

Quantum Computing

How can it save the world?

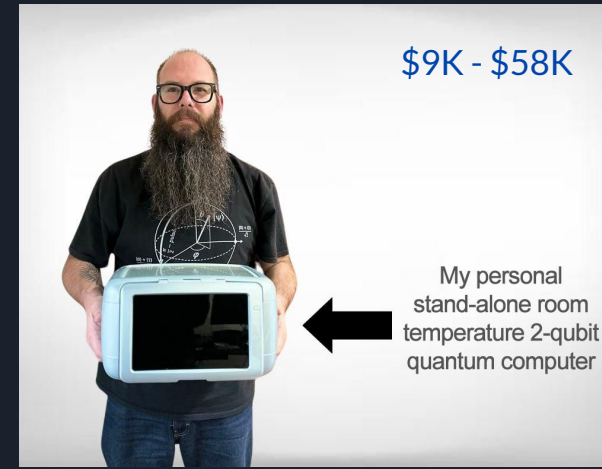
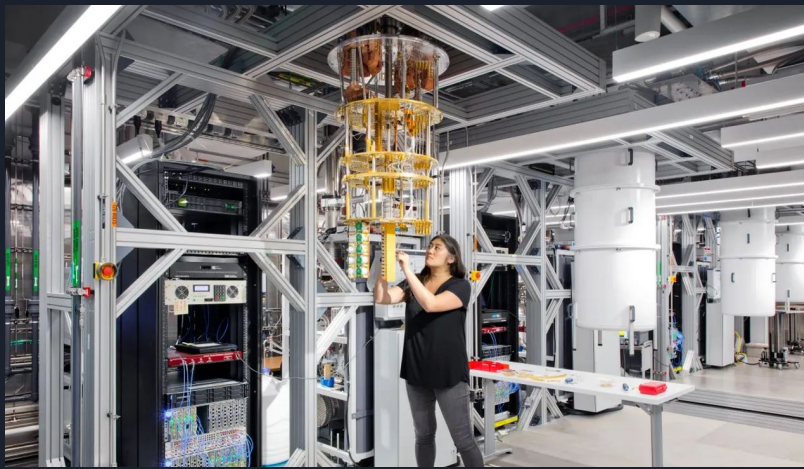
Debajyoti Bera
Department of Computer Sc. & Engg.
Center for Quantum Technologies
IIIT-Delhi



INDRAPRASTHA INSTITUTE of
INFORMATION TECHNOLOGY DELHI

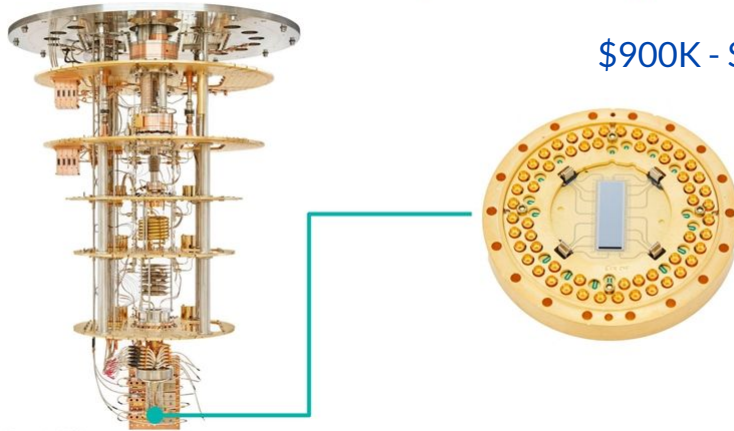
Modern QPU

Company	Technology of qubits	Architecture
IBM	Superconducting	Gate-based
Google	Superconducting	
Intel	Superconducting & semiconductor spin	
IonQ	Trapped ion	
Quantinuum	Trapped ion	
Rigetti	Superconducting transmon	
SpinQ	Nuclear magnetic resonance (NMR)	
DWAVE	Superconducting	Quantum annealing



The Chip is the Heart of the Quantum Computer

\$900K - \$2M



How to access QPUs ?



Amazon Braket Pricing

Hardware Provider	D- Wave	D- Wave	IonQ	Rigetti
QPU Family	2000Q	Advantage	IonQ device	Aspen-8
Per-task price	\$0.30000	\$0.30000	\$0.30000	\$0.30000
Per-shot Price	\$0.00019	\$0.00019	\$0.01000	\$0.00035

WORKFALL



PennyLane supports IBM Quantum, Amazon Braket, IonQ, AQT, Rigetti, QuTech, Honeywell.

