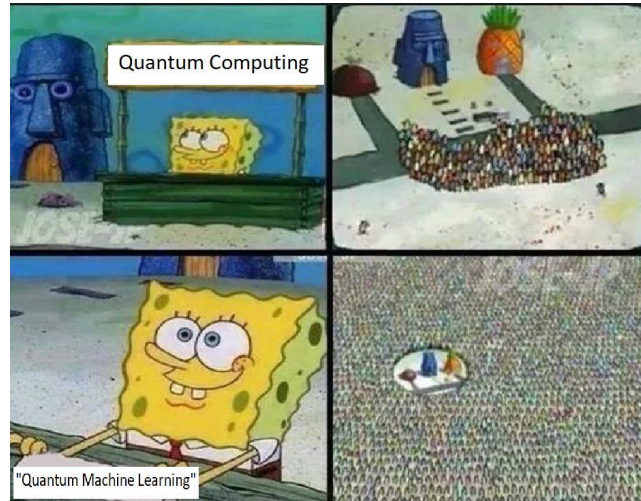


# Quantum Machine Learning Beyond the Hype

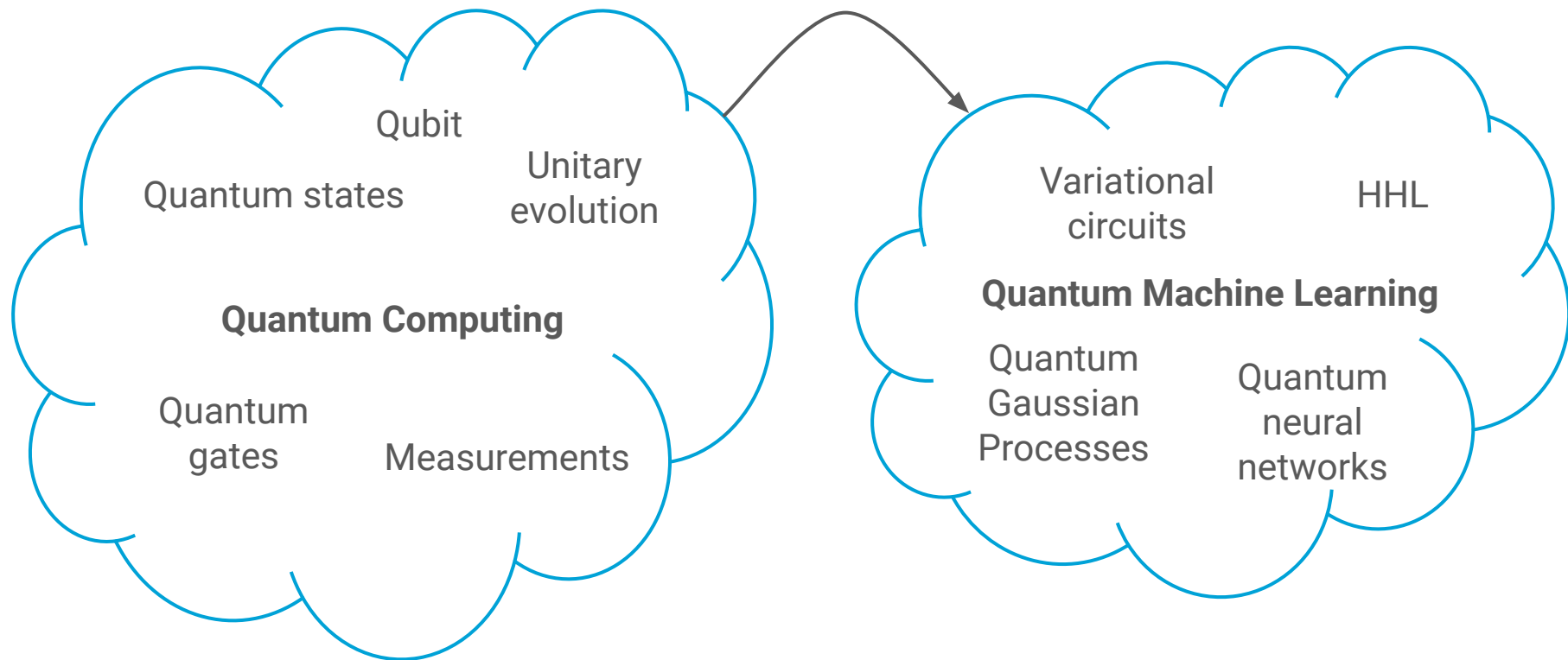
**Arthur Pesah**

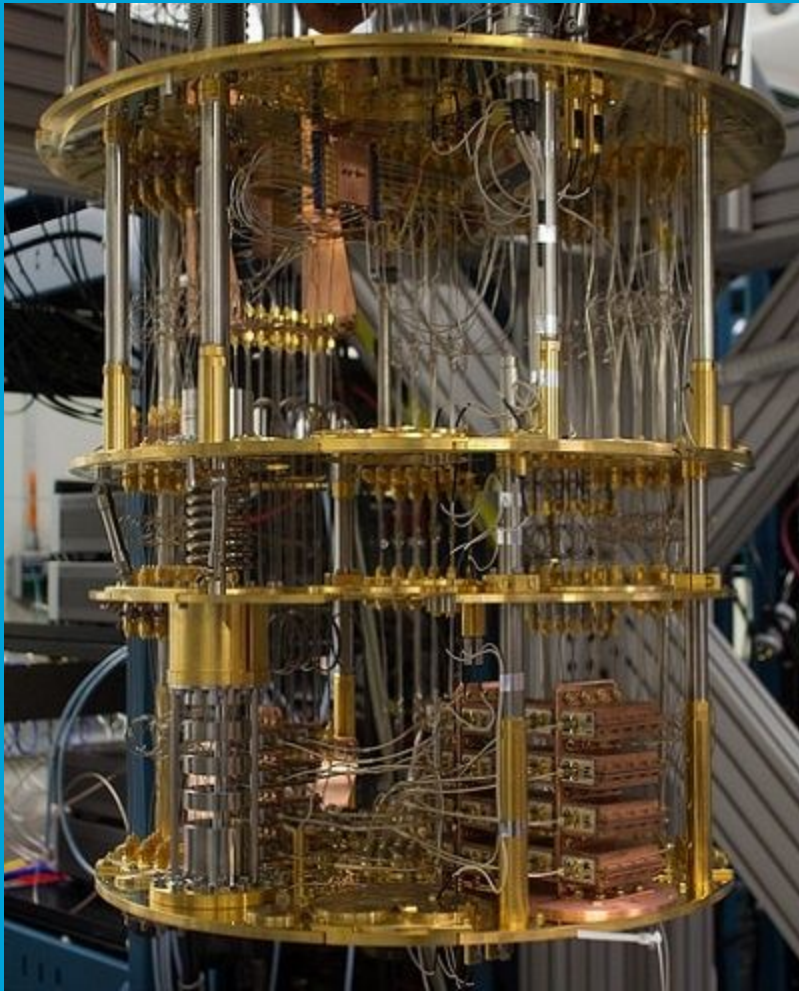
Previously 1QBit, Waterloo, Canada

KTH Royal Institute of Technology, Stockholm, Sweden



# What are you going to learn?





## Quantum Computing Overview

# Quantum computing: overview

## Do we have quantum computers?

Yes!

 Google AI  
53 qubits

 rigetti  
32 qubits

 XANADU  
