



Lecture

Quantum Computing

(CS5070)

What else?

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What else? - Simon's algorithm

- Simon's problem is proven to be solved exponentially faster on a quantum computer than on a classical computer
 - Simon's algorithm, served as the inspiration for Shor's algorithm
 - Simon's algorithm: $O(n)$ queries to the black box
 - Classical algorithm at least $\Omega(2^{\frac{n}{2}})$ queries
- Given a function (implemented by a black box or oracle)
 $f : \{0, 1\}^n \rightarrow \{0, 1\}^n$ with the promise that, for some unknown $s \in \{0, 1\}^n$, for all $x, y \in \{0, 1\}^n$, $f(x) = f(y)$ if and only if $x \oplus y \in \{0^n, s\}$, the goal is to identify s by making as few queries to $f(x)$ as possible.
 - i.e., distinguishing the $s = 0^n$ case, where the function is one-to-one, from the $s \neq 0^n$ case, where the function is two-to-one and satisfies
$$f(x) = f(x \oplus s)$$
 - little practical value

What else? - Bernstein–Vazirani algorithm

- designed to prove an oracle separation between complexity classes BQP and BPP
 - BQP: bounded-error quantum polynomial time is the class of decision problems solvable by a quantum computer in polynomial time, with an error probability of at most $\frac{1}{3}$ for all instances
 - BPP: bounded-error probabilistic polynomial time is the class of decision problems solvable by a probabilistic Turing machine in polynomial time with an error probability bounded away from $\frac{1}{3}$ for all instances
- Given an oracle that implements a function $f: \{0, 1\}^n \rightarrow \{0, 1\}$ in which $f(x)$ is promised to be the dot product between x and a secret string $s \in \{0, 1\}^n$ modulo 2,
$$f(x) = x \cdot s = x_1 s_1 \oplus x_2 s_2 \oplus \dots \oplus x_n s_n,$$
 find s
- Generalization is the [hidden linear function problem](#)

What else? - Quantum Fourier Transform

- used as basic routine in other quantum algorithms for
 - factoring and computing the discrete logarithm,
 - the quantum phase estimation algorithm for estimating the eigenvalues of a unitary operator, and
 - algorithms for the hidden subgroup problem

	Classical discrete Fourier transform	Quantum Fourier transform
Input	vector $(x_0, x_1, \dots, x_{N-1})$	quantum state $ x\rangle = \sum_{i=0}^{N-1} x_i i\rangle$
Notation	$N = 2^n$ with n #(qu)bits, $\omega_N = e^{\frac{2\pi i}{N}}$ and ω_N^n is an N -th root of unity	
Output	vector $(y_0, y_1, \dots, y_{N-1}) \in \mathbb{C}^N$ according to $y_k = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} x_n \omega_N^{-kn},$ $k = 0, 1, 2, \dots, N-1$	quantum state $\sum_{i=0}^{N-1} y_i i\rangle$ according to $y_k = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} x_n \omega_N^{nk},$ $k = 0, 1, 2, \dots, N-1$
Circuit Depth	$O(n \cdot 2^n)$ gates	Simple Decomposition: $O(n^2)$ Hadamard gates and controlled phase shift gates Best known one: $O(n \log n)$ gates [HH'00]

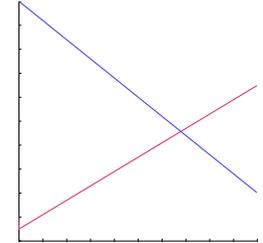
What else? - Hidden subgroup problem

- Given a group G , a subgroup $H \leq G$, and a set X , $f : G \rightarrow X$ **hides** the subgroup H if for all $g_1, g_2 \in G$, $f(g_1) = f(g_2)$ if and only if $g_1H = g_2H$. Equivalently, the function f is constant on the cosets of H , while it is different between the different cosets of H .
- **Hidden subgroup problem (HSP):** Let G be a group, X a finite set, and $f : G \rightarrow X$ a function that hides a subgroup $H \leq G$. The function f is given via an oracle, which uses $O(\log|G| + \log|X|)$ bits. Using information gained from evaluations of f via its oracle, determine a generating set for H .
 - used within Shor's algorithm
 - an efficient quantum algorithm for the HSP for the symmetric group would give a quantum algorithm for the graph isomorphism [EH'99]
 - An efficient quantum algorithm for the HSP for the dihedral group would give a quantum algorithm for the $poly(n)$ unique shortest vector problems [R'03]

