



Quantum Computing Simplified

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Random Bits



Classical randomized algorithms use classical coins (random bits).

Classical randomized algorithms are efficient

Quicksort, Monte-Carlo sampling, ...

Satisfiability of 3SAT Boolean formula

Brute force O(2ⁿ) Deterministic O(1.439ⁿ) Randomized O(1.321ⁿ)

From Bits to Qubits

Photon polarization, Electron spin, direction of current in Josephson junction, ...



Organization for this lecture

- 1. (Recap) The basic principles of quantum computing
 - a. Qubits
 - b. Operations
- 2. Designing quantum algorithms
- 3. Emerging techniques

1 qubit



Basic datatype. Can be observed to be in state-0 and state-1.

Behaviour of a random bit



```
b = random(0,1)
// Q: what is b?
```

```
if b=0, print("0")
```

- 3.
- if b=1, print("1") 4.

If 0 is printed, b must have been 0. If 1 is printed, b must have been 1.







1 qubit

Basic datatype. Can be observed to be in state-0 and state-1.

Behaviour of a qubit

<u>Exercise</u>

Determine the state |b> by only observing the output of the code.



AND







2-qubit state can be mathematically represented as a complex combination of four basis states



Classical randomized algorithms use random variables

Stochastic vector

b = random bit from { $0:\frac{1}{2}$, $1:\frac{1}{2}$ } Mathematical representation of b $\begin{bmatrix} 1/2\\1/2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1\\0 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 0\\1 \end{bmatrix} \text{ represents b=1}$ Another representation of b $\begin{bmatrix} 1/2\\1/2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1/3\\2/3 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 2/3\\1/3 \end{bmatrix} \text{ not clear what can be be done with alternative representations using a different basis}$

Algorithms using random bits can be analysed using L1 norm unit vectors over R



Algorithms using qubits can be analysed using L2 norm unit vectors over C

"Value" of a qubit (state vector)

Observation/measurement changes the state of a qubit !

Linear algebraically, measurement is a projection onto a set of basis states.



$$|\psi\rangle = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \frac{1+i}{2} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} + \frac{1-i}{2} \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$
Measuring in 0/1 basis
Observe |0⟩ with probability $\left|\frac{1}{\sqrt{2}}\right|^2$
Measuring in +/- basis
Observe |1⟩ with probability $\left|\frac{1}{\sqrt{2}}\right|^2$

qubit

Has intrinsic state.

State is a continuum from 2-dimensional numbers. Observation reveals partial information. Observation changes (collapses) states.

Single bit operation

Classical deterministic operations



Classical randomized operations

If x=0, f(x) =	value	0	1
	prob.	1/2	1/2
If x=1, f(x) =	value	0	1
	prob.	1/3	2/3

$$\begin{bmatrix} f(x) \end{bmatrix} = \begin{bmatrix} 1/2 & 1/3 \\ 1/2 & 2/3 \end{bmatrix} \begin{bmatrix} x \end{bmatrix}$$

Multiplication by a stochastic matrix

Qubit operation

Linear operation

Multiplication of state vector by a unitary (L2 length-preserving complex) matrix



Qubit operation

Linear operation

Multiplication of state vector by a unitary (L2 length-preserving complex) matrix



Qubit operation

Linear operation

Multiplication of state vector by a unitary (L2 length-preserving complex) matrix



First magic



 $|+\rangle = \sqrt{(1/2)} |0\rangle + \sqrt{(1/2)} |1\rangle$

- Observe $|0\rangle$ with probability $\frac{1}{2}$
- Observe $|1\rangle$ with probability $1\!\!\!/_2$

|+⟩ is a **superposition** of |0⟩ and |1⟩

$\left| + \right\rangle$ is not $\left| 0 \right\rangle$ OR $\left| 1 \right\rangle$

Suppose $|+\rangle$ is a state that is randomly chosen between $|0\rangle$ and $|1\rangle$ with equal probability.

- $|0\rangle$ with prob. $\frac{1}{2}$
 - After 2nd H is applied...
 - $\circ \quad |0\rangle \rightarrow \sqrt{(1_2)} \, |0\rangle \ + \ \sqrt{(1_2)} \ |1\rangle$
 - Observation will yield
 - $\circ~~|0\rangle$ and $|1\rangle$ with probability $^{1\!\!/_2}$
- $|1\rangle$ with prob. $\frac{1}{2}$
 - After 2nd H is applied...
 - $\circ \quad |1\rangle \rightarrow \sqrt{(1_2')} \; |0\rangle \; \text{--} \; \sqrt{(1_2')} \; \; |1\rangle$
 - Observation will yield
 - $\circ ~~ |0\rangle$ and $|1\rangle$ with probability $^{1\!\!/_2}$
- Overall ...
 - Prob. of observing $|0\rangle = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$
 - Prob. of observing $|1\rangle = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$

state evolution operation

Specify action only on basis states Linearly extrapolate on all other states Unitary, hence reversible