8 qubits

 IONQ  
55 qubits

 IBM Q  
53 qubits

 D:WAVE  
The Quantum Computing Company™

3,000 (non-universal) qubits

Cloud access!

But...

...no quantum advantage has been shown on a practical application yet!

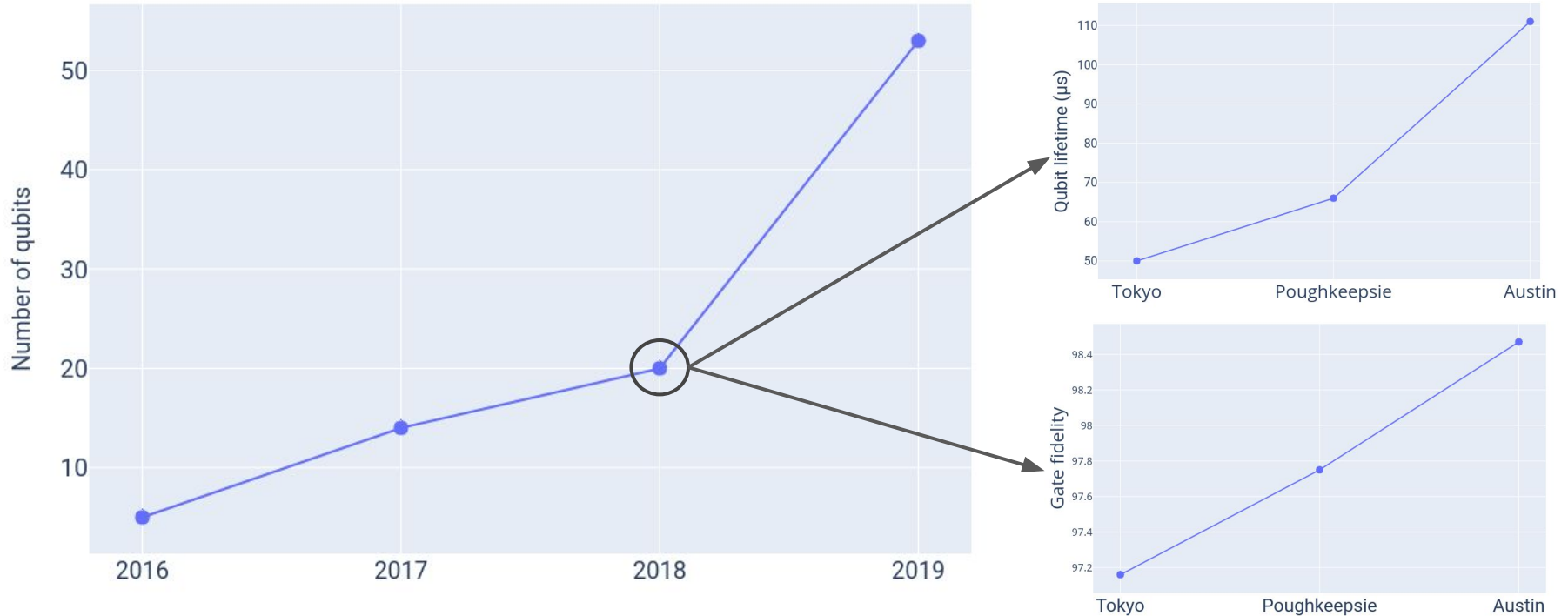
Why?

1. Not enough qubits
2. Qubits too noisy

# Quantum computing: overview

## Do we have quantum computers?

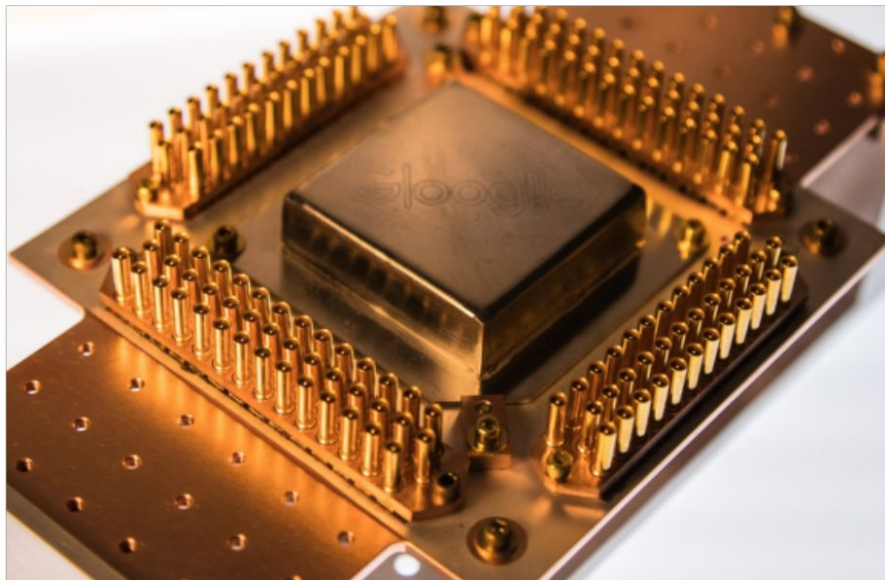
However, huge progress recently (e.g. IBM quantum computers)