How to program QPUs ?

```
import pennylane as qml
```


```
# Create backend to run the quantum node  
dev = qml.device('default.qubit', wires=1)
```

```
# Define quantum node  
@qml.qnode(dev)  
def circuit(var):  
    qml.RX(var, wires=0)  
    return qml.expval.PauliZ(0)
```

```
# Define classical node/cost  
def cost(var):  
    return var[0] * circuit(var[1])
```

```
# Update initial variable  
opt = qml.GradientDescentOptimizer()  
var_new = opt.step(cost, var=[0.1, 0.2])
```

quantum device

$|0\rangle$ — $R_x(v)$ — 

quantum node

classical node

optimizer

Hybrid loop

Input

Training set
Cost function
Ansatz

Quantum computer

ρ_k

$f_k(\theta, \rho_k)$

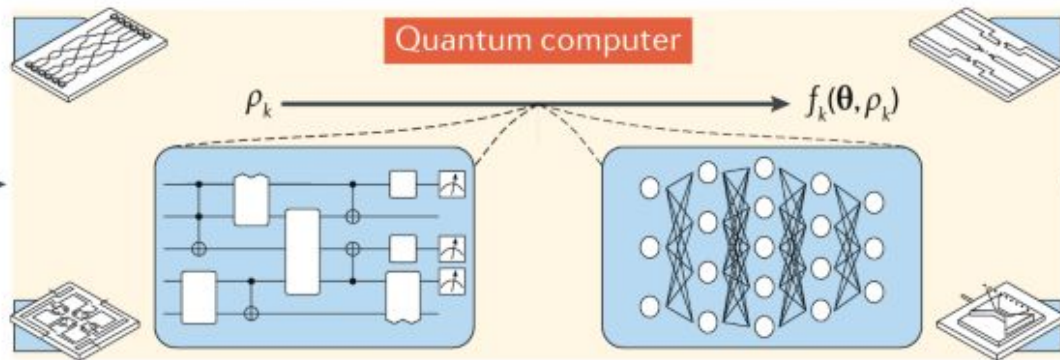
Updated parameters

Classical computer

Optimizer
 $\arg \min_{\theta} C(\theta)$

$$C(\theta) = \sum_k f_k(\theta, \rho_k)$$

Variational
Quantum
Algorithms



Quantum development ecosystem

Many programming languages.

Lot of libraries.

Active community.

Several tutorial, blogs, open Github projects,

SDK

Qiskit by IBM

Ocean by D-Wave

Strawberry Fields & PennyLane

Forest by Rigetti

Circ by Google

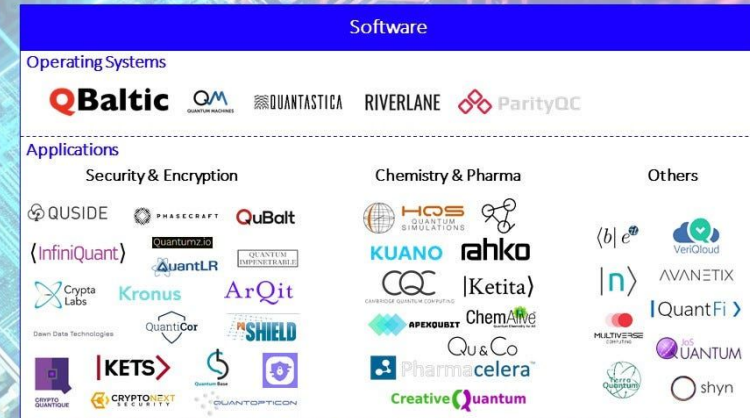
QDK by Microsoft

CUDA-Quantum by NVIDIA


THE EUROPEAN QUANTUM COMPUTING STARTUP LANDSCAPE



UVC UNTERNEHMERTUM VENTURE CAPITAL



@alex_kiltz



Quantum Algorithms for ...

Traditional CS problems

Quantum Algorithm Zoo

Number theoretic algorithms

Linear algebraic algorithms

Algebraic algorithms

Sorting, searching, ...

