



Quantum Machine Learning

Speaker: Chahan Kropf

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19 Dec. 2019

Who are you?

Master



PostDoc



Bachelor



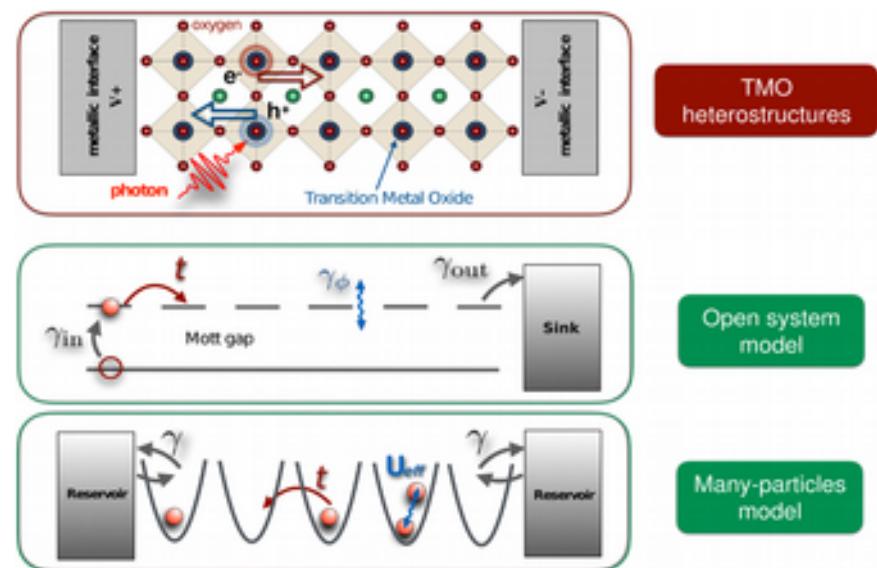
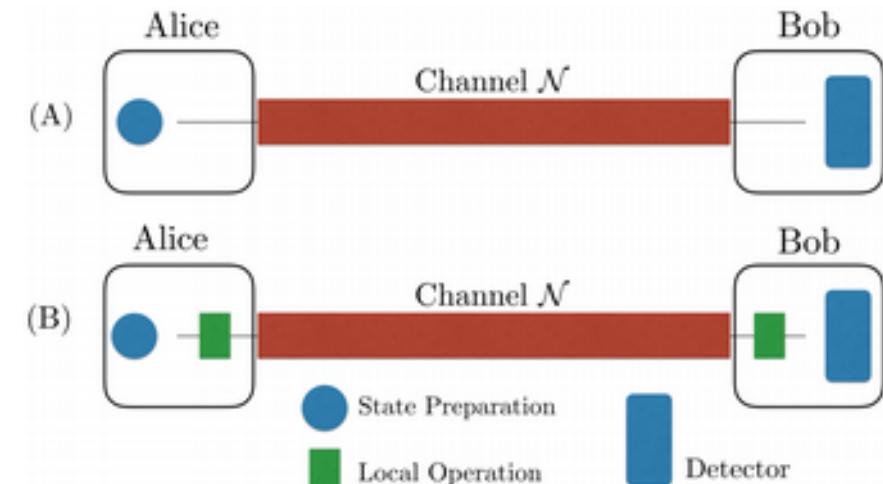
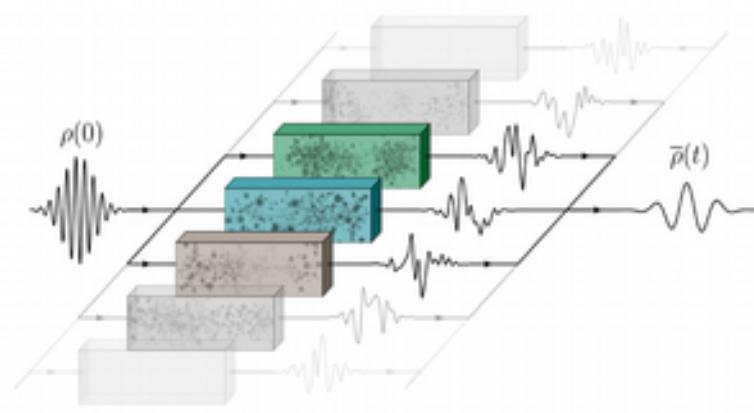
PhD



- **Neuchâtel - Heidelberg - Freiburg - Brescia**

What do you do ? Quantum physics!

- Optimal quantum transport
- Dynamics of disordered quantum systems
- Quantum information and cryptography (channel coding of a measurement)
- Quantum machine learning
- Foundations of quantum mechanics



Outline

- **Quantum machine learning**
 - What?
 - Why?
- **Quantum mechanics**
 - What is it?
 - Quantum computing
- **Example of a quantum machine learning algorithm**
 - Quantum square distance classifier
 - Python code for IBM Qcomputer

Quantum Machine Learning

- Quantum communication and cryptography



Quantum Machine Learning

- Quantum Computing

NATURE BRIEFING · 23 OCTOBER 2019

Daily briefing: Google's quantum-supremacy paper is here

A computing milestone, a 'failed' Alzheimer's drug is back and *Nature*'s new look.

4000 qubits



53 qubits



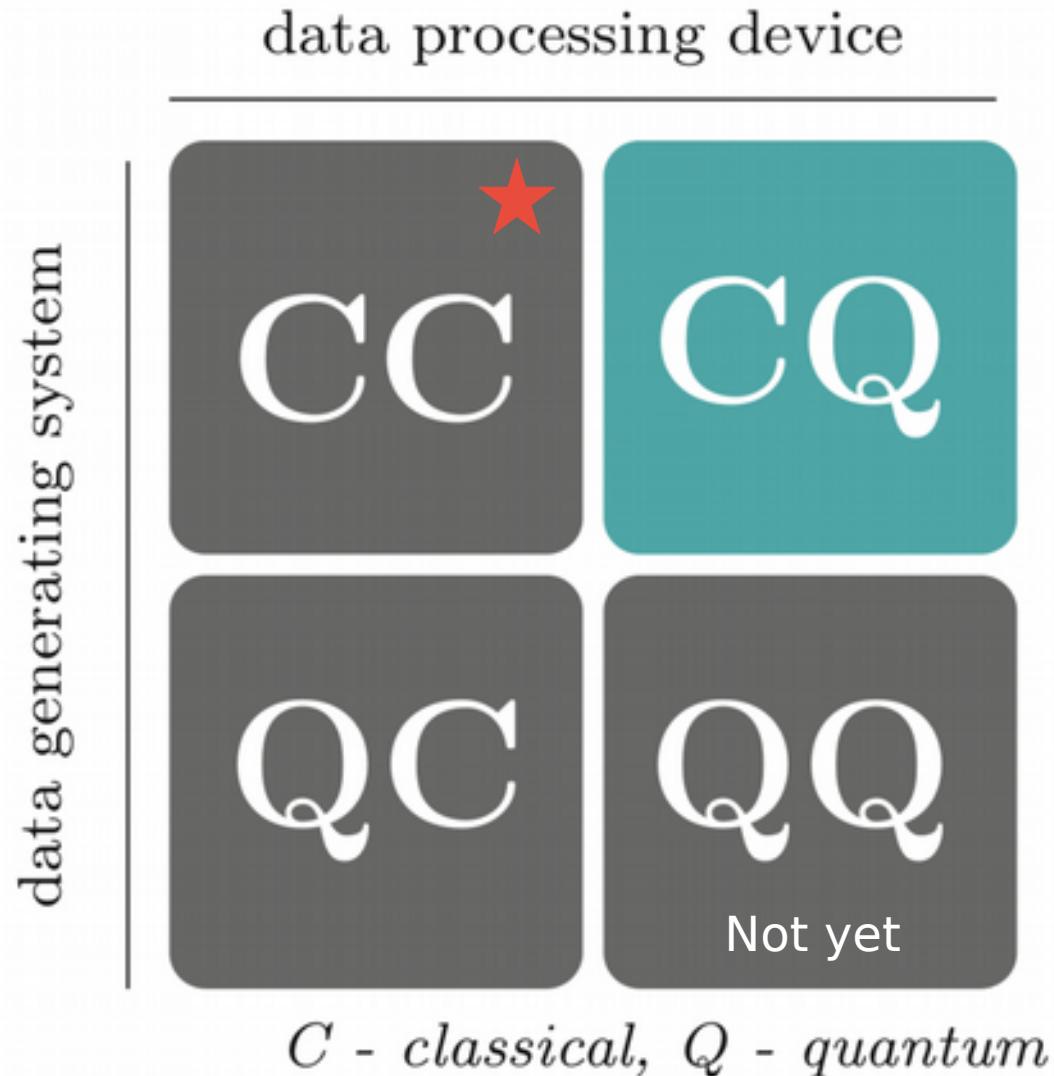
Quantum Machine Learning

Players



Thanks to Maria Schuld from Xanadu

Quantum Machine Learning

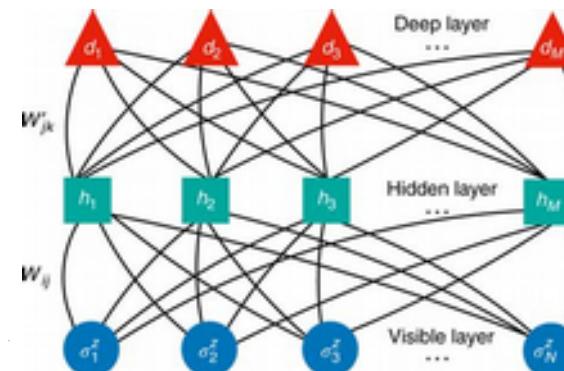


Quantum - Classical (QC)

