix} \cos \frac{\theta}{2} & -\sin \frac{\theta}{2} \\ \sin \frac{\theta}{2} & \cos \frac{\theta}{2} \end{pmatrix}$$

$$R_Z(\theta) = e^{-i\frac{\theta}{2}\sigma_z} = \begin{pmatrix} 1 & 0 \\ 0 & e^{i\theta} \end{pmatrix}$$

Note that $R_X(\pi) \equiv X$, $R_Y(\pi) \equiv Y$, $R_Z(\pi) \equiv Z$.

The rotation gates

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Note that $R_X(\pi) \equiv X$, $R_Y(\pi) \equiv Y$, $R_Z(\pi) \equiv Z$.

Every unitary transformation as decomposed in rotations around the y and z axis:

$$U \equiv R_Z(\theta_1)R_Y(\theta_2)R_Z(\theta_3),$$

for a fixed set of angles θ_1 , θ_2 and θ_3 .

Two-qubit gates

The controlled-NOT (CNOT) gate is a conditional gate defined as

$$\text{CNOT} = \begin{pmatrix} 1 & 0 \\ 0 & \sigma_X \end{pmatrix}$$

We define a control qubit which if at $|1\rangle$ applies X to a target qubit.

Supposing the **first qubit** is the **control** and the **second qubit** the **target**:

$$\begin{aligned} |00\rangle &\rightarrow |00\rangle & |01\rangle &\rightarrow |01\rangle \\ |10\rangle &\rightarrow |11\rangle & |11\rangle &\rightarrow |10\rangle \end{aligned}$$

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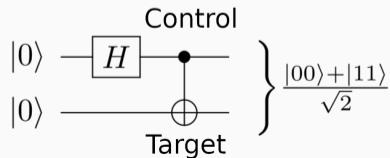
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CNOT allows entangled states, e.g.:



Measurements

So far we have simulated quantum circuits using wave-function propagation.

In real experiments we perform measurements with a preselected number of **shots**.

Shots contribute to the reconstruction of the underlying wave-function distribution.

Measurement (M) gate:

Lets consider the following circuit:



The analytic final state is:

$$\frac{|0\rangle + |1\rangle}{\sqrt{2}}$$

When measuring the final state we obtain 0 or 1 each with 50% probability.

Hands on tutorial

We will use Qibo for a practical demonstration:

The screenshot shows the Qibo website's home page. At the top, there is a search bar and navigation links for 'USER DOCUMENTATION', 'Installing Qibo', 'Basic examples', and 'Advanced examples'. Below this, there is a section for 'Application tutorials' with a list of topics: 'Scaling of variational quantum circuit depth for condensed matter systems', 'Grover's Algorithm for solving Satisfiability Problems', 'Variational Quantum Classifier', 'Data respawning for a universal quantum classifier', 'Quantum autoencoder for data compression', 'Quantum Singular Value Decomposer', 'Measuring the tangle of three-qubit states', and 'Quantum unary approach to option pricing'. A 'Next' button is visible at the bottom right of the list.

The screenshot shows the GitHub repository page for Qibo. The page title is 'Application tutorials' and it includes a list of 12 tutorials. The first tutorial is 'Scaling of variational quantum circuit depth for condensed matter systems'. The page also features a 'Previous' button and a copyright notice: '© Copyright 2020 by Quantum-TIL team Revision 6c9d1aa1. Built with Sphinx using a theme provided by Read the Docs.'

The screenshot shows the GitHub repository page for Qibo, specifically the README file. The README includes the Qibo logo, a description of Qibo as an open-source full stack API for quantum simulation and quantum hardware control, and a list of key features: 'Definition of a standard language for the construction and execution of quantum circuits with device agnostic approach to simulation and quantum hardware control based on plug and play backend drivers', 'A continuously growing code base of quantum algorithm applications, presented with examples and tutorials', 'Efficient simulation backends with GPU, multi-GPU and CPU with multi-threading support', and 'Simple mechanism for the implementation of new simulation and hardware backend drivers'.

Documentation: <https://qibo.readthedocs.io>

GitHub: <https://github.com/qiboteam/qibo>

Visit the tutorial:

<https://colab.research.google.com/drive/1M4HV1RroiHtxh4uZdrSGASv51Tjpe6dT?usp=sharing>

Variational Quantum Circuits

Variational Quantum Circuits

Getting inspiration from **AI**:

- **Supervised** Learning \Rightarrow Regression and classification
- **Unsupervised** Learning \Rightarrow Generative models, autoencoders
- **Reinforcement** Learning \Rightarrow Quantum RL / Q-learning

Variational Quantum Circuits

Getting inspiration from **AI**:

- **Supervised** Learning \Rightarrow Regression and classification
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- **Reinforcement** Learning \Rightarrow Quantum RL / Q-learning

Define new parametric model architectures for quantum hardware:

\Rightarrow **Variational Quantum Circuits / Quantum Machine Learning**

Why Quantum Machine Learning?

Why QML?

- ① Proof-of-concept, study new architectures.
- ② Obtain a hardware representation (analogy with GPU and FPGA).
- ③ Lower power consumption.