# Quantum computing: overview

## Do we have quantum computers?

However, huge progress recently (e.g. quantum supremacy)



Quantum supremacy



Practical quantum advantage

# Quantum computing: overview

## How can we achieve a practical quantum advantage?

### Long-term

#### Traditional quantum algorithms

- Factoring algorithm (Shor)
- Search in unstructured database (Grover)
- Quantum machine learning (QSVM, QPCA...)

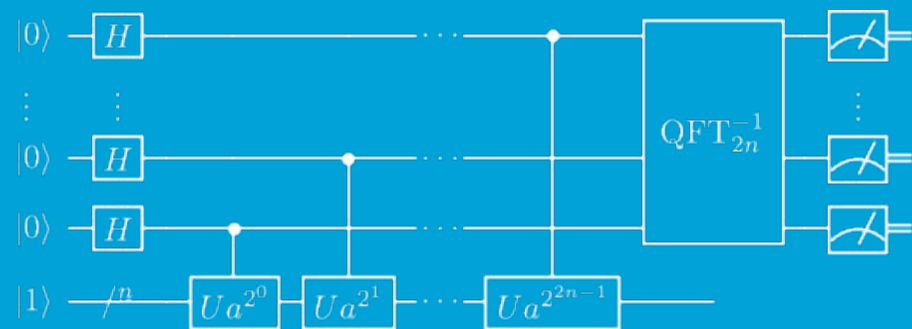
Very well understood algorithms, with precise bounds and complexity-theoretic advantage

### Intermediate-term

#### Noisy Intermediate-term Quantum era (NISQ)

- Quantum simulation (chemistry, material...)
- Optimization problems
- Quantum “neural networks”

Very recent heuristics (>2014), as complicated to analyse as regular neural networks



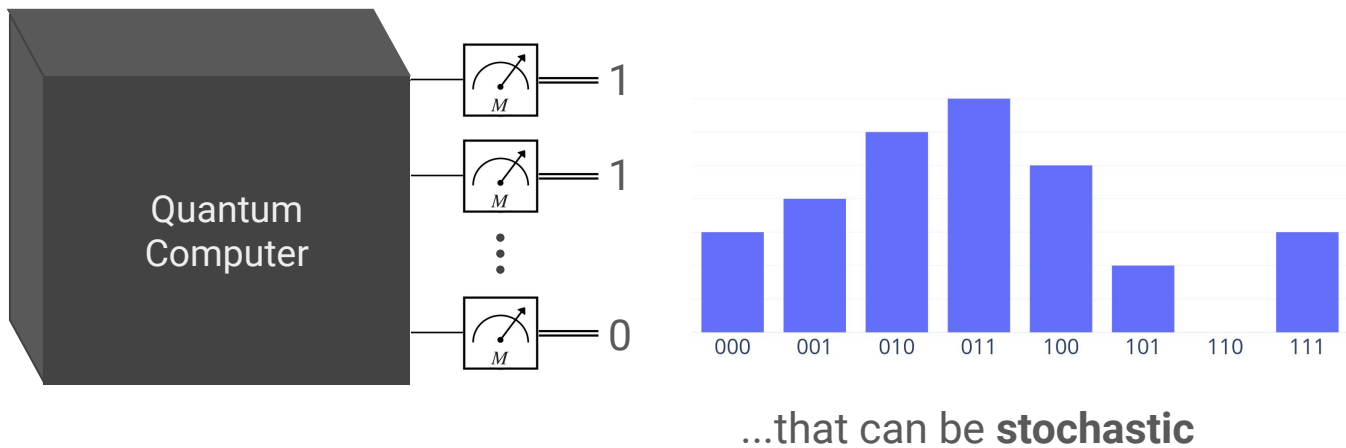
## Quantum Computing Formalism



# Quantum computing: formalism

## What is a quantum computer?

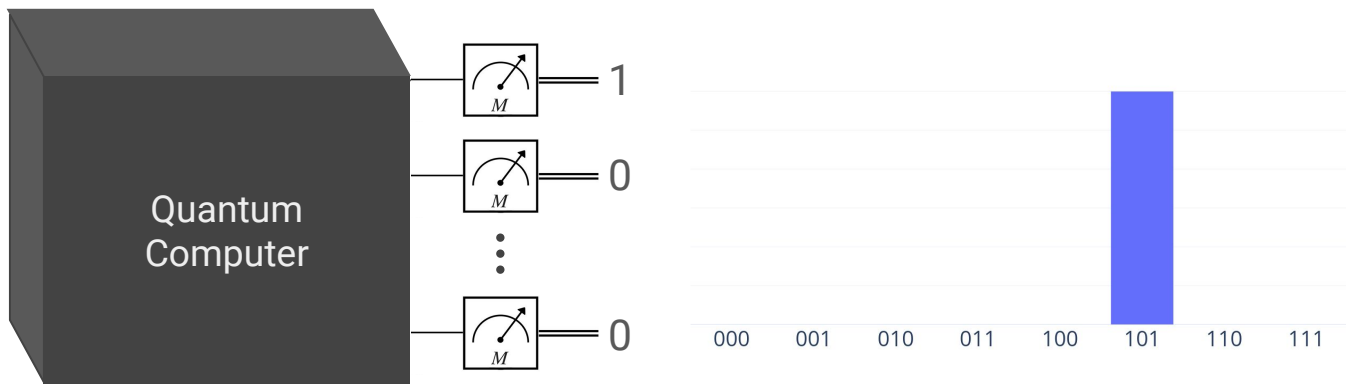
A quantum computer is a **generative model**...



