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Lecture 6

Quantum Fourier Transform and Phase Estimation

- Quantum Fourier transform
- Phase Estimation

1 Quantum Fourier transform

Hadamard transform: $|x\rangle \overset{H^{\otimes n}}{\longmapsto} \frac{1}{\sqrt{2^n}} \sum_{y \in \{0,1\}^n} (-1)^{x \cdot y} |y\rangle$ where x is an integer modulo 2^n . This is a Fourier transform over $\underbrace{\mathbb{Z}_2 \otimes \cdots \otimes \mathbb{Z}_2}_{2}$.

How about Fourier transform over $\mathbb{Z}_{2^n}^n$? That has the form:

$$|x\rangle \longmapsto \frac{1}{\sqrt{2^n}} \sum_{y \in \mathbb{Z}_{2^n}} e^{\frac{2\pi i x y}{2^n}} |y\rangle := |\tilde{x}\rangle$$

where $x \in \mathbb{Z}_{2^n}$ represents an integer modulo 2^n .

For n = 1, the transform is

$$|0\rangle \longmapsto \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle), \quad |1\rangle \longmapsto \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$$
$$|x\rangle \stackrel{H}{\longmapsto} \frac{1}{\sqrt{2^n}} \sum_{y \in \{0,1\}} e^{\frac{2\pi i x y}{2^n}}$$

In general, the $|\tilde{x}\rangle$ states form an orthonormal basis, the Fourier basis: $\langle \tilde{x} | \tilde{x}' \rangle = \delta_{x,x'}$. When do we need the quantum Fourier transform?

1.1 Phase estimation

• Given: Ability to implement a controlled unitary operator U and a quartum state $|\psi\rangle$ with $U|\psi\rangle=e^{i\theta}|\psi\rangle$.



• Problem: Learn θ .

1.2 Hadamard test

$$\begin{split} |0\rangle|\psi\rangle & \stackrel{H\otimes I}{\longmapsto} \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)|\psi\rangle & \stackrel{\text{controlled-U}}{\longmapsto} \frac{1}{\sqrt{2}}(|0\rangle|\psi\rangle + |1\rangle U|\psi\rangle) \\ & = \frac{1}{\sqrt{2}}\left(|0\rangle + e^{i\theta}|1\rangle\right)|\psi\rangle \\ & \stackrel{H\otimes I}{\longmapsto} \frac{1}{2}\left[(|0\rangle + |1\rangle) + e^{i\theta}(|0\rangle - |1\rangle)\right]|\psi\rangle = \left(\frac{1 + e^{i\theta}}{2}|0\rangle + \frac{1 - e^{i\theta}}{2}|1\rangle\right)|\psi\rangle \end{split}$$

$$\Pr(0) = \left| \frac{1 + e^{i\theta}}{2} \right|^2 = \frac{1}{4} \left[(1 + \cos \theta)^2 + \sin^2 \theta \right] \quad \Pr(1) = \left| \frac{1 - e^{i\theta}}{2} \right|^2 = \sin^2 \frac{\theta}{2}.$$
$$= \frac{1}{4} [2 + 2\cos \theta] = \cos^2 \frac{\theta}{2}$$

If $\theta = 0$ or $\theta = \pi$, we learn θ perfectly: After we measure the state in the computational basis, we get $|0\rangle$ with probability 1 when $\theta = 0$ and $|1\rangle$ with probability 1 when $\theta = 1$.

If $0 < \theta < \pi$. learn the probability distribution by samples to get information of θ .

Now, consider $\theta = 2\pi \cdot \sum_{j=1}^{n} \frac{x_j}{2^j}$, $x_j \in \{0,1\}$. i.e., $\theta = 2\pi \cdot 0$. $x_1 \dots x_n$ Consider what happens if we apply U^{2^k} .

$$U^{2^k}|\psi\rangle = e^{i2^k\theta}|\psi\rangle = e^{2\pi i \cdot x_1 \dots x_k \cdot x_{k+1} \dots x_n}|\psi\rangle = e^{2\pi i \cdot 0 \cdot x_{k+1} \dots x_n}|\psi\rangle$$

In particular, $U^{2^{n-1}}|\psi\rangle = e^{2\pi i \cdot 0.x_n}|\psi\rangle$.

In other words,
$$U^{2^{n-1}} |\psi\rangle = \begin{cases} |\psi\rangle & x_n = 0 \\ -|\psi\rangle & x_n = 1 \end{cases}$$
 $|0\rangle - H$ $|\psi\rangle - U^{2^{n-1}}$

Idea: Combine n such experiments for different exponents of U:

$$|0\rangle - H \qquad \cdots \qquad \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0 \cdot x_n} |1\rangle \right)$$

$$|0\rangle - H \qquad \cdots \qquad \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0 \cdot x_{n-1} x_n} |1\rangle \right)$$

$$\vdots \qquad \vdots \qquad \vdots \qquad \vdots$$

$$|0\rangle - H \qquad \cdots \qquad \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0 \cdot x_2 \cdot \cdots x_n} |1\rangle \right)$$

$$|0\rangle - H \qquad \cdots \qquad \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0 \cdot x_2 \cdot \cdots x_n} |1\rangle \right)$$

$$|\psi\rangle - U \qquad U^2 \qquad \cdots \qquad U^{2^{n-2}} \qquad |\psi\rangle$$

