CSE 392/CS 395T/M 397C: Matrix and Tensor Algorithms for Data

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Lecture 26: Introduction to quantum computing II



- 1 Complexity Classes
- **2** Quantum Fourier Transform
- (3) Quantum Phase Estimation
- 4 Linear system solver

Complexity Classes

• There are many intractable problems where the best known algorithm has runtime that scales exponentially with input size



Complexity Classes

• Quantum Computers are the Only Novel Hardware which Changes the Game



The Complexity Zoo

• Should We Focus on NP-hard Problems ?

 Counter to the layman belief, there is a consensus among quantum computing researchers that quantum computing is not likely to exponentially speed-up computation of NP-hard problems [C.H. Bennett, E. Bernstein, G. Brassard, U. Vazirani, Strengths and Weaknesses of Quantum Computing, 1996]



Fourier Transform - Background

- Fourier Transform: Decomposes a function or a signal in one domain (e.g. time) into its constituent frequency representation
- **Instrumental** in signal processing, image analysis, (convolutional) neural networks, etc
- Gilbert Strang described the FFT as "the most important numerical algorithm of our lifetime"
- Inducted in Top 10 Algorithms of 20th Century by the IEEE journal Computing in Science & Engineering
- Classically, the Fast Fourier Transform (FFT) can perform the task in $N \log(N)$ run-time [Cooley and Tukey, 1965]
- Qunatumly, the Quantum Fast Fourier Transform (FFT) is due to [Coppersmith, 1994]



- Similarly to the classical, the quantum Fourier Transform, (QFT) performs a discrete Fourier transform on the complex valued vector $|\psi\rangle$, yet it can achieve runtime of $\mathcal{O}(n \log n)$
- Given: an *n*-qubit state as a superposition of basis states $|0\rangle, |1\rangle, \dots, |2^n 1\rangle$
- **Map** each basis state $|j\rangle$

$$QFT(|j\rangle) = \frac{1}{\sqrt{2^n}} \sum_{k=0}^{2^n-1} e^{\frac{2\pi i jk}{2^n}} |k\rangle$$

• Notation: Fractional Binary Notation :

$$[0.x_1...x_m] = \sum_{k=1}^m x_k 2^{-k}$$

- For instance, $[0.x_1] = \frac{x_1}{2}$ and $[0.x_1x_2] = \frac{x_1}{2} + \frac{x_2}{2^2}$
- With this notation, the action of the quantum Fourier transform can be expressed in a compact manner:

$$QFT(|x_1x_2...x_n\rangle) = \frac{1}{\sqrt{N}} \left(|0\rangle + e^{2\pi i [0.x_n]}|1\rangle\right) \otimes \left(|0\rangle + e^{2\pi i [0.x_{n-1}x_n]}|1\rangle\right) \otimes \cdots \otimes \left(|0\rangle + e^{2\pi i [0.x_1x_2...x_n]}|1\rangle\right)$$

 or

$$QFT(|x_1x_2\dots x_n\rangle) = \frac{1}{\sqrt{N}} \ (|0\rangle + \omega_1^x|1\rangle) \otimes (|0\rangle + \omega_2^x|1\rangle) \otimes \cdots \otimes (|0\rangle + \omega_n^x|1\rangle)$$

- The algorithm effectively takes the 2^n amplitudes of an *n*-qubit state as a vector of size 2^n and performs a **discrete Fourier Transform** so that the result is encoded in the amplitudes of the output state
- The simplest way to show that the normalized Fourier Transform is a unitary operation is to demonstrate the quantum circuit that performs the QFT



• The input register contains an *n*-qubit basis state $|x\rangle$ expressed as the tensor product of the individual qubits in its binary expansion:

$$|x\rangle \equiv |x_1x_2\cdots x_n\rangle \equiv |x_1\rangle \otimes |x_2\rangle \otimes \cdots \otimes |x_n\rangle$$