# Quantum computing: formalism

## What is a quantum computer?

A quantum computer is a **generative model**...

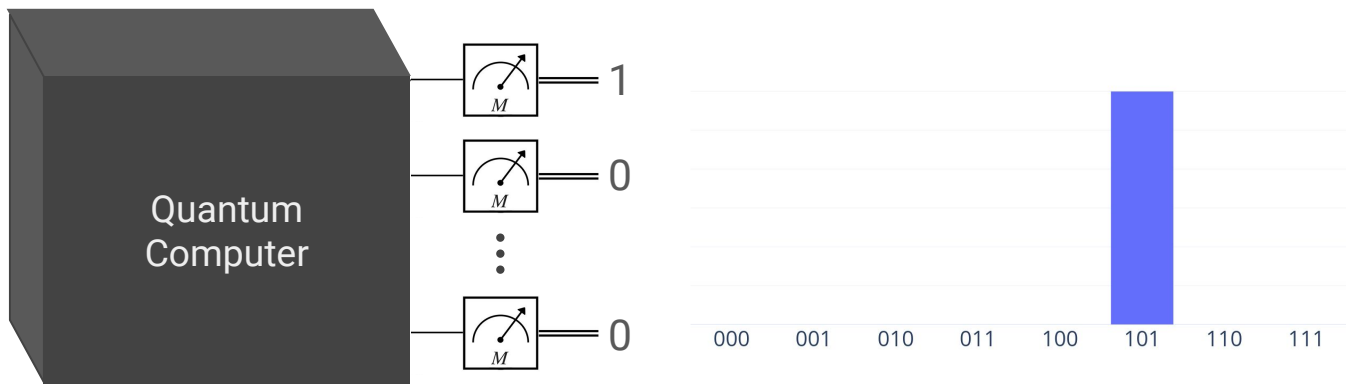


...that can be **deterministic**

# Quantum computing: formalism

## What is a quantum computer?

A quantum computer is a **generative model**...

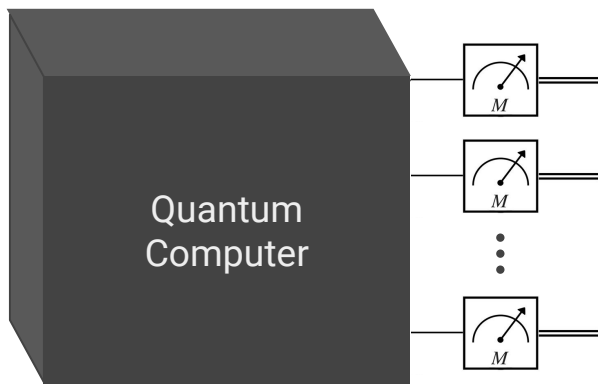


...that is very **efficient** on **specific problems**

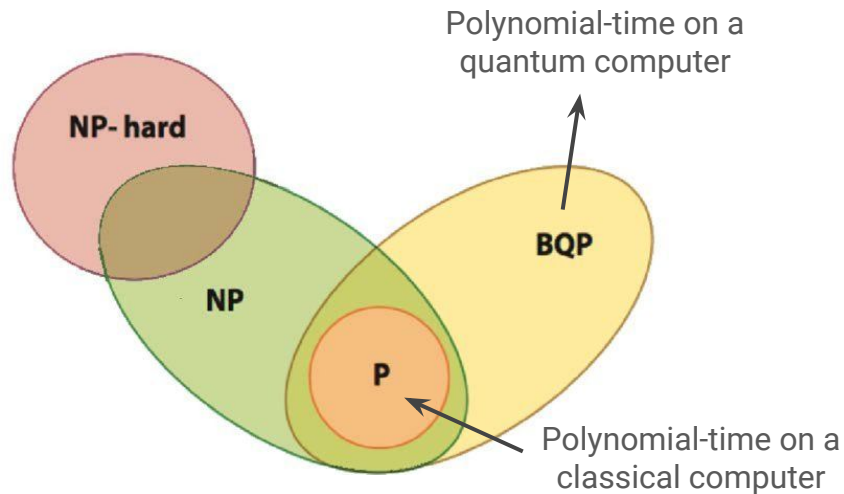
# Quantum computing: formalism

## What is a quantum computer?

A quantum computer is **not** equivalent to a Turing machine:



It **cannot** be efficiently simulated  
by a classical computer...



...and has different **complexity classes**

# Quantum computing: formalism

## Quantum physics as a generalized probability theory

	Probabilistic bit	Quantum bit (Qubit)
State	$\mathbf{p} = \begin{pmatrix} p_0 \\ p_1 \end{pmatrix} \in \mathbb{R}^2, \mathbf{p} \geq 0, \ \mathbf{p}\ _1 = 1$	$\psi = \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} \in \mathbb{C}^2, \ \psi\ _2 = 1$

# Quantum computing: formalism

## Quantum physics as a generalized probability theory

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Probability of event i	$p_i$	$ a_i ^2$

# Quantum computing: formalism

## Quantum physics as a generalized probability theory

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Probability of event i	$p_i$	$ a_i ^2$
Evolution	<b>Stochastic matrix</b> $\ S\mathbf{p}\ _1 = \ \mathbf{p}\ _1$	<b>Unitary matrix</b> $\ U\psi\ _2 = \ \psi\ _2$

**Notation:**  $|0\rangle := e_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$$|1\rangle := e_1 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$|\psi\rangle := \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} = a_0|0\rangle + a_1|1\rangle$$

# Quantum computing: formalism

## Quantum circuits

### Notation:

$$|\psi_{out}\rangle = U|\psi_{in}\rangle : \quad |\psi_{in}\rangle \text{ --- } \boxed{U} \text{ --- } |\psi_{out}\rangle$$