Experimental data



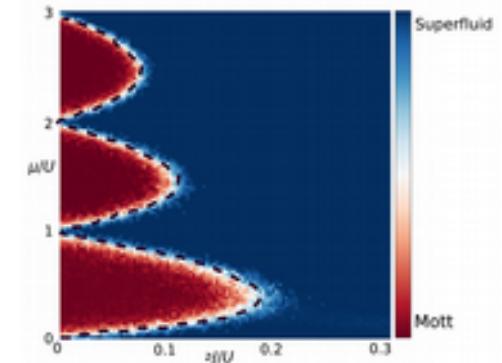
ML algorithm



Theoretical Model

$$i\hbar \frac{d}{dt} \psi(\vec{r}, t) = \left[\frac{-\hbar^2}{2m} \nabla^2 + V(\vec{r}, t) \right] \psi(\vec{r}, t)$$

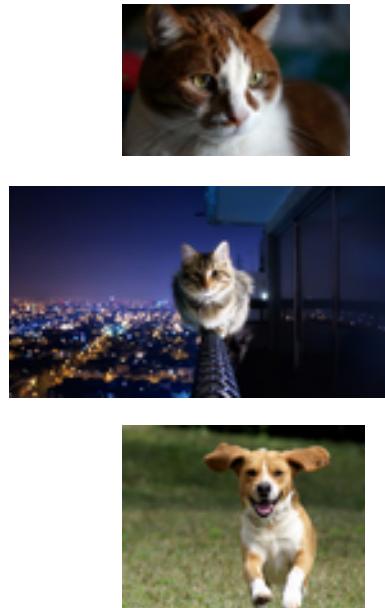
Result



- Quantum processor/experiment optimal design
- Material physics - phase transitions (high-temperature superconductivity)

Classical - Quantum (CQ)

Normal data



QML algorithm



Result

Cat
Or
Dog



- **Optimization (annealing)**
- **QPCA, QSVM, QNN, Qclustering, QBM, ...**

Quantum inspired Classical

- **Quantum information theory can help to understand classical machine learning and improve it.**
- **Example: learn correlations between two datasets and learn relevant features**

Learning relevant features for statistical inference

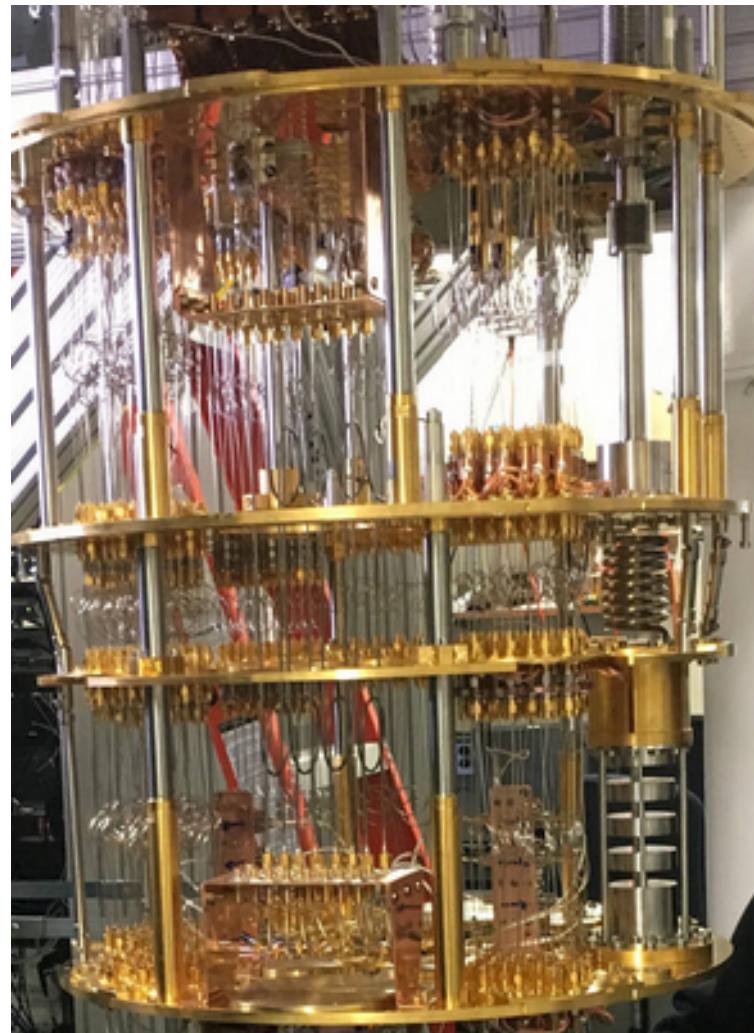
Cédric Bény¹

¹*Department of Applied Mathematics, Hanyang University (ERICA),
55 Hanyangdaehak-ro, Ansan, Gyeonggi-do, 426-791, Korea.*

We introduce an algorithm that learns correlations between two datasets, in a way which can be used to infer one type of data given the other. The approach allows for the computation of expectation values over the inferred conditional distributions, such as Bayesian estimators and their standard deviations. This is done by learning feature maps which span hyperplanes in the spaces of probabilities for both types of data, optimized to optimally represent correlations. When applied to supervised learning, this yields a new objective function which automatically provides regularization and results in faster convergence. We propose that, in addition to many applications where two correlated variables appear naturally, this approach could also be used to identify dominant independent features of a single dataset in an unsupervised fashion: in this scenario, the second variables should be produced from the original data by adding noise in a manner which defines an appropriate information metric.

Why QML?