Qubits cannot be copied!

```
is the code that does the bubble sort.
for (int i = ar.length - 1; i > 0; i--) {
  for (int j = 0; j < i; j++) {
    if (ar[j] > ar[j + 1]) {
      temp = ar[j];
      ar[j] = ar[j + 1];
      ar[j + 1] = temp;
  }
}
```

Searching

Input is an binary array A of size 100.

```
Find any index i for which A[i] = 1
```

var b = random index from $\{1 \dots 100\}$ Makes 1 probe to A // b = 1 with prob. 0.01 Success probability = m/100Where, m = number of 1s in A// b = 2 with prob. 0.01 // b = 100 with prob. 0.01 Run the code k times. var c = A[b]Prob. of getting "not found" in all 10 if c = 1: $runs = (1 - m/100)^{k}$ print (b) else: Prob. of finding good index ≅ print ("not found") k*(m/100) (linear in k)

k = 100/m iterations ensure success

Quantum searching

Input is an binary array A of size 100.

```
Find any index i for which A[i] = 1
```

Define UA gate on two "registers":

- 1. 10-qubit register 1 to store position
- 1-qubit register 2 to store value 2.

 $|p\rangle |0\rangle \rightarrow |p\rangle |A[p]\rangle$ 1 probe to A $|p\rangle |1\rangle \rightarrow |p\rangle |1-A[p]\rangle$

```
var |b\rangle = 1/10 [|1\rangle + |2\rangle + ... + |100\rangle ]
```

Apply UA on $|b\rangle|0\rangle$?

1 probe to A

 $1/10 \left[|1\rangle|A[1]\rangle + |2\rangle|A[2]\rangle + \dots \right]$

 $1/10 \left[|1\rangle|A[1]\rangle + |2\rangle|A[2]\rangle + \dots \right]$ Observation yields any $|b\rangle|A[b]\rangle$ with probability 1/100

Not better than classical

Do not observe (yet). Run Grover's search algorithm on this state and then observe.

Constant success probability can be achieved using $\sqrt{(100/m)}$ probes.

Programming a QC



Early style of designing (efficient) solvers



Linear system of equations



Given N equation in the form of Ax = b.

- Classical : conjugate gradient descent ~ O(N)
- Quantum algorithm by HHL : O(log(N))

Gives the solution vector x

Gives a random sample from the solution vector x

Variational Quantum Eigensolver



Quantum Approximate Optimization Algorithms



Quantum neural network



Quantum annealing



Promises and Prospects

1981 Feynman proposed quantum computer to efficiently onshot. simulate many-body quantum systems

- **1984** Bennett and Brassard designed quantum protocol for BB84 running on **2000KM** fiber-optic cable in China secret key sharing
- **1991** Another QKD protocol by Ekert

QKD networks : DARPA, Tokyo, Vienna, Japan, ...



Promises and Prospects

race

 $1994 \sqrt{N}$ is the best

Technology not clear quantum computer

• **1992** Deutsch and Jozsa solved a (toy) problem in half the time taken by the best classical algorithm

1993 Simon designed algorithm that is efficient on quantum computer but inefficient classically

1996 Grover designed algorithm to search in a database of N elements using \sqrt{N} "lookups" (classical best is N/2)

... better-than-classical algorithms for problems on numbers, graphs, geometric objects, strings, statistics, communication, data structures, ... but <u>limited</u> speedup

Promises and Prospects

2001 15 factored using 10¹⁸ identical molecules

Requires • **1994** Shor designed algorithms to factor n-bit number in high-precision O(n²) time (classical best is O(exp(n^{1/3})))

- **1995** Shor and Steane designed error-correcting codes
- **1998** Gottesman and Knill showed how to <u>efficiently</u> simulate certain quantum algorithms <u>classically</u>
- 2017 Microsoft releases 40-qubit classical simulator ODE, PDE,
- machine learning • 2009 Harrow+ designed linear system "solver" Quantum ML
- **2015** Grassl showed 3000-7000 qubits needed to search Attacks on AES key using Grover's algorithm
 - **2016** Google simulated a Hydrogen molecule with 9 qubits

PORTFOLIO OPTIMIZATION on D-Wave

SIMULATION OF HYDROGEN MOLECULE by Google

(simulation of quantum-mechanical systems was

the initial motivation of Richard Feynman to

propose a quantum computer)

TRAFFIC OPTIMIZATION & EXPLORE MATERIAL STRUCTURE FOR E-VEHICLE BATTERY by Volkswagen Group and Google * Maybe you believe in the experiments yet disagree with the meaning

Summary

Quantum mechanics that drive quantum computing is mysterious

But if you are a believer ... (*)

Quantum algorithm design and analysis possible using knowledge of algorithms, probability and linear algebra.

Thanks to physicists, material scientists, engineers,

mathematicians, ... in universities, R&D labs and corporates ...

These algorithms can be implemented on real quantum computers and experimented with.

Too early to say how and where QC will become useful ...

Just the right time to enter the game.



Thank you for listening. Questions?

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