Graph algorithms

String algorithms

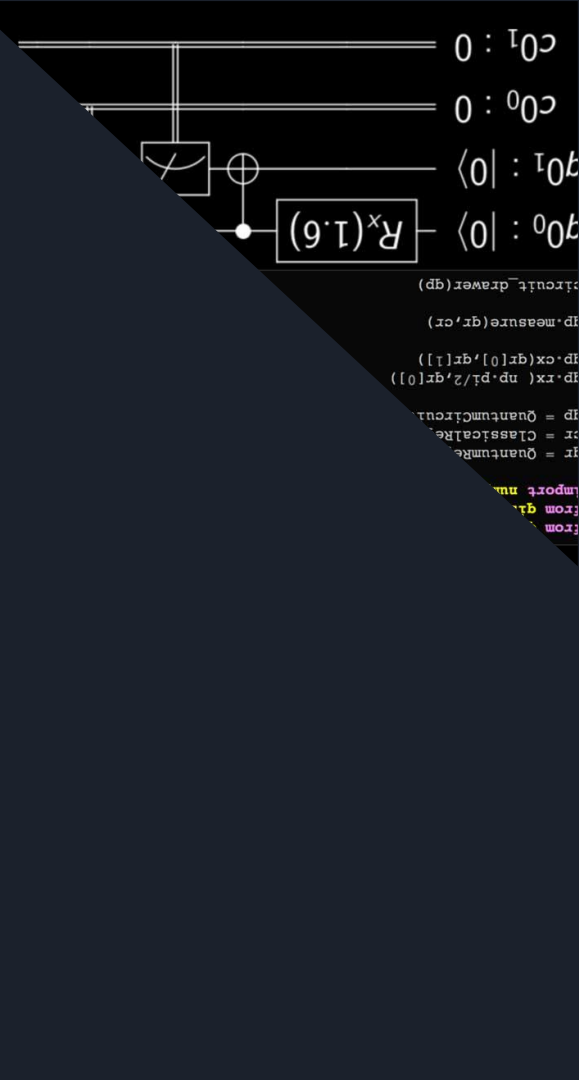
Group-theoretic algorithms

Linear programs

Semi-definite programs

...

<https://quantumalgorithmzoo.org/>

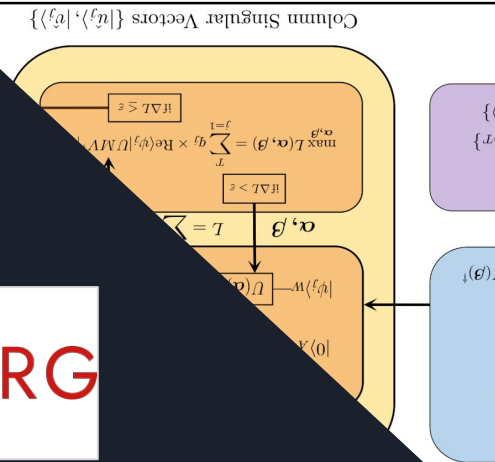


Quantum algorithms for ...

Data analysis



- Singular-value estimation based algorithms
- Monte-carlo techniques
- PCA and other dimensionality reduction methods
- Clustering algorithms like k-means, k-median, ...
- Matrix operations like inversion, solving linear system, ...



Quantum algorithms for ...

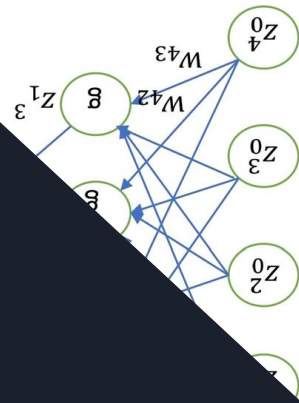
Machine learning & optimization

https://en.wikipedia.org/wiki/Quantum_machine_learning

- Quantum-enhanced reinforcement learning
- Quantum annealing
- Training Boltzmann machines
- Quantum convolutional neural networks (QCNN)
- Dissipative QNN
- Quantum generative adversarial networks (QGAN)
- Hidden quantum markov models
- Quantum graph neural networks
- Quantum physics-inspired neural networks (QPINN)

		Type of Algorithm	
		classical	quantum
Type of Data	classical	CC	CQ
	quantum	QC	QQ

$$(\sum_{i=1}^n \beta_i) \mathbf{z} = \mathbf{z}$$



Quantum Online Portfolio Optimization

Input: n, s, η, T, C, δ

1: Initialize $U_{\rho(0)} = \mathbb{I}$, $\tilde{I} = 1$ and $q^{(1)} = (1, \dots, 1)$, $U_{q^{(1)}} = \mathbb{I}$.

2: **for** $t = 1$ to T **do**

3: $q_{\max} \leftarrow$ Find the largest element of $q^{(t)}$ using $U_{q^{(t)}}$ and quantum maximum finding [56] with success probability $1 - \frac{\delta}{4T}$.

4: $\tilde{Z}^{(t)} \leftarrow$ Estimate the norm of $\frac{q^{(t)}}{q_{\max}}$ using $U_{q^{(t)}}$, q_{\max} , Lemma 2, and Lemma 3(i), with relative error $\epsilon_Z = \frac{\eta^2}{r_{\min}^2}$ and success probability $1 - \frac{\delta}{4T}$.

5: $U_{w^{(t)}} \leftarrow$ Prepare quantum circuit for approximating $|\tilde{w}^{(t)}\rangle$ of the quantum state $|w^{(t)}\rangle$, where $w^{(t)} = \frac{q^{(t)}}{\|q^{(t)}\|_1}$, using $U_{q^{(t)}}$, q_{\max} , $\tilde{Z}^{(t)}$, Lemma 2, Lemma 3(ii), with success probability $1 - \frac{\delta}{4T}$.