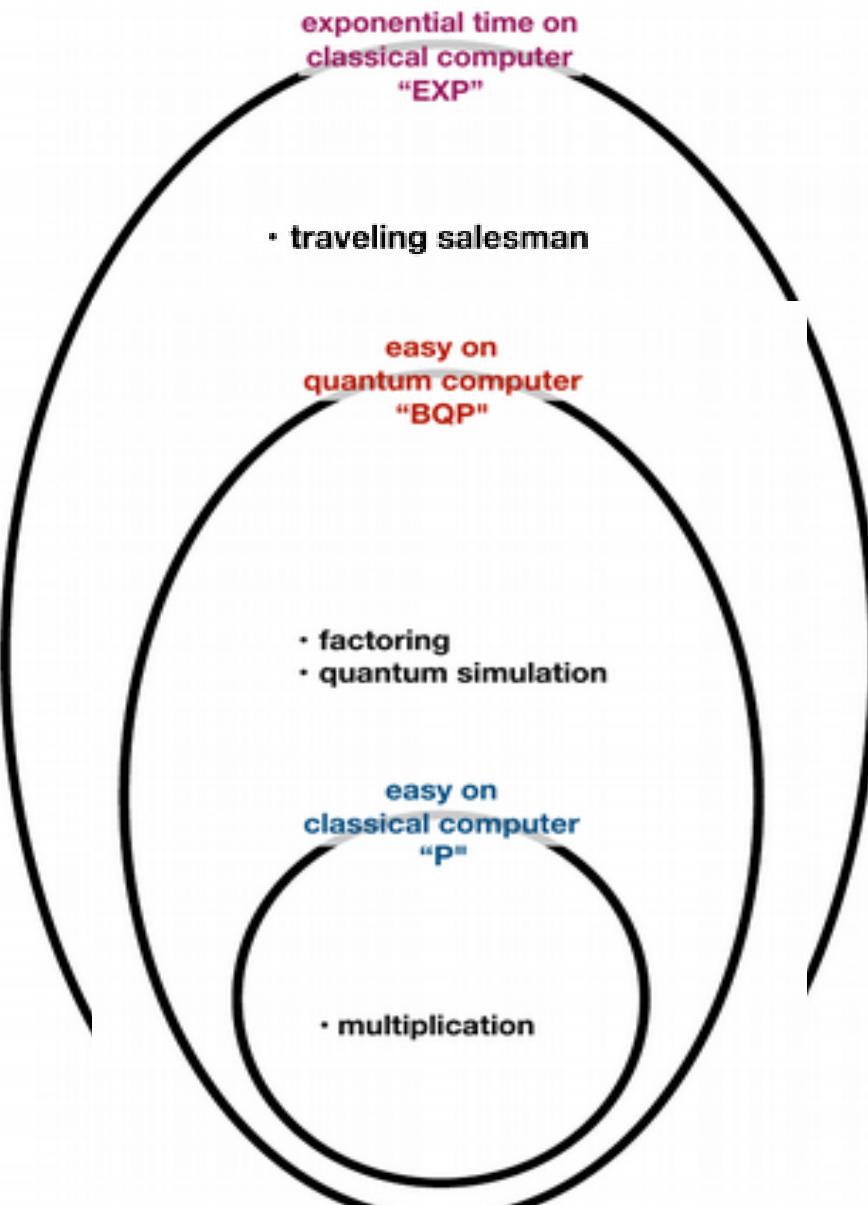
- **Solve complex quantum problems with ML**
- **Better ML from quantum information theory**
- **Faster full ML algorithms on quantum computer**
- **Faster optimization routines**
- **Better sampling methods**
- **Entirely new type of AI and algorithms**
- **QML may be the current best reason to actually develop quantum computers**



The power of quantum computers

- **Potential:**
 - Algorithm speed-up
 - Quantum sampling
 - Direct quantum analysis
- **Ingredients**
 - Coherence
 - Entanglement
 - Interference
 - Stochasticity
 - Linear dynamics
 - Non-linear measurement
- **Difficulties**
 - Coherence is destroyed by measurement
 - No cloning makes error correction hard
 - Classical: 10^{-24} errors per gate, quantum 10^{-2} errors per gate

The power of quantum computer



- **BQP contains P:** a quantum computer can (quickly) do everything a classical computer can (quickly) do
- **P does not contain BQP:** a quantum computer can do some things quickly that a classical computer cannot do quickly
- **EXP contains BQP :** a classical computer can eventually do everything a quantum computer can do
- **BQP does not contain EXP:** a quantum computer does not give generic exponential speed-up

Timeline of quantum algorithms

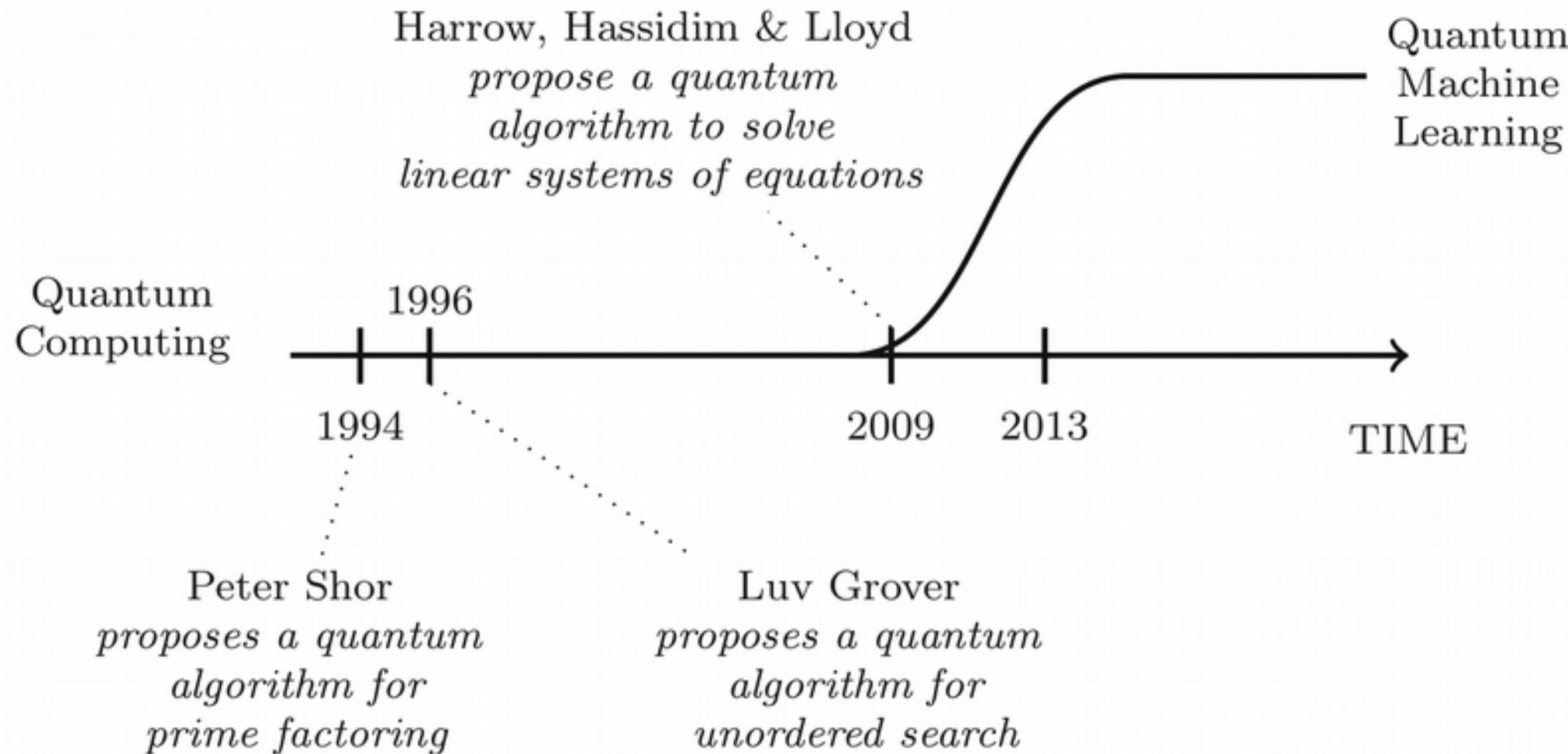
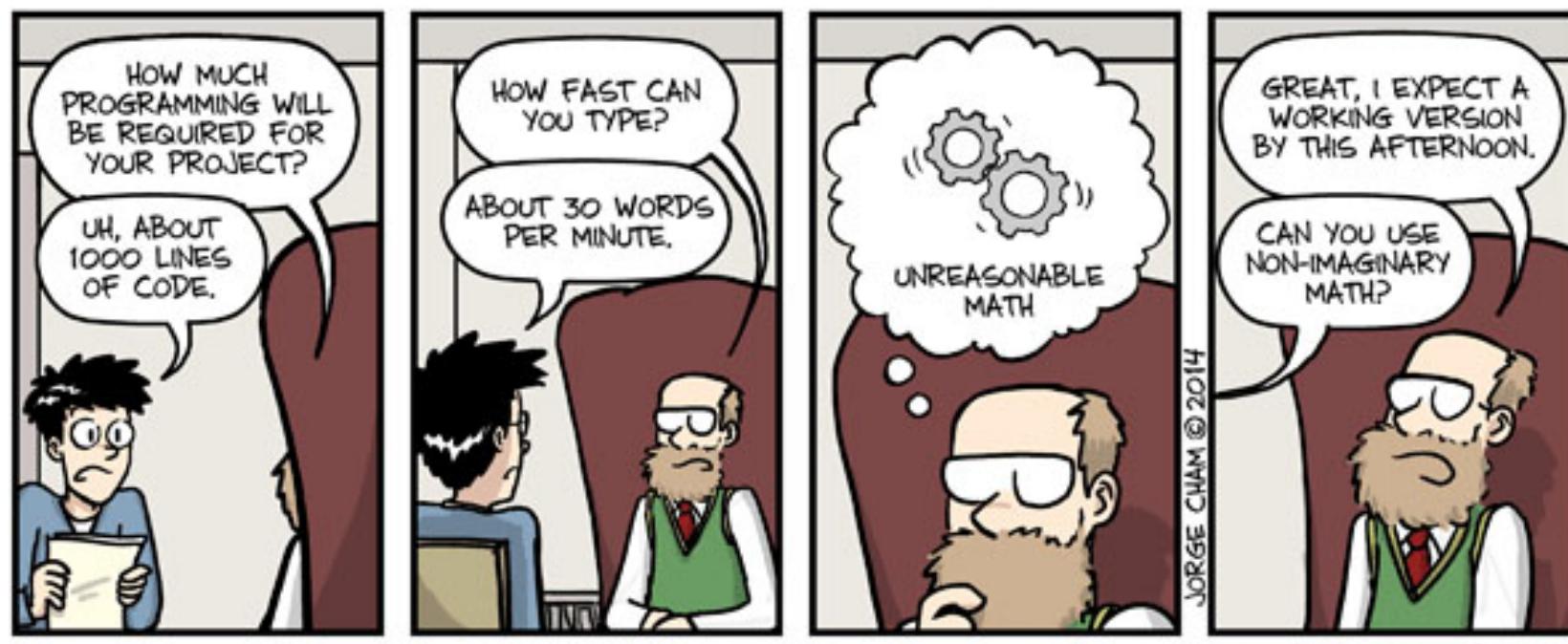