What else? - Phase Estimation

- **Eigenvector/-value:** If T is a linear transformation from a vector space V over a field F into itself and v is a nonzero vector in V , then v is an eigenvector of T if $T(v)$ is a scalar multiple of v , i.e., $T(v) = \lambda v$, where λ is a scalar in F (called eigenvalue)
- Let U be a unitary operator that operates on m qubits with an eigenvector $|\psi\rangle$, such that $U|\psi\rangle = e^{2\pi i \theta} |\psi\rangle$, $0 \leq \theta < 1$.
Phase estimation finds the eigenvalue $e^{2\pi i \theta}$ of $|\psi\rangle$, which in this case is equivalent to estimating the phase θ , to a finite level of precision ε .
 - $O(\log(\frac{1}{\varepsilon}))$ qubits (without counting the ones used to encode the eigenvector state) and $O(\frac{1}{\varepsilon})$ controlled- U operations
- used as a subroutine in other quantum algorithms, such as
 - Shor's algorithm and
 - HHL
- Other algos for eigenvalues: **variational quantum eigensolver (VQE)**

What else? - Quantum Linear Systems Problem (QLSP)



Find \vec{x} , such that $A \cdot \vec{x} = \vec{b}$, where \vec{x}, \vec{b} are N -dimensional vectors, A is $N \times N$ matrix

Reference	Complexity	Subroutines					Note
		H	P	L	A	V	
Harrow, Hassidim, Lloyd 2008 arXiv	$O\left(\frac{\kappa^2 \cdot \log(N)}{\epsilon}\right)$	✓	✓		✓		known as HHL algorithm, quadratic scaling in condition number
Ambainis 2012 arXiv	$O\left(\frac{\kappa \cdot \log(N)}{\epsilon^3}\right)$	✓	✓			✓	better scaling in condition number & worse scaling in precision
Childs, Kothari, Somma 2017 arXiv	$O\left(\kappa \cdot \log(N) \cdot \text{poly} \log\left(\frac{1}{\epsilon}\right)\right)$	✓		✓		✓	Exp. improvement in precision & still linear in condition number
Subaşı, Somma, Orsucci 2018 arXiv	$O\left(\frac{\kappa \cdot \log(N)}{\epsilon}\right)$	✓					simple "adiabatic-inspired" randomized algorithm
An, Li 2019 arXiv	$O\left(\frac{\kappa \cdot \log(N)}{\epsilon}\right)$	✓					adiabatic algorithm

Legend: **H:** Hamiltonian Simulation **P:** Phase Estimation **A:** Amplitude Amplification
V: Variable-Time Amplitude Amplification **L:** Linear Combination of Unitaries

What else? - Linear Regression

- Problem to be solved: $\min_{w \in \mathbb{R}^{d+1}} E(w)$
 - $E(w) = \|X \cdot w - Y\|^2$ Euclidean error function
 - $X \in \mathbb{R}^{N \times (d+1)}$, i.e.
 - X contains N data points ($N \in \mathbb{N}$) along its rows, and
 - each data point is a d dimensional row vector ($d \in \mathbb{N}$, #features), augmented by unity, having a total length of $d + 1$
 - Y : Regression labels ($Y \in \mathbb{R}^N$), i.e. the dependant variable in linear regression
 - w : Regression weights to be learned, $w \in \mathbb{R}^{d+1}$
- Best classical algorithm $O(N \cdot d^{1.37})$ time, where N is the size of the training data (using a fast matrix multiplication algorithm, such as Coppersmith–Winograd)
- Quantum linear regression algorithm [W'17]: $O(\text{poly}(\log(N), d, \kappa, \frac{1}{\epsilon}))$
- Linear regression can be formulated as QUBO problem [DP'21] (to be run on quantum annealer)