- The gates labeled R_m represent a series of single-qubit phase rotations
- For each integer $m \ge 2$, the gate R_m shifts the phase of the $|1\rangle$ component of the input qubit by a factor of $e^{\frac{2\pi i}{2m}}$, representing the unitary transformation

$$\begin{pmatrix} 1 & 0 \\ 0 & e^{\frac{2\pi i}{2^m}} \end{pmatrix}$$

- However, in the QFT circuit, each R^m gate is **controlled** by another qubit (indicated by a large dot connected to the gate by a vertical line)
- Given a two-qubit state, $|\psi_1\rangle|\psi_2\rangle$, composed of the controlling qubit, $|\psi_1\rangle$, and the input qubit, $|\psi_2\rangle$, the **controlled**- R^m gate represents the unitary transformation

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & e^{\frac{2\pi i}{2^{m}}} \end{pmatrix}$$

- If the controlled- R_m gate is being applied to a **basis state**, $|x_\ell\rangle$, where x_ℓ is either 0 or 1, then depending on the value of x_ℓ , the controlled- R_m gate performs the identity transformation, or the R_m transformation
- However, we may combine the two and equivalently say that the controlled- R_m gate performs the transformation

$$\begin{pmatrix} 1 & 0 \\ 0 & e^{\frac{2\pi i x_{\ell}}{2^m}} \end{pmatrix}$$

on the qubit $|\psi_2\rangle$, effectively performing a **data-dependent phase rotation**

Quantum Fourier Transform - Circuit Analysis



1 After the first Hadamard gate on qubit 1, the state is transformed from the input state to

$$H \otimes I_{n-1} | x_1 x_2 \dots x_n \rangle = \frac{1}{\sqrt{2}} \left(|0\rangle + e^{\frac{2\pi i}{2}x_1} |1\rangle \right) \otimes | x_2 x_3 \dots x_n \rangle$$

2 Following application of R_2 on qubit 1 controlled by qubit 2, the state becomes

$$\frac{1}{\sqrt{2}}\left(|0\rangle + e^{\frac{2\pi i}{2^2}x_2 + \frac{2\pi i}{2}x_1}|1\rangle\right) \otimes |x_2x_3\dots x_n\rangle$$

3 After the application of the last R_n gate on qubit 1 controlled by qubit n, the state is

$$\frac{1}{\sqrt{2}}\left(|0\rangle + e^{\frac{2\pi i}{2^n}x_n + \frac{2\pi i}{2^{n-1}}x_{n-1} + \dots + \frac{2\pi i}{2^2}x_2 + \frac{2\pi i}{2}x_1}|1\rangle\right) \otimes |x_2 x_3 \dots x_n\rangle$$

Quantum Fourier Transform - Circuit Analysis



4 Application of a similar sequence of gates for qubits $2 \dots n$, the final state is:

$$\frac{1}{\sqrt{2}}\left(|0\rangle + e^{\frac{2\pi i}{2^n}x}|1\rangle\right) \otimes \frac{1}{\sqrt{2}}\left(|0\rangle + e^{\frac{2\pi i}{2^{n-1}}x}|1\rangle\right) \otimes \cdots \otimes \frac{1}{\sqrt{2}}\left(|0\rangle + e^{\frac{2\pi i}{2^2}x}|1\rangle\right) \otimes \frac{1}{\sqrt{2}}\left(|0\rangle + e^{\frac{2\pi i}{2^1}x}|1\rangle\right)$$

which is the QFT of the input state in reversed order

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Quantum Phase Estimation

- The **Quantum Phase Estimation** algorithm is one of the most instrumental algorithms in Quantum Computing
- Proposed first in **von Neuman's** Mathematical Foundations of Quantum Mechanics book (aka von Neumann measurement)
- $\bullet\,$ Quantum Computing formulation is due to ${\bf Kitaev}$
- **Applications** range from factoring, through eigenvalue decomposition, linear system solver and more...