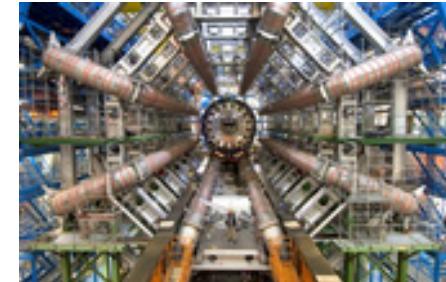
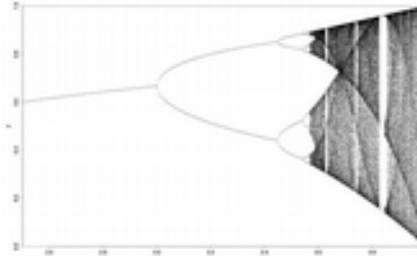
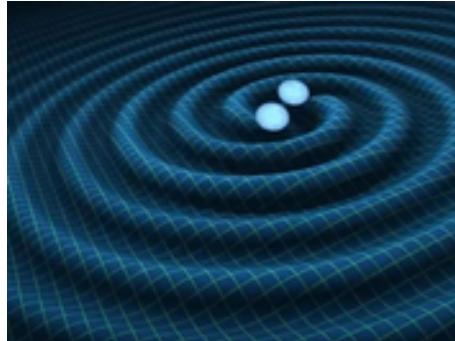
Fig. 3.3 Timeline of quantum computing and quantum machine learning

List of QML algorithms

- qSearch [Grover 1996]
- qPrincipal Component Analysis [Lloyd, Mohseni, Rebentrost 2013]
- qSupport Vector Machine [Rebentrost, Mohseni, Lloyd, 2013]
- qBayesian Network [Tucci 2012, Moreira, Wichert 2018]
- qHidden Markov Model [Clark, Huang, Barlow, Beige 2014, Cholewa, Gawron, Glomb, Kurzyk 2015, Srinivasan, Gordon, Boots 2017]
- qMatrix Operations, qBLAS [Harrow, Hassidim, Lloyd 2009, Le Gall 2012, Zhang, Zhang, Xue 2018, Shao 2018]
- qNeural Network [Schuld, Sinayskiy, Petruccione 2014, Neukart 2013, Ricks, Ventura 2004, Dorozhinsky, Pavlovsky 2018, Farhi, Neven 2018]
- qAnnealing [Behrman, Steck, Moustafa 2016, Li, Felice, Rohs, Lidar 2018]



But what is quantum?



- **General relativity**

- Curved space-time
 - Time dilation

- **Classical physics**

- Newton
 - Chaos

- **Quantum mechanics**

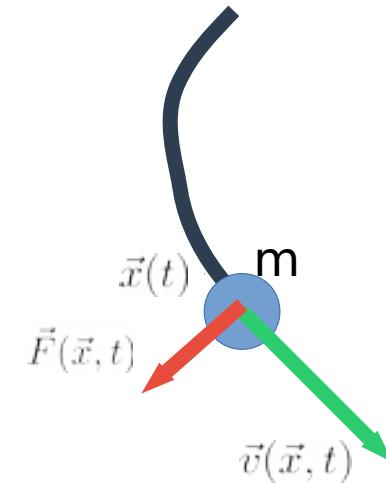
- Non-locality
 - Tunneling

Classical mechanics

- ‘points’ with masses and velocity
- Forces acting on ‘points’
- Newton’s law of motion

$$\vec{x}(t) \qquad \vec{v}(\vec{x}, t)$$

$$\vec{F} = m\vec{a}$$



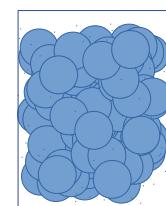
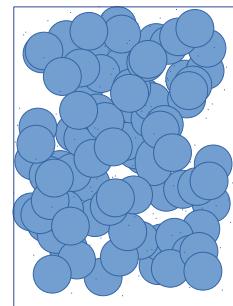
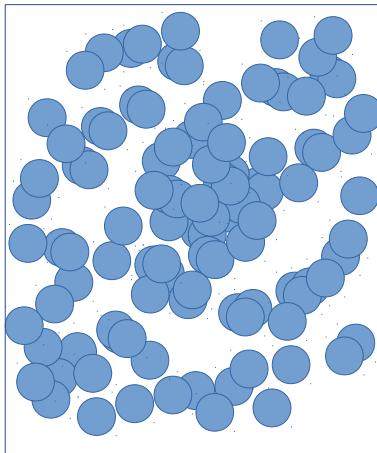
Statistical physics

- **Probability distribution of points**
- **Liouville equation for the distribution W (equivalent of Newton)**

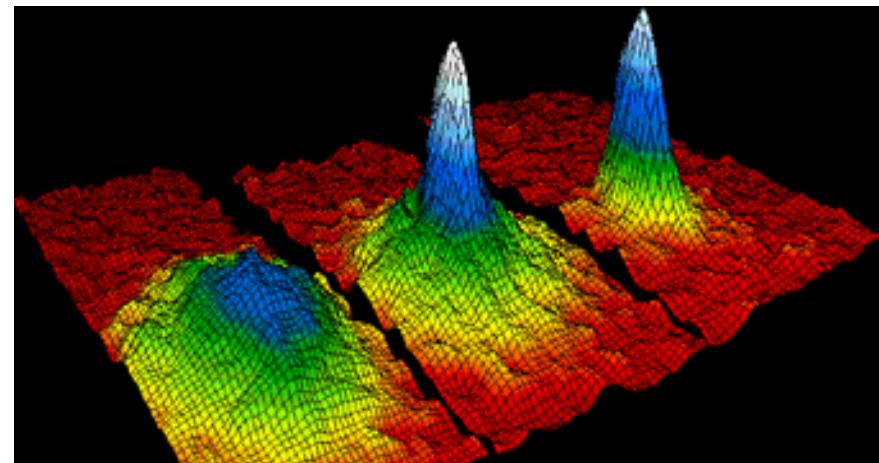
$$W(\vec{q}, \vec{p}, t)$$

$$\frac{\partial}{\partial t} W(\vec{q}, \vec{p}, t) = \{W(\vec{q}, \vec{p}, t), H(\vec{q}, t)\}$$