Quantum gate  
↙

### Examples:

- Hadamard Gate  $H = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$

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

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

- Rotation Gate  $R(\theta) = \begin{pmatrix} \cos\left(\frac{\theta}{2}\right) & -\sin\left(\frac{\theta}{2}\right) \\ \sin\left(\frac{\theta}{2}\right) & \cos\left(\frac{\theta}{2}\right) \end{pmatrix}$

$$|0\rangle \text{ --- } \boxed{R(\theta)} \text{ --- } \cos\left(\frac{\theta}{2}\right)|0\rangle + \sin\left(\frac{\theta}{2}\right)|1\rangle$$

$$|1\rangle \text{ --- } \boxed{R(\theta)} \text{ --- } -\sin\left(\frac{\theta}{2}\right)|0\rangle + \cos\left(\frac{\theta}{2}\right)|1\rangle$$

- Phase Gate  $P(\phi) = \begin{pmatrix} 1 & 0 \\ 0 & e^{i\phi} \end{pmatrix}$

$$|0\rangle \text{ --- } \boxed{P(\phi)} \text{ --- } |0\rangle$$

$$|1\rangle \text{ --- } \boxed{P(\phi)} \text{ --- } e^{i\phi}|1\rangle$$



# Quantum computing: formalism

## Multiple qubits

**2-qubit state:**  $|\psi\rangle = \begin{pmatrix} a_{00} \\ a_{01} \\ a_{10} \\ a_{11} \end{pmatrix} = a_{00}|00\rangle + a_{01}|01\rangle + a_{10}|10\rangle + a_{11}|11\rangle \in \mathbb{C}^4$

**3-qubit state:**  $|\psi\rangle = a_{000}|000\rangle + a_{001}|001\rangle + \dots + a_{111}|111\rangle \in \mathbb{C}^8$

$\vdots$

**n-qubit state:**  $|\psi\rangle = \sum_{i=1}^{2^n} a_i |i\rangle \in \mathbb{C}^{2^n} \longrightarrow \text{Exponentially-large storage}$

Could we use qubits as a memory for classical data?

**Not simple:** it can be exponentially-hard to retrieve those amplitudes!

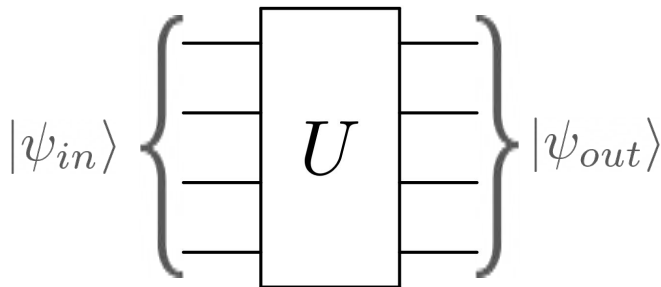
$\longrightarrow$  **Quantum state tomography**

# Quantum computing: formalism

## Multiple qubits

What about the **gates**?

$$U \in \mathcal{U}(2^n)$$



Physically **efficient** operations...