What else? - Quantum Counting

- quantum algorithm for efficiently counting the number of solutions for a given search problem
 - Consider $f : \{0, 1\}^n \rightarrow \{0, 1\}$ to be the oracle function returning 1 if the input is a solution
 - Calculate the number of solutions $M = |f^{-1}(1)|$
- based on
 - the quantum phase estimation algorithm,
 - quantum Fourier transform and
 - on Grover's search algorithm
- **Runtime Quantum Solution:** $O\left(\sqrt{\frac{N}{M}}\right)$, where $N = 2^n$
- **Runtime Classical Solution:** $\Omega(N)$
 - without any prior knowledge of the structure of the function f

What else? - Quantum Random Walk

- (Classical) random walk [P1905] is a random process that describes a path that consists of a succession of random steps on some mathematical space
 - Example: random walk on the integer number line \mathbb{Z} which starts at 0, and at each step moves ± 1 with equal probability
- Quantum walks: random process is achieved by
 1. quantum superposition of states,
 2. non-random, reversible unitary evolution and
 3. state measurements
- Speedups of quantum walks over any classical algorithm
 - Exponential for specific problems [C+'03] [C+'07]
 - Polynomial for many practical problems like
 - the element distinctness problem [A'07],
 - the triangle finding problem [MSS'05],
 - and evaluating NAND trees [FGG'07].
 - Grover's search can also be regarded as some kind of quantum walk algorithm

What else? - Quantum Artificial Life

- Algorithm for simulating life processes according to Darwinian Evolution including
 - self replications of individuals
 - interactions between individuals
 - mutations and
 - death of individuals
- Realization [R+'16] [R+'18]:
 - Realization of individuals by two qubits for their genotype and phenotype
 - Setup: the state of the genotype is instantiated by some rotation of an ancillary state $|0\rangle\langle 0|$
 - Copying the genotype for transmission of genetic information through generations
 - Phenotype is dependent on the genotype as well as the individual's interactions with their environment
 - Individuals move throughout a two-dimensional spatial grid and occupy cells randomly
 - two or more individuals occupy the same cell \Rightarrow they interact with each other

What else? - Quantum approximate optimization algorithm (QAOA)

- **To be maximized:** $C(z) = \sum_{\alpha=1}^m C_{\alpha}(z)$
 - $z = z_1 z_2 \dots z_n$ is the bit string and
 - $C_{\alpha}(z) = 1$ if z satisfies clause α and 0 otherwise
 - **Typically:** C_{α} depends on only a few of the n bits
 - **Satisfiability:** Is there a string that satisfies every clause?
 - **MaxSat:** Determine string that maximizes $C(z)$
 - **Approximate optimization:** Determine string z for which $C(z)$ is close to the maximum of C
- **Mapping from quadratic binary optimization (QUBO) - problems to QAOA**
 - \Rightarrow In practice universal quantum computers can directly process problems proposed for quantum annealing

What else?

- Clustering
 - Quantum variant of k-means: q-means [K+'18]
 - Support Vector Clustering (SVC) [B-H+'02], Quantum Clustering (QC) [HG'01], Approximate Quantum Clustering (AQC) [S'13], Dynamic Quantum Clustering (DQC) [WH'09]
- Nearest Neighbor Search (e.g. [WKS'14], [L+'21])
- ...

Summary & Conclusions

- Simon's algorithm
- Bernstein–Vazirani algorithm
- Quantum Fourier Transform
- Hidden subgroup problem
- Phase Estimation
- Quantum Linear Systems Problem (QLSP)
- Quantum Linear Regression
- Quantum Counting
- Quantum Random Walk
- Quantum Artificial Life
- Quantum approximate optimization algorithm (QAOA)
- Clustering
- Nearest Neighbor Search