Particles moving around



Density probability distribution



$$W(\vec{q}, \vec{p}, t)$$

Quantum mechanics

- **Wavefunctions (complex object!)**

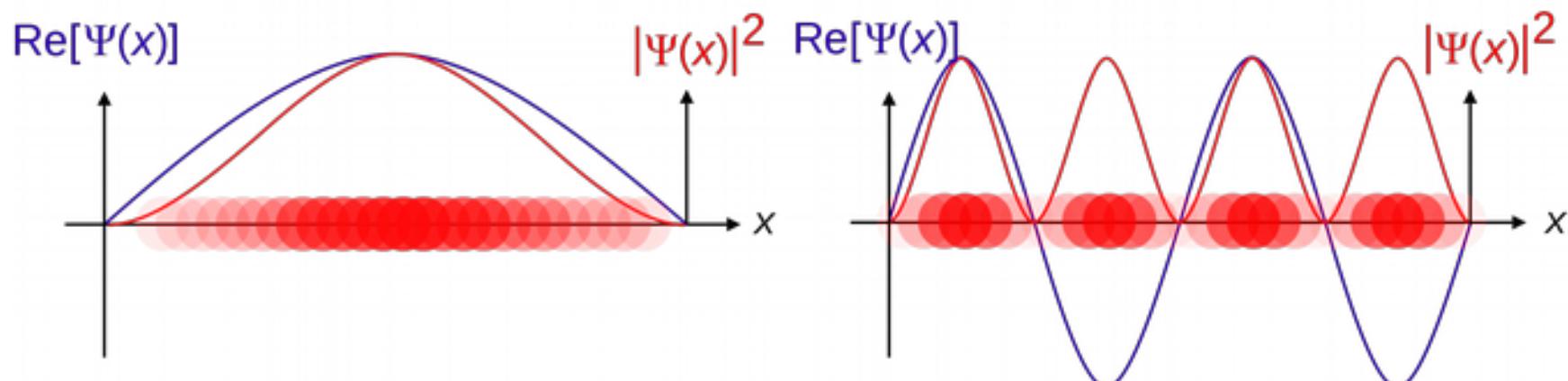
$$\psi(\vec{x}, t) \in \mathbb{C}$$

- **Schrödinger equation (linear!)**

$$i\hbar \frac{d}{dt} \psi(\vec{x}, t) = \left[-\frac{i\hbar}{2m} \nabla^2 + V(\vec{x}) \right] \psi(\vec{x}, t)$$

- **Outcome: probabilities**

$$p(\vec{x}, t) = |\psi(\vec{x}, t)|^2 = \psi(\vec{x}, t) \psi^*(\vec{x}, t)$$



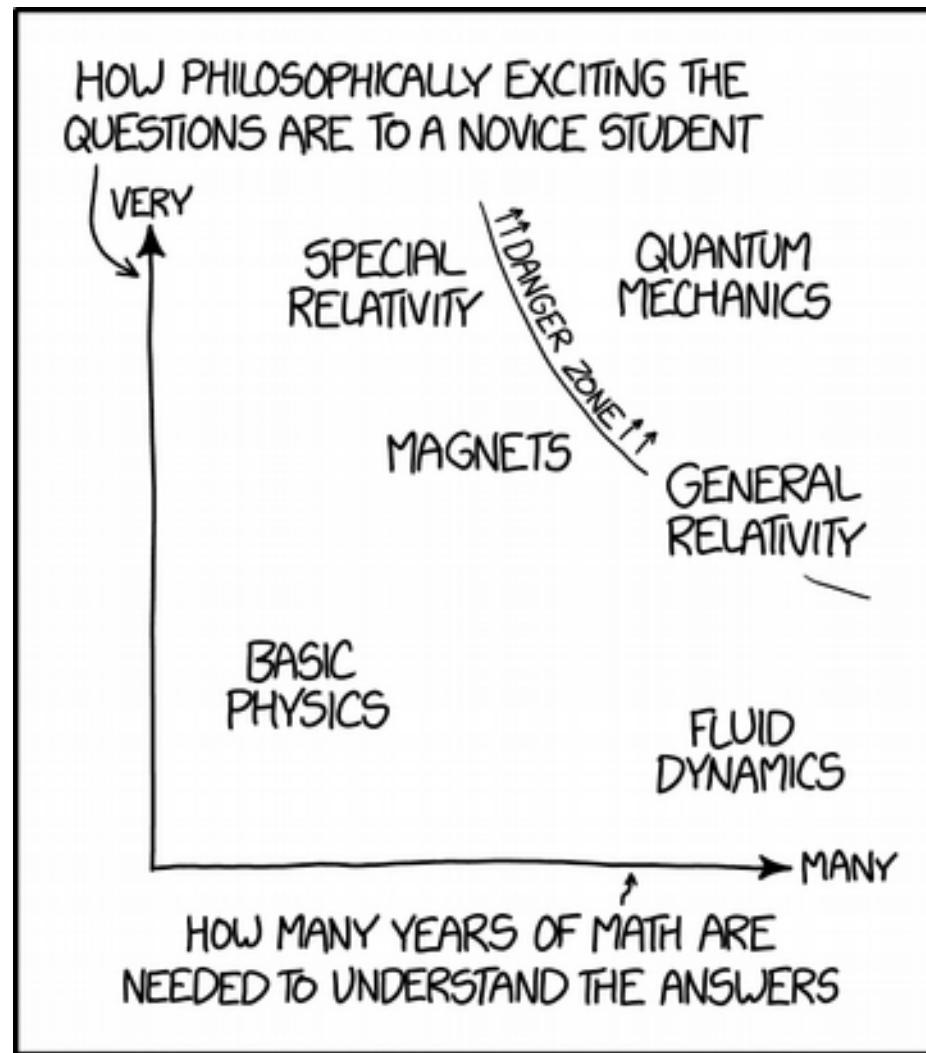
When does quantum start?

- Where does the beachtstart/end?



- Ultra-cold
- Very-high Vacuum
- Ultra-short





WHY SO MANY PEOPLE HAVE WEIRD IDEAS ABOUT QUANTUM MECHANICS

Qubit

- **Binary equivalent of bit**

$$0 \rightarrow |0\rangle$$

$$|0\rangle, |1\rangle \in \mathcal{H}$$

$$1 \rightarrow |1\rangle$$

- **All superposition also exist**

$$|q\rangle = \alpha |0\rangle + \beta |1\rangle \quad \alpha, \beta \in \mathbb{C} \quad \alpha^2 + \beta^2 = 1$$

- **Probability to obtain 0 or 1**

$$p(0) = \alpha^2$$

$$p(1) = \beta^2$$

- **After measurement**

$$|q\rangle \rightarrow |0\rangle$$

$$|q\rangle \rightarrow |1\rangle$$

Qubits

- **Two qubits:**

$$|\psi\rangle = \alpha_1 |00\rangle + \alpha_2 |01\rangle + \alpha_3 |10\rangle + \alpha_4 |11\rangle$$

- **Many qubits:**

$$|\psi\rangle = \alpha_1 |0\dots00\rangle + \alpha_2 |0\dots01\rangle + \dots + \alpha_{2^n} |1\dots11\rangle$$



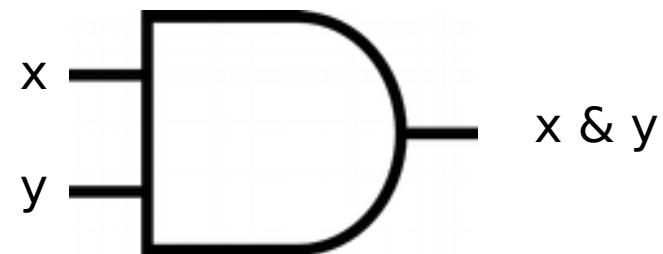
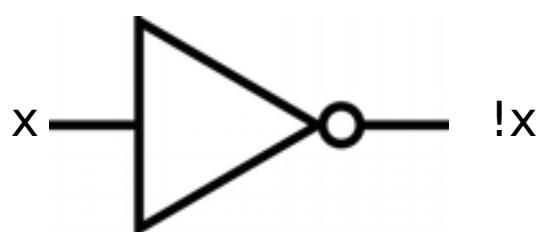
n qubits = 2^n complex amplitudes α_i

- **Superposition → Quantum parallelism**
- **Normalization, but otherwise free → entanglement (non-classical correlations)**

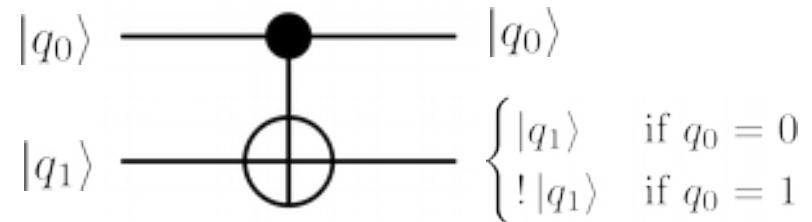
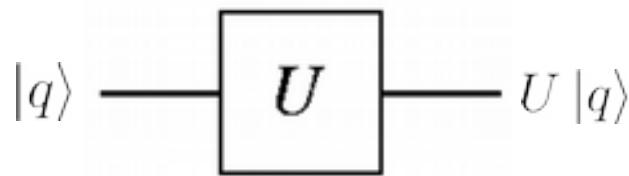
$$\sum_{i=1}^n |\alpha_i|^2 = 1$$