...that can perform **exponentially-large** matrix multiplications

Quantum “parallelism”? **Yes and No**

It's what makes quantum computers powerful

You can only retrieve the result efficiently in very specific cases

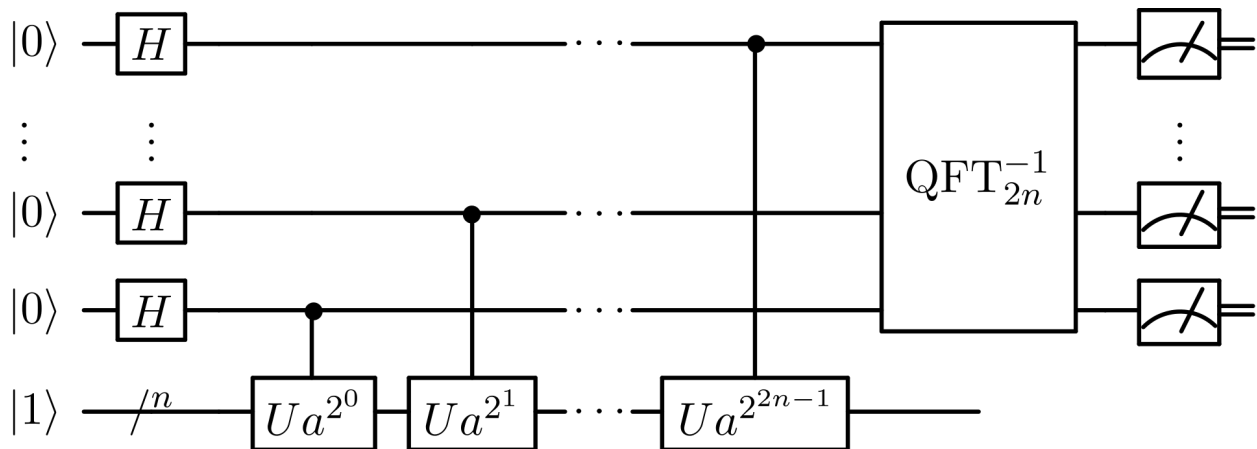
# Quantum computing: formalism

## Quantum Algorithms

### Shor's Algorithm

**Goal:** Factor a number  $N$  into its prime factors

**Speed-up:** exponential



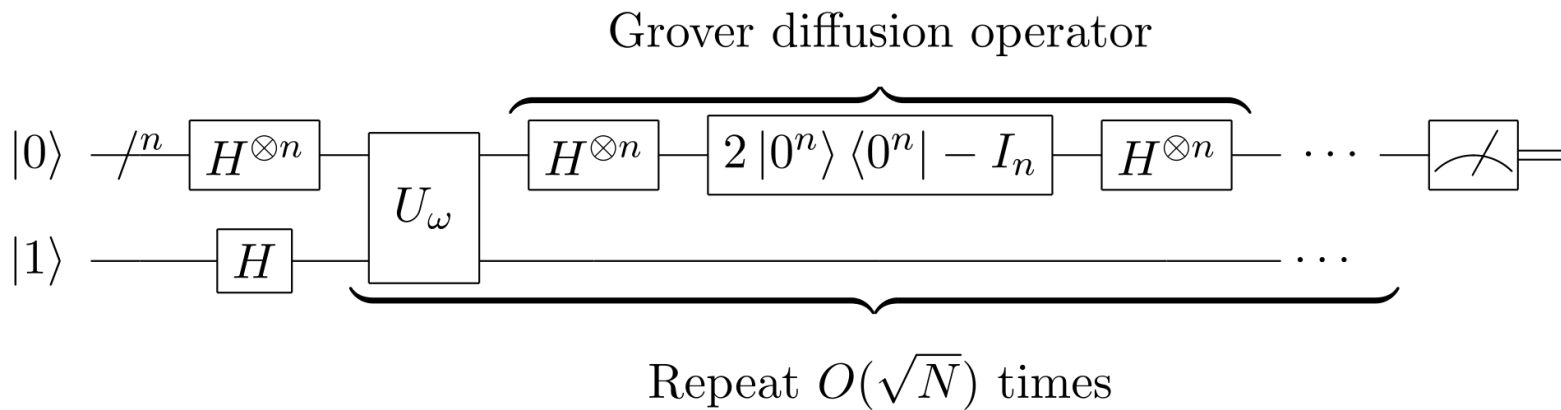
# Quantum computing: formalism

## Quantum Algorithms

### Grover Algorithm

**Goal:** Search an element in an unstructured list

**Speed-up:** quadratic



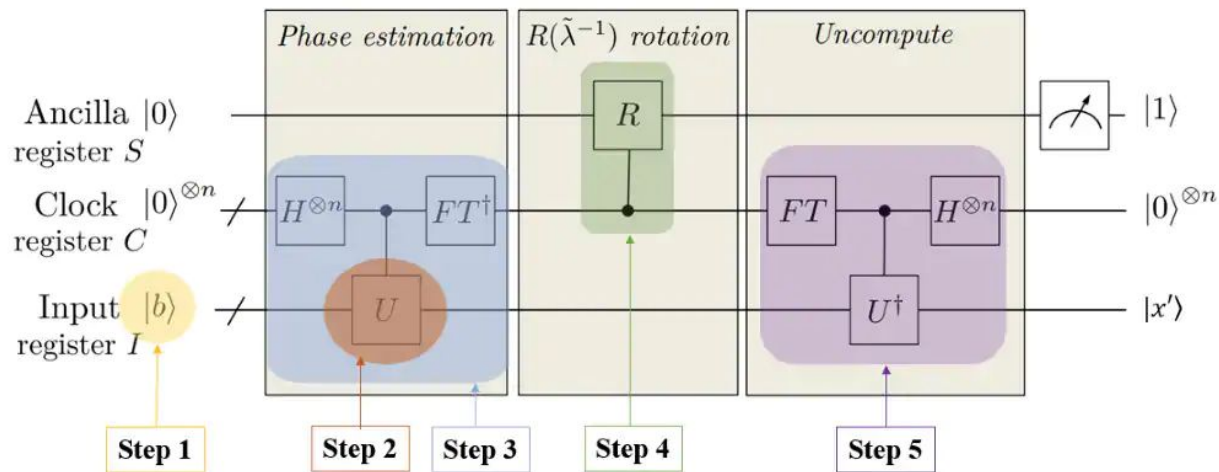
# Quantum computing: formalism

## Quantum Algorithms

### HHL Algorithm

**Goal:** Solve well-conditioned linear system of equations  $A\mathbf{x} = \mathbf{b}$

**Speed-up:** exponential (with some major caveats)



“Most **overhyped** and **underestimated**  
field in quantum computing”,  
Iordanis Kerenidis (Paris Diderot)

## Quantum Machine Learning

# Quantum machine learning: overview

## What is quantum machine learning?

		Type of Algorithm	
		<i>classical</i>	<i>quantum</i>
Type of Data	<i>classical</i>	CC	CQ
	<i>quantum</i>	QC	QQ

# Quantum machine learning: overview

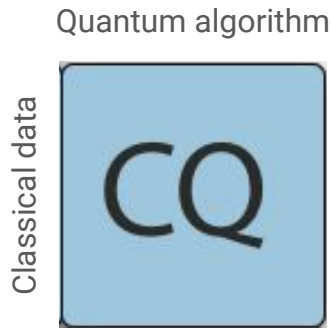
## What is quantum machine learning?

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Type of Data	<i>classical</i>	CC	CQ
	<i>quantum</i>	QC	QQ



# Quantum machine learning: overview

## What is quantum machine learning?



### **First-wave QML**

*(from ~2010)*

Fault-tolerance devices

Theoretical guarantees

Requires QRAM

QSVM, QPCA, QBoost, Q-Means, etc.

### **Second-wave QML**

*(from ~2016)*

Noisy devices

Heuristics

No QRAM

Quantum NN, Quantum Kernels

# Quantum machine learning: first-wave

## Example: Quantum Gaussian Processes (GP)

### Classical GP:

The mean and std of a GP with kernel  $K$  is given by:

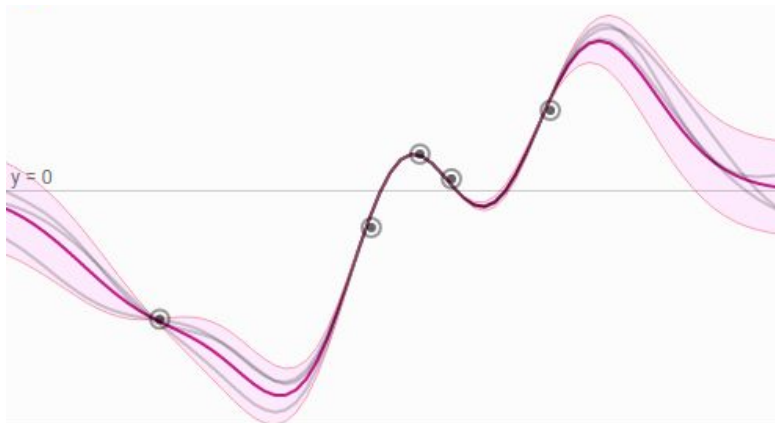
$$\mu_* = \mathbf{k}_*^T (K + \sigma^2 I_n)^{-1} \mathbf{y}$$

$$\sigma_* = k(\mathbf{x}_*, \mathbf{x}_*) - \mathbf{k}_*^T (K + \sigma^2 I_n)^{-1} \mathbf{k}_*$$