6: $\Gamma, W, V \leftarrow$ Determine using Lem. 5 applied to $\tilde{w}^{(t)}$ with probability $1 - \frac{\delta}{4T}$.

7: $i_1^{(t)}, \dots, i_s^{(t)}$ Perform multi-sampling using Γ, W, V and Lemma 6 with probability $1 - \frac{\delta}{4T}$.

8: Invest the amount $1/s$ in each asset $i_1^{(t)}, \dots, i_s^{(t)}$ at cost C each.

9: Wait until end of day.

10: Receive price relative oracle $U_{\rho^{(t)}}$.

11: $\rho_{j^{(t)}}^{(t)} \leftarrow$ Query $U_{\rho^{(t)}}$ with $|j^{(t)}\rangle|0\rangle$.

12: $\tilde{I}^{(t)} \leftarrow$ Estimate $\tilde{w}^{(t)} \cdot \rho^{(t)}$ using $U_{w^{(t)}}$, $U_{\rho^{(t)}}$, and Lemma 4, with relative error success probability $1 - \frac{\delta}{4T}$.

13: $U_{q^{t+1}} \leftarrow$ Prepare quantum circuit to compute $q^{(t+1)} = \exp\left(\eta \sum_{t'=1}^t \frac{\rho^{(t')}}{\tilde{I}^{(t')}}\right)$ using

Lemma 2.

14: **end for**

Output: $LS_{\text{samp}}^Q := \frac{1}{T} \sum_{t=1}^T \log \left(\frac{1}{s} \sum_{\ell=1}^s \rho_{i_{\ell}^{(t)}}^{(t)} \right).$

Name	Alg.	Regret	Run time
Online	1	$\frac{1}{r_{\min}} \sqrt{\frac{\log n}{2T}}$	$O(Tn)$
Sampling-based Online	2	$\frac{2}{r_{\min}} \sqrt{\frac{\log n}{2T}}$	$O(T^2 n \log \frac{T}{\delta})$
Approximate Sampling-based Online	3	$\frac{8}{r_{\min}} \sqrt{\frac{\log n}{2T}}$	$O\left(Tn + \frac{T^2}{r_{\min}} \log \frac{T}{\delta}\right)$
Quantum Online	4	$\frac{12}{r_{\min}} \sqrt{\frac{\log n}{2T}}$	$\tilde{O}\left(T^3 \sqrt{\frac{n}{r_{\min}}} \log^{1.5}\left(\frac{1}{\delta}\right)\right)$



Why quantum ?

Fundamentally different point of view! Backed by quantum mechanics.

10 qubits represent a distribution over 1024 elements \equiv 1024-sized “stochastic” vector

+

Distribution allows “negative” probabilities \Rightarrow Cancellation of probabilistic scenarios

+

Certain impossibility results advocated by quantum mechanics

Classically impossible tasks are now possible.

Better speedup compared to classical techniques.

Better quality of optimization solutions.

Why not (yet) quantum ?

No Quantum RAM (yet)!

Lots of errors - T1, T2, Gate, Readout, ...

Barren plateau observed in variational QA

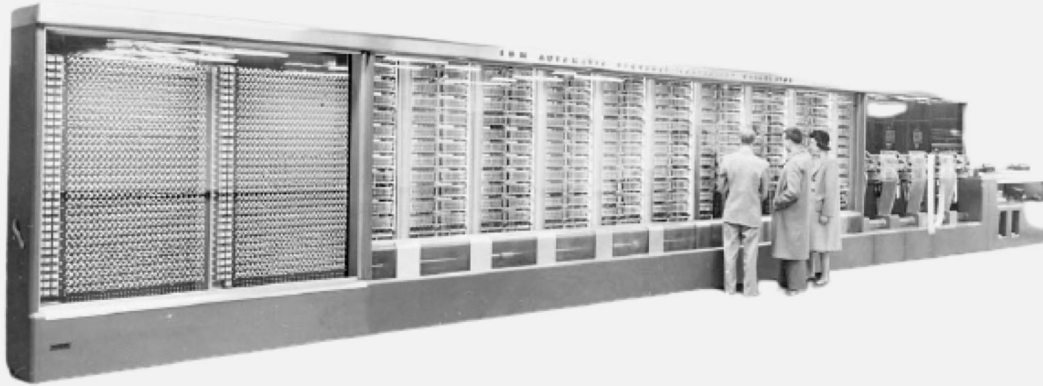
No clear case of quantum supremacy!

Perceived learning curve.



First generation CPU

Harvard Mark I



Specification

- + : 0.3 sec
- x : 6 secs
- ÷ : 15 secs
- sin, cos, etc. : 1 min +
- Operated on 72 73-bit registers
- 51' x 8' x 2'
- 4.3 ton



Thank you!

<https://braqiiit.github.io/>