Universal quantum computer - universal gate algebra

- Not
- And (2 bits operation)



- Single qubit operation (unitary)
- 2 qubits gate CNOT (unitary)



Convenient representation

- Qubit as complex unit vector

$$|0\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \in \mathbb{C}^2 \quad |1\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \in \mathbb{C}^2 \quad |\psi\rangle = \frac{1}{\sqrt{2}}(1|0\rangle + i|1\rangle) = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ i \end{pmatrix}$$

- Unitary Gate as Hermitian matrix

- Not $X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$

- Phase not $Y = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix}$

$$X|0\rangle = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} = |1\rangle$$

$$Y|0\rangle = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ i \end{pmatrix} = i|1\rangle$$

$$X|1\rangle = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = |0\rangle$$

$$Y|1\rangle = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -i \\ 0 \end{pmatrix} = -i|0\rangle$$

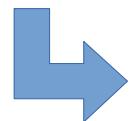
Hadamard Gate

- **One Qubit**

$$H = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \quad H|0\rangle = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = |0\rangle \quad H|1\rangle = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ -1 \end{pmatrix} = -|1\rangle$$

- **Two qubits**

$$H_2 = \frac{1}{\sqrt{2}} H \otimes \mathbb{I}_{2 \times 2} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{pmatrix} \quad |\psi\rangle = \alpha_1|00\rangle + \alpha_2|01\rangle + \alpha_3|10\rangle + \alpha_4|11\rangle = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \alpha_4 \end{pmatrix}$$



$$\begin{aligned} H|\psi\rangle &= \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \alpha_4 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} \alpha_1 + \alpha_3 \\ \alpha_2 + \alpha_4 \\ \alpha_1 - \alpha_3 \\ \alpha_2 - \alpha_4 \end{pmatrix} \\ &= (\alpha_1 + \alpha_2)|00\rangle + (\alpha_2 + \alpha_4)|01\rangle + (\alpha_1 - \alpha_3)|10\rangle + (\alpha_2 - \alpha_4)|11\rangle \end{aligned}$$

Hadamard Gate

- **n qubits: apply Hadamard to first one only**

$$|\psi\rangle = \alpha_1 |0\dots00\rangle + \alpha_2 |0\dots01\rangle + \dots + \alpha_N |1\dots11\rangle \quad N = 2^n$$

$$H = \frac{1}{\sqrt{2}} H_{2 \times 2} \otimes \mathbb{I}_{2 \times 2} \dots \mathbb{I}_{2 \times 2} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 & \dots & 1 & 0 & \dots \\ 0 & 1 & \dots & 0 & 1 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 1 & 0 & \dots & -1 & 0 & \dots \\ 0 & 1 & \dots & 0 & -1 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} \mathbb{I} & \mathbb{I} \\ \mathbb{I} & -\mathbb{I} \end{pmatrix} \quad \mathbb{I} : \frac{N}{2} \times \frac{N}{2}$$

$$\begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_{\frac{N}{2}} \\ \alpha_{\frac{N}{2}+1} \\ \vdots \\ \alpha_N \end{pmatrix} \rightarrow \frac{1}{\sqrt{2}} \begin{pmatrix} \alpha_1 + \alpha_{\frac{N}{2}+1} \\ \vdots \\ \alpha_{\frac{N}{2}} + \alpha_N \\ \alpha_1 - \alpha_{\frac{N}{2}+1} \\ \vdots \\ \alpha_{\frac{N}{2}} - \alpha_N \end{pmatrix}$$

Measurement vs Evolution (gate)

- Measure first qubit

$$|\psi\rangle = \alpha_1 |00\rangle + \alpha_2 |01\rangle + \alpha_3 |10\rangle + \alpha_4 |11\rangle$$

$$p(q_0 = 0) = \alpha_1^2 + \alpha_2^2$$

$$p(q_0 = 1) = \alpha_3^2 + \alpha_4^2$$

→ $|\psi\rangle = \frac{1}{\sqrt{\alpha_1^2 + \alpha_2^2}} (\alpha_1 |00\rangle + \alpha_2 |01\rangle)$

$$\alpha_3 = 0, \alpha_4 = 0$$

$$|\psi\rangle = \frac{1}{\sqrt{\alpha_3^2 + \alpha_4^2}} (\alpha_3 |10\rangle + \alpha_4 |11\rangle)$$

$$\alpha_1 = 0, \alpha_2 = 0$$

Non-Linear!

- Gate (quantum evolution): matrix multiplication

$$X |0\rangle = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} = |1\rangle$$

Linear!

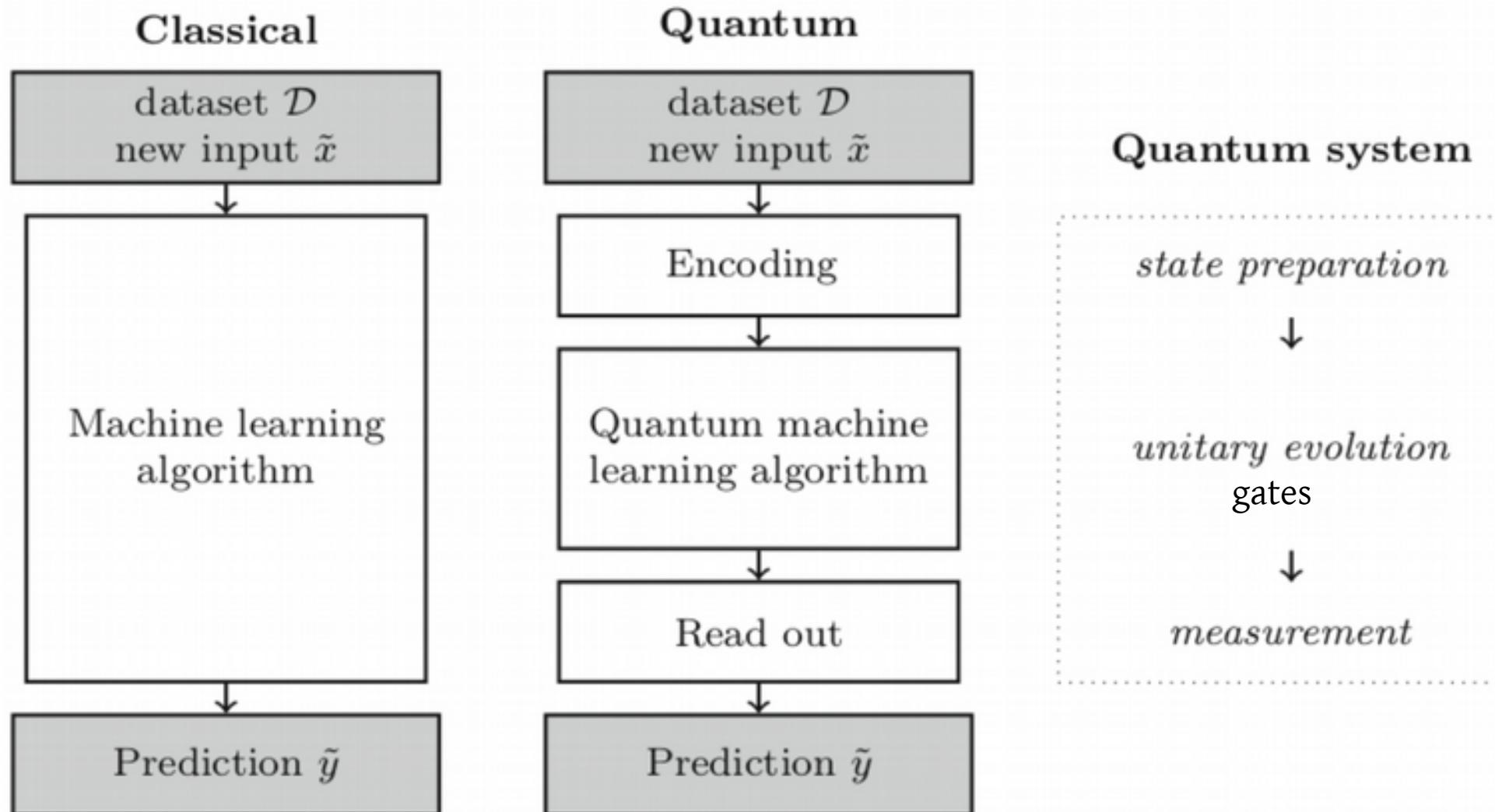
Key ingredients

- **Complex wavefunction**
 - Interferences
 - Entanglement
 - Superposition → large space of possibilities
- **Linear evolution equation**
- **Measurements**
 - Destroys superpositions
 - Only probabilistic predictions
- **Constraints**
 - No cloning theorem
 - One Qubit = One bit of information



It was surprisingly easy to get \$100 million from NASA.