Why Quantum Machine Learning?

Why QML?

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NISQ era Warning...

- Quantum devices implement few qubits, noise is a bottleneck.
- We can simulate quantum computation on classical hardware.

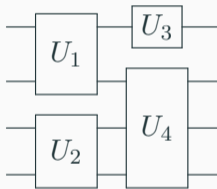
Rational

Rational for Variational Quantum Circuits

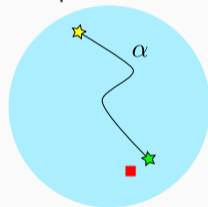
Rational:

Deliver variational quantum states \rightarrow explore a large Hilbert space.

$$U(\vec{\alpha}) = U_n \dots U_2 U_1$$



Near optimal solution

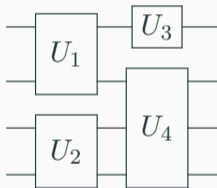


Rational for Variational Quantum Circuits

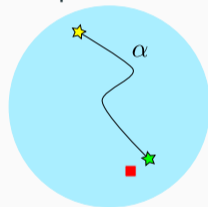
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Near optimal solution



Idea:

Quantum Computer is a machine that generates variational states.

\Rightarrow **Variational Quantum Computer!**

Solovay-Kitaev Theorem

Let $\{U_i\}$ be a dense set of unitaries.

Define a circuit approximation to V :

$$|U_k \dots U_2 U_1 - V| < \delta$$

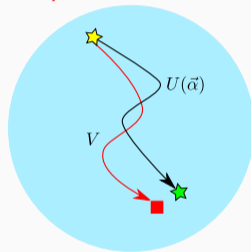
Scaling to best approximation

$$k \sim \mathcal{O}\left(\log^c \frac{1}{\delta}\right)$$

where $c < 4$.

⇒ The approximation is **efficient** and requires a **finite number of gates**.

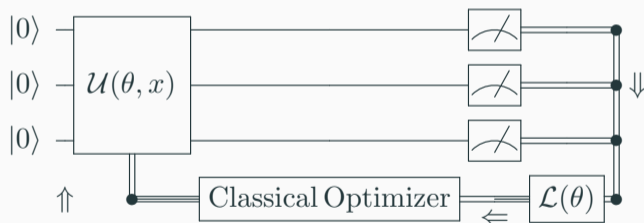
Optimal solution



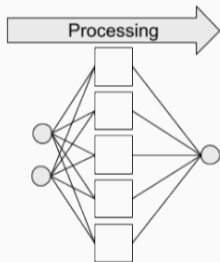
Why Quantum Machine Learning?

How do we parametrize models using a quantum computer?

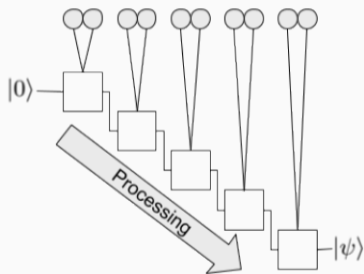
Using variational quantum circuits and data re-uploading algorithms:



Encode data directly “inside” circuit parameters:

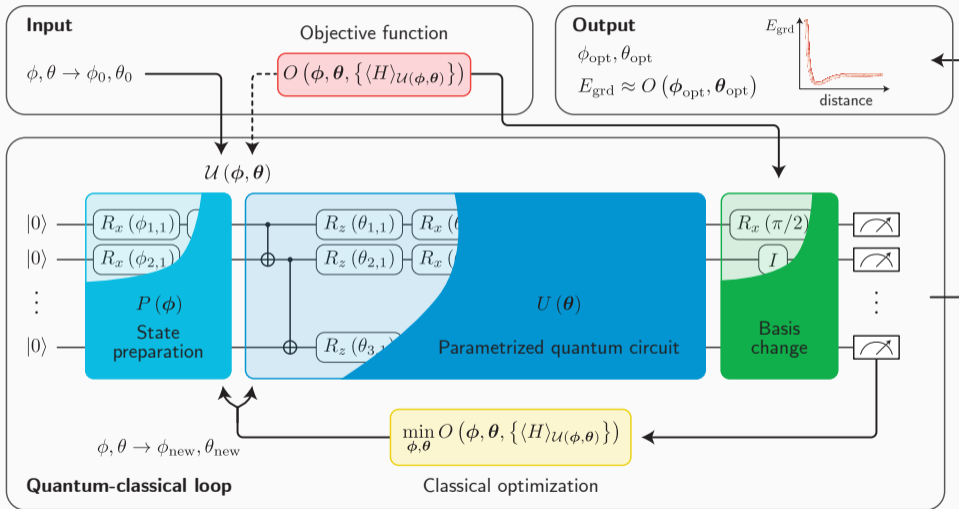


(a) Neural network



(b) Quantum classifier

Variational quantum algorithm



Visit the tutorial:

<https://colab.research.google.com/drive/1M4HV1RroiHtxh4uZdrSGASv51Tjpe6dT?usp=sharing>