### Quantum GP:

To calculate the mean:

1. Prepare states  $|\mathbf{k}_*\rangle$  and  $|\mathbf{y}\rangle$  on a quantum RAM
2. Calculate  $|\mathbf{b}\rangle = (K + \sigma^2 I_n)^{-1} |\mathbf{y}\rangle$  using HHL
3. Calculate the inner product  $\langle \mathbf{b} | \mathbf{y} \rangle$



Source: [distill.pub/2019/visual-exploration-gaussian-processes/](https://distill.pub/2019/visual-exploration-gaussian-processes/)

# Quantum machine learning: first-wave

## Caveat 1: QRAM might not be feasible

1. It might be too slow to have any advantage
2. Requires too many qubits for short-term applications



# Quantum machine learning: first-wave

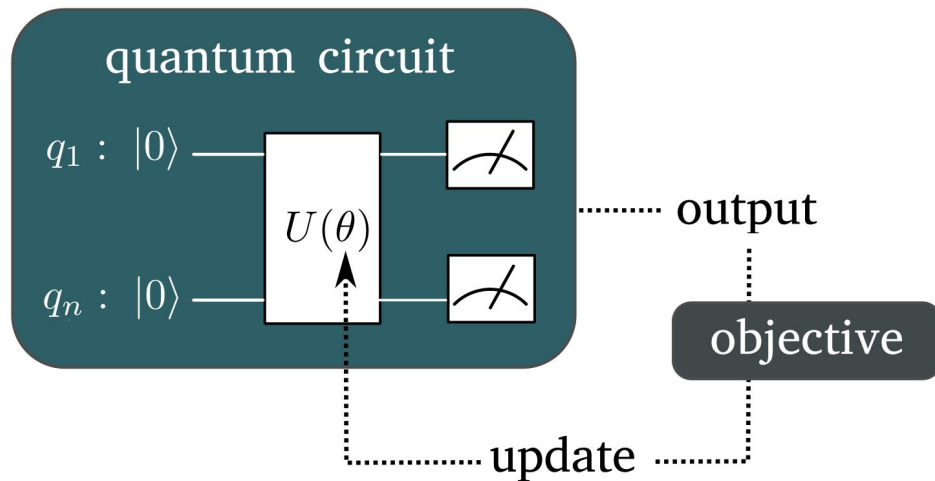
## Caveat 2: Beware dequantization!



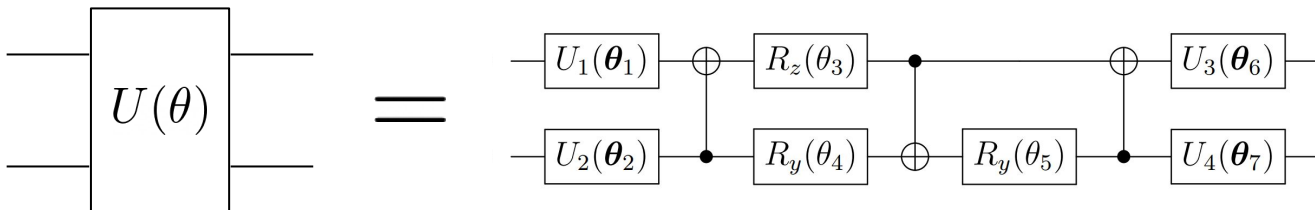
1. Quantum-inspired algorithms with same performance as purely quantum can be constructed
2. Dequantized algorithms: recommendation systems, low-rank HHL, PCA...
3. But constant factors matter!