QML design



Thanks to Maria Schuld from Xanadu

Quantum (ML) algorithm - example

- **Square-distance classifier**
- **Titanic dataset**



Thanks to Maria Schuld from Xanadu

Quantum (ML) algorithm - example

- **Square-distance classifier**
- **Titanic dataset**

raw data			
	price	cabin	survival
Passenger 1	8,500	0910	1 (yes)
Passenger 2	1,200	2105	0 (no)
Passenger 3	7,800	1121	?

Quantum (ML) algorithm - example

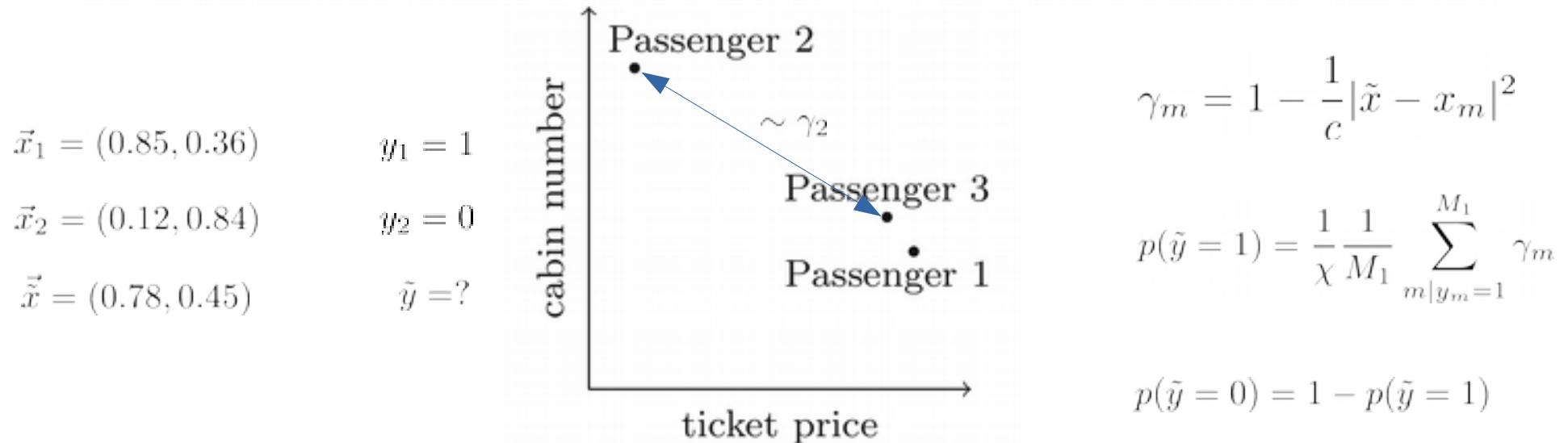
- **Square-distance classifier**
- **Titanic dataset**

	raw data		preprocessed data		
	price	cabin	price	cabin	survival
Passenger 1	8,500	0910	0.85	0.36	1 (yes)
Passenger 2	1,200	2105	0.12	0.84	0 (no)
Passenger 3	7,800	1121	0.78	0.45	?

 \vec{x} y

Classical square distance classifier

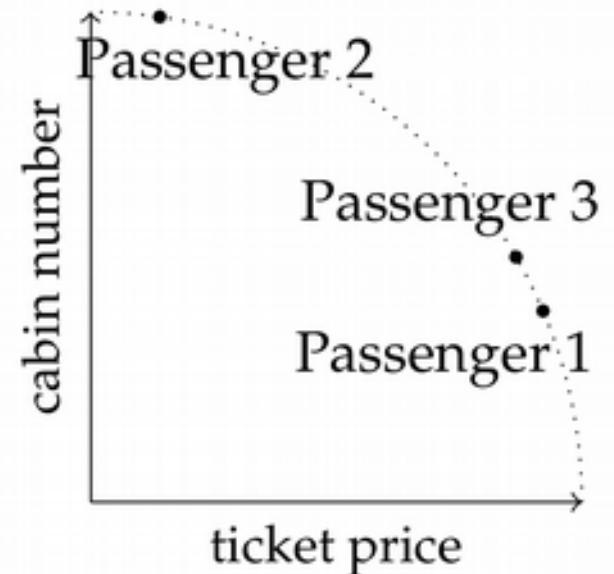
	raw data		preprocessed data		
	price	cabin	price	cabin	survival
Passenger 1	8,500	0910	0.85	0.36	1 (yes)
Passenger 2	1,200	2105	0.12	0.84	0 (no)
Passenger 3	7,800	1121	0.78	0.45	?



Quantum square-distance classifier

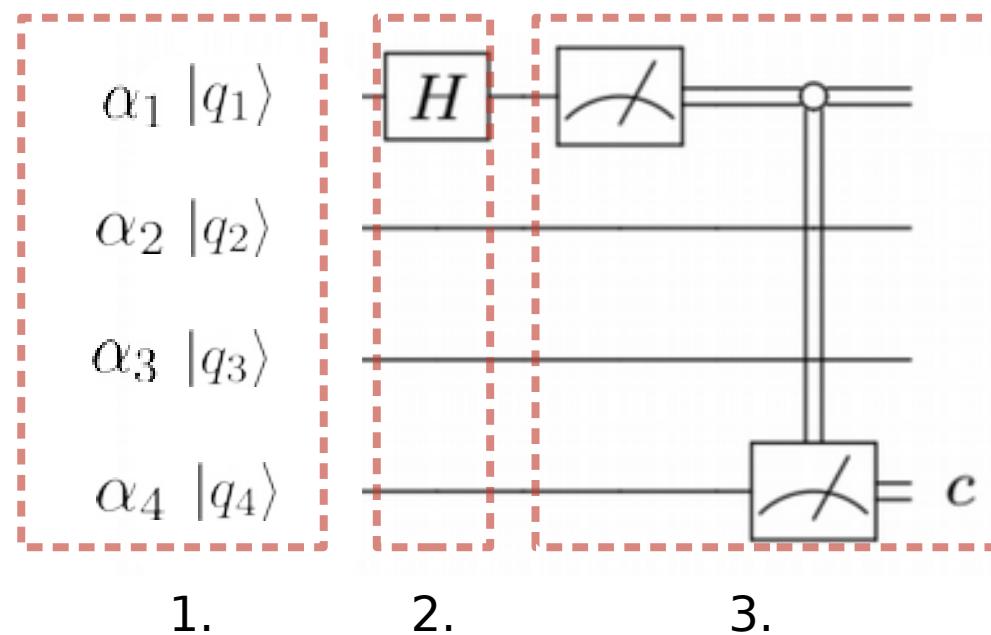
- Data preprocessing normalized vectors (0-1)

	price	room	survival
Passenger 1	0.921	0.390	yes (1)
Passenger 2	0.141	0.990	no (0)
Passenger 3	0.866	0.500	?



Algorithm

- (1. Amplitude encoding on four qubits)
- 2. Apply one Hadamard to the first qubit
- 3. Measure first qubit
 - If 0, measure qubit fourth qubit
 - If 1, discard
- Repeat to obtain statistics



Algorithm results

				Step 1
0	0	0	0	0
0	0	0	1	$\frac{1}{\sqrt{4}} 0.921$
0	0	1	0	0
0	0	1	1	$\frac{1}{\sqrt{4}} 0.390$
0	1	0	0	$\frac{1}{\sqrt{4}} 0.141$
0	1	0	1	0
0	1	1	0	$\frac{1}{\sqrt{4}} 0.990$
0	1	1	1	0
1	0	0	0	0
1	0	0	1	$\frac{1}{\sqrt{4}} 0.866$
1	0	1	0	0
1	0	1	1	$\frac{1}{\sqrt{4}} 0.500$
1	1	0	0	$\frac{1}{\sqrt{4}} 0.866$
1	1	0	1	0
1	1	1	0	$\frac{1}{\sqrt{4}} 0.500$
1	1	1	1	0

	price	room	survival
Passenger 1	0.921	0.390	yes (1)
Passenger 2	0.141	0.990	no (0)
Passenger 3	0.866	0.500	?