Quantum Phase Estimation

- Recall : Unitary Eigenvalues : Let U be a $N \times N$ unitary transformation. U has an orthonormal basis of eigenvectors $|\psi_1\rangle, |\psi_2\rangle, \ldots, |\psi_N\rangle$ with eigenvalues $\lambda_1, \lambda_2, \ldots, \lambda_N$, where $\lambda_j = e^{2\pi i \theta_j}$ for some θ_j
- **Proof** : U, being **unitary**, **maps unit vectors to unit vectors** and hence all the eigenvalues have **unit magnitude**, i.e. they are of the form $e^{2\pi i\theta}$ for some θ
 - Let $|\psi_j\rangle$ and $|\psi_k\rangle$ be two **distinct** eigenvectors with **distinct** eigenvalues λ_j and λ_k

▶ We have that

 $\lambda_j \langle \psi_j, \psi_k \rangle = \langle \lambda_j \psi_j, \psi_k \rangle = \langle U \psi_j, \psi_k \rangle = \langle \psi_j, U \psi_k \rangle = \langle \psi_j, \lambda_k \psi_k \rangle = \lambda_k \langle \psi_j, \psi_k \rangle$

• Since $\lambda_j \neq \lambda_k$, the inner product $\langle \psi_j, \psi_k \rangle$ is 0, i.e. the eigenvectors $|\psi_j\rangle$ and $|\psi_k\rangle$ are orthonormal

Quantum Phase Estimation

- Goal: Phase Estimation : Given a unitary transformation U, and one of its eigenstate $|\psi_j\rangle$
- Find: the corresponding eigenvalue $\lambda_j = e^{2\pi i \theta_j}$ (or, equivalently, $\theta_j \in \mathbb{R}$)
- Reminder : Controlled U : For any unitary transformation U, the controlled U gate, CU, transforms the target register $|\psi\rangle$ to $U|\psi\rangle$ conditionally upon the control input qubit



• Estimation of the phase θ can be performed by the following simple prototype circuit



Quantum Phase Estimation Prototype - Circuit Analysis



• The application of **H** gate upon the control qubit, transfers the controller into a **uniform superposition** state

$$H \otimes I_n |0\rangle |\psi\rangle = \frac{1}{\sqrt{2}} (|0\rangle + |1\rangle) |\psi\rangle$$

• Consequent application of the controlled **CU** entails

$$CU \ \frac{1}{\sqrt{2}} \left(|0\rangle + |1\rangle \right) |\psi\rangle = \frac{1}{\sqrt{2}} |0\rangle |\psi\rangle + \frac{1}{\sqrt{2}} |1\rangle \lambda |\psi\rangle$$
$$= \left(\frac{1}{\sqrt{2}} |0\rangle + \frac{\lambda}{\sqrt{2}} |1\rangle \right) \otimes |\psi\rangle$$

• Note : After application of CU gate, the eigenstate, $|\psi\rangle$ remained unchanged while were able to push λ into the phase (phase kickback) of the controller qubit

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Quantum Phase Estimation Prototype - Circuit Analysis



• Application of an additional **Hadamard** gate upon the controller qubit will **transform** the state into a **measurable amplitude** in the Z basis

$$H \left(\frac{1}{\sqrt{2}}|0\rangle + \frac{\lambda}{\sqrt{2}}|1\rangle\right) = \frac{1+\lambda}{2}|0\rangle + \frac{1-\lambda}{2}|1\rangle$$

- To perform a (more) efficient implementation of the phase estimation algorithm we need to extend the set of ancillary qubits
- Definition: m-Controlled U : For any unitary transformation U, m-controlled U gate, $C_m U$, performs the transformation $C_m U |k\rangle \otimes |\psi\rangle = |k\rangle \otimes U^k |\psi\rangle$



where $k \in \{0, 1, ..., 2^m - 1\}$

• Estimation of θ within *m* bits of **precision** is equivalent to estimating the integer *j*, where $\frac{j}{2^m}$ is the **closest approximation** to θ