# Quantum machine learning: second-wave

## Variational circuits

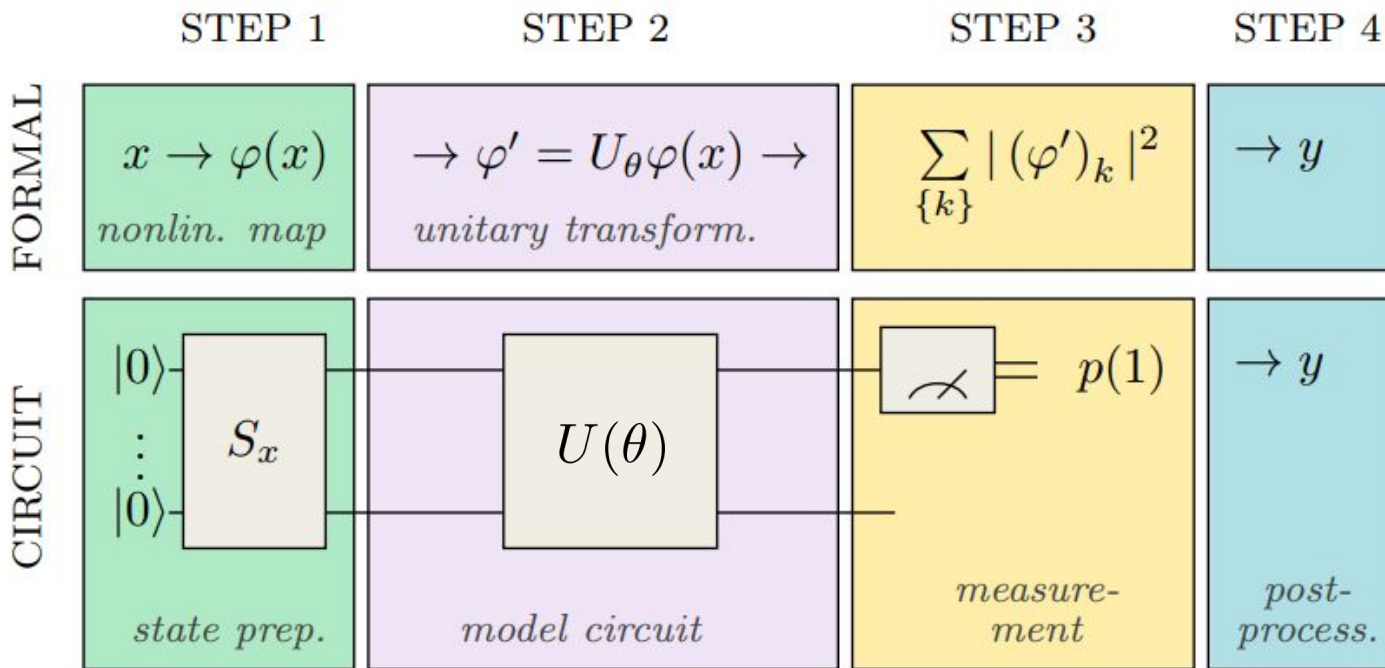


Example of **ansatz**:



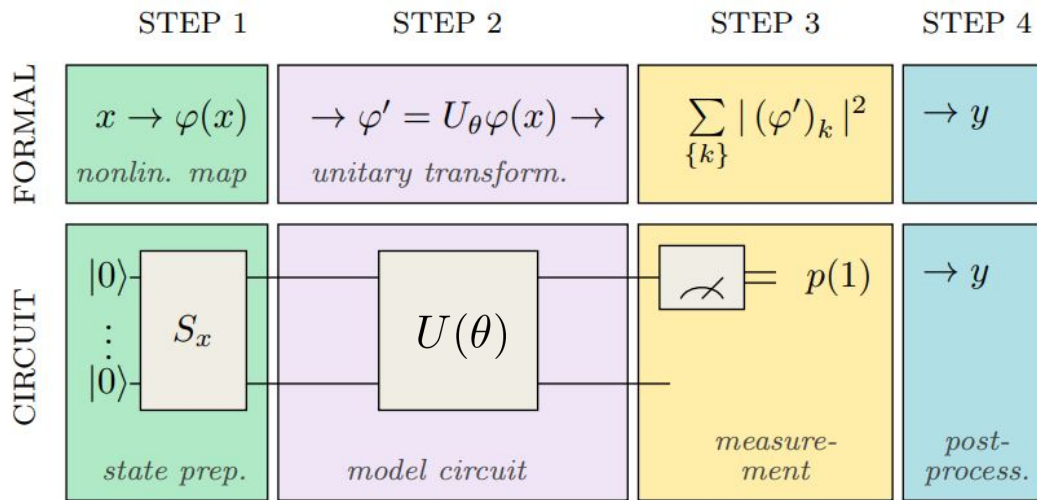
# Quantum machine learning: second-wave

## Variational quantum classifier



# Quantum machine learning: second-wave

## Variational quantum classifier



### Optimization

Quantum stochastic gradient descent

### Potential advantage

Better inference time for some problems

### Caveat

No proof of advantage, empirical demonstration on toy models

# Quantum machine learning: second-wave

## How can I try it myself?

### My favorite libraries for quantum ML

#### PennyLane (Python):

- Easy to use
- Can compute quantum and classical gradients
- Interface with PyTorch and Tensorflow
- Can connect to other simulators and real devices

#### Yao (Julia):

- Very modular and flexible
- Fastest simulator currently available
- Automatic differentiation as well

```
import pennylane as qml
from pennylane import numpy as np

# create a quantum device
dev1 = qml.device('default.qubit', wires=1)

@qml.qnode(dev1)
def circuit(phi1, phi2):
    # a quantum node
    qml.RX(phi1, wires=0)
    qml.RY(phi2, wires=0)
    return qml.expval(qml.PauliZ(0))

def cost(x, y):
    # classical processing
    return np.sin(np.abs(circuit(x, y))) - 1

# calculate the gradient
dcost = qml.grad(cost, argnum=[0, 1])
```



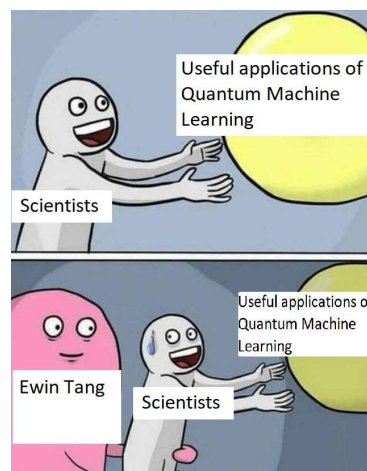
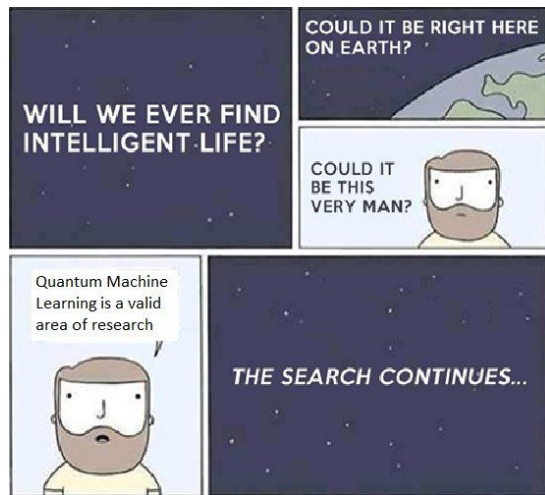
# Discussion

QML is the “most **overhyped** and **underestimated** field in quantum computing” (Iordanis Kerenidis)

**Overhyped:** lot of fuss, but too early to predict if QML will ever be useful: no perfect QML algorithm has been found so far

**Underestimated:**

- Research in this field  $\Rightarrow$  new classical algorithms discovered!
- Dequantization only discovered in 2018  $\Rightarrow$  non-dequantizable algorithms might still be found!
- Second-wave QML very similar to early deep learning research!
- Only a small community actively working on QML!



When someone tries to show you a Quantum Machine Learning paper