$$|\psi\rangle = \frac{1}{\sqrt{4}} (0.921 |0001\rangle + 0.390 |0011\rangle + 0.141 |0100\rangle + 0.990 |0110\rangle + 0.866 |1001\rangle + 0.500 |1011\rangle + 0.866 |1100\rangle + 0.500 |1110\rangle)$$

- (1. Amplitude encoding) 2. Apply one Hadamard to the first qubit
3. Measure first qubit**

Thanks to Maria Shuld from Xanadu

Algorithm results

				Step 1	Step 2
0	0	0	0	0	0
0	0	0	1	$\frac{1}{\sqrt{4}} 0.921$	$\frac{1}{\sqrt{4}} (0.921 + 0.866)$
0	0	1	0	0	0
0	0	1	1	$\frac{1}{\sqrt{4}} 0.390$	$\frac{1}{\sqrt{4}} (0.390 + 0.500)$
0	1	0	0	$\frac{1}{\sqrt{4}} 0.141$	$\frac{1}{\sqrt{4}} (0.141 + 0.866)$
0	1	0	1	0	0
0	1	1	0	$\frac{1}{\sqrt{4}} 0.990$	$\frac{1}{\sqrt{4}} (0.990 + 0.500)$
0	1	1	1	0	0
1	0	0	0	0	0
1	0	0	1	$\frac{1}{\sqrt{4}} 0.866$	$\frac{1}{\sqrt{4}} (0.921 - 0.866)$
1	0	1	0	0	0
1	0	1	1	$\frac{1}{\sqrt{4}} 0.500$	$\frac{1}{\sqrt{4}} (0.390 - 0.500)$
1	1	0	0	$\frac{1}{\sqrt{4}} 0.866$	$\frac{1}{\sqrt{4}} (0.141 - 0.866)$
1	1	0	1	0	0
1	1	1	0	$\frac{1}{\sqrt{4}} 0.500$	$\frac{1}{\sqrt{4}} (0.990 - 0.500)$
1	1	1	1	0	0

- (1. Amplitude encoding) 2. Apply one Hadamard to the first qubit
3. Measure first qubit**

Thanks to Maria Schuld from Xanadu

Algorithm results

				Step 1	Step 2	Step 3
0	0	0	0	0	0	0
0	0	0	1	$\frac{1}{\sqrt{4}} 0.921$	$\frac{1}{\sqrt{4}} (0.921 + 0.866)$	$\frac{1}{\sqrt{4}\chi} (0.921 + 0.866)$
0	0	1	0	0	0	0
0	0	1	1	$\frac{1}{\sqrt{4}} 0.390$	$\frac{1}{\sqrt{4}} (0.390 + 0.500)$	$\frac{1}{\sqrt{4}\chi} (0.390 + 0.500)$
0	1	0	0	$\frac{1}{\sqrt{4}} 0.141$	$\frac{1}{\sqrt{4}} (0.141 + 0.866)$	$\frac{1}{\sqrt{4}\chi} (0.141 + 0.866)$
0	1	0	1	0	0	0
0	1	1	0	$\frac{1}{\sqrt{4}} 0.990$	$\frac{1}{\sqrt{4}} (0.990 + 0.500)$	$\frac{1}{\sqrt{4}\chi} (0.990 + 0.500)$
0	1	1	1	0	0	0
1	0	0	0	0	0	0
1	0	0	1	$\frac{1}{\sqrt{4}} 0.866$	$\frac{1}{\sqrt{4}} (0.921 - 0.866)$	0
1	0	1	0	0	0	0
1	0	1	1	$\frac{1}{\sqrt{4}} 0.500$	$\frac{1}{\sqrt{4}} (0.390 - 0.500)$	0
1	1	0	0	$\frac{1}{\sqrt{4}} 0.866$	$\frac{1}{\sqrt{4}} (0.141 - 0.866)$	0
1	1	0	1	0	0	0
1	1	1	0	$\frac{1}{\sqrt{4}} 0.500$	$\frac{1}{\sqrt{4}} (0.990 - 0.500)$	0
1	1	1	1	0	0	0

- (1. Amplitude encoding) 2. Apply one Hadamard to the first qubit
 3. Measure first qubit

Thanks to Maria Schuld from Xanadu

Result

				Step 1	Step 2	Step 3
0	0	0	0	0	0	0
0	0	0	1	$\frac{1}{\sqrt{4}} 0.921$	$\frac{1}{\sqrt{4}} (0.921 + 0.866)$	$\frac{1}{\sqrt{4}\chi} (0.921 + 0.866)$
0	0	1	0	0	0	0
0	0	1	1	$\frac{1}{\sqrt{4}} 0.390$	$\frac{1}{\sqrt{4}} (0.390 + 0.500)$	$\frac{1}{\sqrt{4}\chi} (0.390 + 0.500)$
0	1	0	0	$\frac{1}{\sqrt{4}} 0.141$	$\frac{1}{\sqrt{4}} (0.141 + 0.866)$	$\frac{1}{\sqrt{4}\chi} (0.141 + 0.866)$
0	1	0	1	0	0	0
0	1	1	0	$\frac{1}{\sqrt{4}} 0.990$	$\frac{1}{\sqrt{4}} (0.990 + 0.500)$	$\frac{1}{\sqrt{4}\chi} (0.990 + 0.500)$
0	1	1	1	0	0	0

Probability to measure the last qubit in state 0

$$p(q_4 = 0) = \frac{1}{4\chi} (|0.141 + 0.866|^2 + |0.990 + 0.500|^2) \approx 0.448$$

This is the same as our goal

$$p(q_4 = 0) = \frac{1}{\chi} \left(1 - \frac{1}{4} |0.141 - 0.866|^2 + |0.990 - 0.500|^2 \right) = \frac{1}{\chi} \frac{1}{M_1} \sum_{m|y_m=0}^{M_0} \gamma_m \approx 0.448$$

Result

Passenger 3 survived!



Insights

- **Only one Hadamard and two simple measurements!**
- **Remains the same for any number of points**
- **But:**
 - Data encoding is a crucial step (often the runtime bottleneck)
 - Constraints on the preprocessing (unit length vectors)
 - Result is a probabilistic measurement
 - Not actually faster than classical one
- **Remark:**
 - Use of interferences
 - QM often inspired by particular variations of ML models which suite quantum computing

Python code

```
import numpy as np
from qiskit import QuantumCircuit, ClassicalRegister,
QuantumRegister
from qiskit import execute
import math
import numpy
#
# State and circuit creation
#
# Create a 4 qubit quantum register
q = QuantumRegister(4)
# Create a classical register for measurements
c = ClassicalRegister(2)
# Create a quantum circuit with input q & output c
circ = QuantumCircuit(q,c)
# Initialize the register content
v = numpy.zeros(16)
v[ 1] = 0.920843009649141/2 # |0001>
v[ 3] = 0.389933522001265/2 # |0011>
v[ 4] = 0.141001339519088/2 # |0100>
v[ 6] = 0.990009405134023/2 # |0110>
v[ 9] = 0.866019052628739/2 # |1001>
v[11] = 0.500011000363013/2 # |1011>
v[12] = 0.866019052628739/2 # |1100>
v[14] = 0.500011000363013/2 # |1110>
circ.initialize(v, q)
# Add a H gate on qubit 3 (leftmost)
circ.h(q[3])
# Measure leftmost qubit
circ.measure(q[3], c[1])
# Measure rightmost qubit
circ.measure(q[0], c[0])
```

```
# Running on IBM Q
#
from qiskit import IBMQ
# Load credentials
IBMQ.load_accounts()
# Choose device
from qiskit.providers.ibmq import least_busy
large_enough_devices = IBMQ.backends(filters=lambda x:
x.configuration().n_qubits > 3 and not x.configuration().simulator)
backend = least_busy(large_enough_devices)
print("The best backend is " + backend.name())
# Execute the circuit
from qiskit.tools.monitor import job_monitor
# Number of shots to run the program
shots = 1024
# Maximum number of credits to spend on executions
max_credits = 3
job = execute(circ, backend=backend, shots=shots,
max_credits=max_credits)
job_monitor(job)
# Wait until the job has finished
result = job.result()
#
# Print the results
#
counts = result.get_counts(circ)
# Calculate probability of label 1
# q4=0 & q0=1 over q4=0 & (q0=0 or q0=1)
p1 = counts.get('01',0) / (counts.get('00',0)+counts.get('01'))
print("Probability of label 0 (1) is %.3f (%.3f) % ((1-p1),p1))
```

Code does not work as is. New Qiskit convention for connection to the IBM devices. See jupyter notebook.

Conclusion

Thank You!