• Let $w_m = e^{\frac{2\pi i}{2m}}$, the circuit below **estimates** the **phase** efficiently



• The Hadamard (over m qubits this time) results in a uniform superposition

$$H^{\otimes m} \otimes I_n |0^m\rangle |\psi\rangle = \left(\frac{1}{\sqrt{2^m}} \sum_{k=0}^{2^m-1} |k\rangle\right) \otimes |\psi\rangle$$

 $\bullet~$ Next, application of the m-controlled~U gate

$$C_m U\left(\frac{1}{\sqrt{2^m}}\sum_{k=0}^{2^m-1}|k\rangle\right)\otimes|\psi\rangle = \left(\frac{1}{\sqrt{2^m}}\sum_{k=0}^{2^m-1}\lambda^k|k\rangle\right)\otimes|\psi\rangle = \left(\frac{1}{\sqrt{2^m}}\sum_{k=0}^{2^m-1}w_m^{jk}|k\rangle\right)\otimes|\psi\rangle$$

- Following the controlled operation the ancillary register contains the Fourier Transform mod 2^m of j
- How do we retrieve j back ?

- Following the controlled operation the ancillary register contains the Fourier Transform mod 2^m of j
- How do we retrieve j back? Apply the inverse of the Fourier Transform mod 2^m
- Recall that quantum circuits are **reversible**, thus, following the **inverse QFT** we get back j

$$QFT_{2^m}^{-1} \otimes I_n \left(\frac{1}{\sqrt{2^m}} \sum_{k=0}^{2^m-1} w_m^{jk} |k\rangle \right) \otimes |\psi\rangle = |j\rangle \otimes |\psi\rangle$$

- If $\theta = \frac{j}{2^m}$, then the circuit outputs j
- If $\theta \approx \frac{j}{2^m}$, then the circuit outputs j with high probability

Quantum Phase Estimation - Circuit Description



Linear System of Equations - Problem Definition

- Given a matrix A and a vector $|b\rangle$
- Find a vector $|x\rangle$ such that $A|x\rangle = |b\rangle$
- Solution of linear systems of equations is instrumental across most disciplines of science and engineering
- The study of Harrow, Hassidim and Lloyd (2008) provided algorithmic framework for linear regression with an exponential speed-up
- Note: it is a quantum **algorithm** but not a realizable **quantum computation**. i.e. it does not map classical input to a classical result, but rather manipulates quantum data only



Assumptions and Disclaimers

- The algorithm exemplifies the **gap** between the **desired computation** of providing the system A, and RHS b, and extracting x, vs. the quantum computation of $|x\rangle$ given A and $|b\rangle$. This difference is not as subtle as it may appear at first glance
- **Data loading** into the quantum computer, is assumed to be performed in **logarithmic cost** with respect to the **problem size** (an unrealistic assumption for general data), or alternatively the data is (somehow) already stored in a so-called QRAM (quantum RAM)
- Classical output is limited to a low dimensional function of the solution (e.g. an expectation)
- Known approximation of the expectation value of some operator associated with x, e.g., $x^{\dagger}Mx$ for some matrix M
- A is sparse, ($s \ll N$ in entries / row), Hermitian $N \times N$ with condition number κ (this assumption can be avoided)

High Level Algorithmic Schematics



• Solve Ax = b, where $|x\rangle$ and $|b\rangle$ are quantum states, and A represents the Hamiltonian

High Level Algorithmic - Steps



- **()** State preparation prepare the state $|b\rangle$ (amplitude encoding), *n* ancilla qubits at $|0\rangle$, additional ancillar qubit at $|0\rangle$
- **2** Quantum Phase Estimation perform phase estimation upon the state $|b\rangle$, using the *n* ancillar qubits extract eigenvalues of A $QPE_A|b\rangle|0\rangle^{\otimes n} = \sum_j \beta_j |\psi_j\rangle|\bar{\lambda_j}\rangle$
- **3** Conditional rotation performs $\sum \beta_j |\psi_j\rangle |\bar{\lambda_j}\rangle |0\rangle \rightarrow \sum \beta_j |\psi_j\rangle |\bar{\lambda_j}\rangle \left(\sqrt{1 \frac{C^2}{\lambda_j^2}} |0\rangle + \frac{C}{\lambda_j} |1\rangle\right)$
- **Olymptic QPE** uncompute eigenvalue register with the **inverted phase** $\sum \beta_j |\psi_j\rangle |0\rangle \left(\sqrt{1 - \frac{C^2}{\lambda_j^2}} |0\rangle + \frac{C}{\lambda_j} |1\rangle \right)$
- Rejection sampling identify cases in which the conditional rotation was successful
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- Assume $A = A^{\dagger}$
- Otherwise, solve instead

$$\begin{pmatrix} 0 & A \\ A^{\dagger} & 0 \end{pmatrix} \begin{pmatrix} 0 \\ x \end{pmatrix} = \begin{pmatrix} b \\ 0 \end{pmatrix}$$

which implies that Ax = b for all A (not necessarily square, can be over-determined or under-determined)

- Per the sparsity assumption upon $A, s \ll N, A = A^{\dagger}$ is local
- Then exponentiation of the operator A (aka Hamiltonian simulation) e^{-iAt} can be performed in time $\mathcal{O}(\log(N))$ [Lloyd 1996]

• If we know how to diagonalize A, i.e.

$$UAU^{\dagger} = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_N \end{pmatrix}$$

then inverse is just the inverse of the diagonal elements

$$U^{\dagger} A^{-1} U = \begin{pmatrix} \lambda_1^{-1} & 0 \\ & \ddots & \\ 0 & & \lambda_N^{-1} \end{pmatrix}$$

- Based on Kitaev's QPE algorithm for finding eigenpairs, a momentum operator p is used to advance the system $|b\rangle|0\rangle$ by a distance proportional to the eigenvalue of A
- Let the state $|b\rangle$ be represented by an **eigen decomposition** of A, i.e. $|b\rangle = \sum_{i} \beta_{i} |\psi_{i}\rangle$

$$QPE_A|b\rangle|0\rangle = \sum_j \beta_j |\psi_j\rangle|\lambda_j\rangle$$

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• Next, we pick the inverse of the eigenvalues λ_j and turn them into a phase

$$\sum_{j}\beta_{j}e^{i\delta\lambda_{j}^{-1}}|\psi_{j}\rangle|\lambda_{j}\rangle$$

with a small δ

• Next, following the swap of λ and λ^{-1} undo the phase estimation operation

$$\beta_j e^{i\delta\lambda_j^{-1}} |\psi_j\rangle |0\rangle$$

• The above term is essentially

$$e^{i\delta A^{-1}}|b\rangle|0\rangle$$

• If δ is small enough

$$e^{i\delta A^{-1}}|b\rangle|0\rangle\approx(I+i\delta A^{-1})|b\rangle|0\rangle=(|b\rangle+i\delta\underbrace{A^{-1}|b\rangle}_{|x\rangle})|0\rangle$$

• Thus, within the expression we have the desired $A^{-1}|b\rangle$ which can be extracted with probability of δ^2

- The classical algorithms can find x and estimate $x^{\dagger}Mx$ in $\tilde{\mathcal{O}}(N\sqrt{\kappa})$ run time
- For A of condition number, κ , in k steps we get $A^{-1}|b\rangle$ to accuracy of $\mathcal{O}(\frac{\kappa^2 s^2}{\epsilon} log N)$
- This is indeed a remarkable exponential acceleration
- Consequent work improved upon the condition number dependency [A. Childs, R. Kothari and R. Somma, 2015]
- Extension to non-sparse settings and further complexity reduction based on quantum singular value estimation are due to $\mathcal{O}(\sqrt{N}\log N\kappa^2)$ [L. Wossing et al, 2018]

Questions