The output state is:

$$\bigotimes_{i=1}^{n} \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0.x_{n+1-i} \dots x_n} |1\rangle \right)$$

$$= \frac{1}{\sqrt{2^n}} \sum_{y=0}^{2^{n-1}} e^{\frac{2\pi i \cdot 2^{n-1} x y_{n-1}}{2^n}} |y_{n-1}\rangle e^{\frac{2\pi i \cdot 2^{n-2} x y_{n-2}}{2^n}} |y_{n-2}\rangle \dots e^{\frac{2\pi i x y_0}{2^n}} |y_0\rangle$$

$$= \frac{1}{\sqrt{2^n}} \sum_{y=0}^{2^{n-1}} e^{\frac{2\pi i x y}{2^n}} |y\rangle = |\tilde{x}\rangle$$

Conclusion: Phase estimation can be solved by inverse of quantum Fourier transform.

Note:
$$QFT: |x\rangle \longrightarrow |\tilde{x}\rangle = \frac{1}{\sqrt{2^n}} \sum_{y=0}^{2^n-1} e^{\frac{2\pi i xy}{2^n}} |y\rangle$$
 $x \in \{0, 1, \dots, 2^n-1\}$

For phase estimation: Ideally, we want to have x_1, \dots, x_n directly as outputs. How do we implement this?

The first qubit is $\frac{|0\rangle + e^{2\pi i \cdot 0.x_n} |1\rangle}{\sqrt{2}} = H |x_n\rangle$. Therefore, applying H reveals x_n . The second qubit is $\frac{|0\rangle + e^{2\pi i \cdot 0.x_{n-1}x_n} |1\rangle}{\sqrt{2}}$:

$$e^{2\pi i \cdot 0 \cdot x_{n-1} x_n} = e^{2\pi i \cdot 0 \cdot x_{n-1}} \cdot e^{2\pi i \cdot 0 \cdot 0 x_n} = e^{x_{n-1} \cdot \pi i} \cdot e^{x_n \cdot \frac{\pi i}{2}} = (-1)^{x_{n-1}} \cdot i^{x_n}.$$

We can remove the dependence on x_n :

If
$$x_n = 0$$
, do I ; if $x_n = 1$, do $\begin{pmatrix} 1 & 0 \\ 0 & i \end{pmatrix}^{\mathsf{T}}$.

Then the state becomes $\frac{1}{\sqrt{2}}(|0\rangle + (-1)^{x_{n-1}}|1\rangle)$. Then Hadamard gate H reveals x_{n-1} .

Inverse QFT with n=2:

$$\frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0.x_2} |1\rangle \right) - \boxed{H} \qquad |x_2\rangle$$

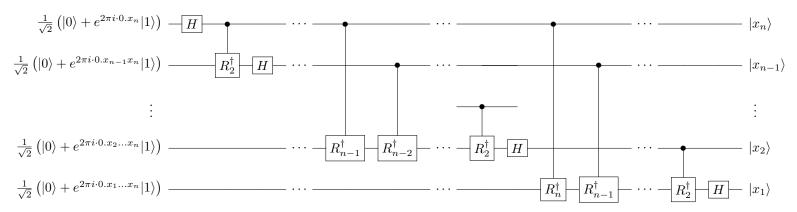
$$\frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0.x_1x_2} |1\rangle \right) - \boxed{\begin{pmatrix} 1 & 0 \\ 0 & i \end{pmatrix}^{\dagger} - \boxed{H} - |x_1\rangle}$$

For convenience, denote $R_k = \begin{pmatrix} 1 & 0 \\ 0 & e^{\frac{2\pi i}{2^k}} \end{pmatrix}$.

Move generally, since $e^{2\pi i \cdot 0 \cdot x_{n-k+1} \cdot \dots x_n} = e^{2\pi i \left(\frac{x_{n-k+1}}{2^1} + \frac{x_{n+12}}{2^2} + \dots + \frac{x_n}{2^k}\right)}$, the k^{th} qubit is

$$\frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i \cdot 0.x_{n-k+1} \dots x_n} |1\rangle \right) = R_k^{x_n} \dots R_3^{x_{n-k+3}} R_2^{x_{n-k+2}} H |x_{n-k+1}\rangle.$$

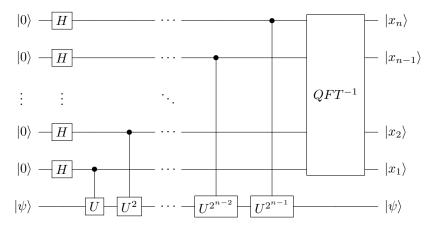
The circuit of inverse QFT:



Gate complexity of QFT^{-1} : $O(n^2)$

2 Phase Estimation

Big picture of phase estimation:



If $U|\psi\rangle = e^{2\pi i \cdot 0.x_1...x_n}|\psi\rangle$, this works. How about $e^{i\varphi}$ for general $\varphi \in [0, 2\pi)$?

$$|0\rangle^{\otimes n} \stackrel{H^{\otimes n}}{\longmapsto} \frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n - 1} |x\rangle \qquad x = \sum_{i=1}^n x_i \cdot 2^{n-i} \quad |x_i\rangle \stackrel{c-U^{2^{i-1}}}{\longmapsto} \begin{cases} |x_i\rangle & \text{if } x_i = 0 \\ e^{i\varphi^{2^{i-1}}} |x_i\rangle & \text{if } x_i = 1 \end{cases}$$

$$\stackrel{c-U_s}{\longmapsto} \frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n - 1} e^{i\varphi x} |x\rangle \qquad |x_1 \dots x_n\rangle \longmapsto \prod_{i=1}^n e^{i\varphi_{x_i} 2^{i-1}} |x_i\rangle$$

The QFT^{-1} is $\sum_{x=0}^{2^{n}-1} |x\rangle \langle \tilde{x}| = \frac{1}{\sqrt{2^{n}}} \sum_{x,y=0}^{2^{n}-1} e^{-\frac{2\pi i x y}{2^{n}}} |x\rangle \langle y|$.

$$\frac{1}{\sqrt{2^{n}}} \sum_{x=0}^{2^{n}-1} e^{i\varphi x} |x\rangle \stackrel{QFT^{-1}}{\longmapsto} \frac{1}{2^{n}} \sum_{x,y=0}^{2^{n}-1} e^{i\varphi x} e^{-\frac{2\pi i x y}{2^{n}}} |y\rangle
= \sum_{y=0}^{2^{n}-1} \alpha_{y} |y\rangle \quad \alpha_{y} := \frac{1}{2^{n}} \sum_{x=0}^{2^{n}-1} e^{i-\frac{2\pi y}{2^{n}} x} e^{i\varphi x} = \frac{1}{2^{n}} \sum_{x=0}^{2^{n}-1} e^{i\left(\varphi - \frac{2\pi y}{2^{n}}\right) x}.$$

where α_y is a geometric series. Denote $\tilde{\varphi} = \varphi - \frac{2\pi y}{2^n}$:

$$\sum_{x=0}^{2^{n}-1} e^{i\tilde{\varphi}x} = \frac{1 - e^{i\tilde{\varphi}2^{n}}}{1 - e^{i\tilde{\varphi}}} = \frac{e^{i\tilde{\varphi}2^{n-1}} \left(e^{-i\tilde{\varphi}2^{n-1}} - e^{i\tilde{\varphi}2^{n-1}}\right)}{e^{i\tilde{\varphi}/2} \left(e^{-i\tilde{\varphi}/2} - e^{i\tilde{\varphi}/2}\right)} \Rightarrow \left|\sum_{x=0}^{2^{n-1}} e^{i\tilde{\varphi}x}\right|^{2} = \frac{\sin^{2}\left(\tilde{\varphi}2^{n-1}\right)}{\sin^{2}(\tilde{\varphi}/2)}$$

Therefore: $\Pr(y) = \left|\alpha_y\right|^2 = \frac{1}{2^{2n}} \frac{\sin^2\left(\left(\varphi - \frac{2\pi y}{2n}\right) \cdot 2^{n-1}\right)}{\sin^2\left(\varphi - \frac{2\pi y}{2}\right) \cdot \frac{1}{2}\right)}.$ $\frac{\sin mx}{\sin x} \to m \text{ when } x \to 0.$

This distribution is tightly peaked around those y for which $\frac{2\pi y}{2^n} \approx \varphi$.

If
$$\varphi = \frac{2\pi y}{2^n}$$
, $\Pr(y) = \frac{1}{2^{2n}} \cdot (2^{n-1} \cdot 2)^2 = 1$.

Claim: Let $\frac{2\pi k}{2^n} \leqslant \varphi \leqslant \frac{2\pi (k+1)}{2^n}$. Then the probability of outputting either k or k+1 is at least $\frac{8}{\pi^2}$.

Proof. The probability of success is Pr(k) + Pr(k+1).

$$\Pr(k) + \Pr(k+1) = \frac{1}{2^{2n}} \left(\frac{\sin^2 \left(2^{n-1} \varphi - \pi k \right)}{\sin^2 \left(\frac{\varphi}{2} - \frac{\pi k}{2^n} \right)} + \frac{\sin^2 \left(2^{n-1} \varphi - \pi (k+1) \right)}{\sin^2 \left(\frac{\varphi}{2} - \frac{\pi (k+1)}{2^n} \right)} \right)$$

$$\left(\min \text{ when } \varphi = \frac{2\pi \left(k + \frac{1}{2} \right)}{2^n} \right) \geqslant \frac{2}{2^{2n}} \frac{\sin^2 \left(\pi \left(k + \frac{1}{2} \right) - \pi k \right)}{\sin^2 \left(\frac{\pi (k + \frac{1}{2})}{2^n} - \frac{\pi k}{2^n} \right)}$$

$$= \frac{1}{2^{2n-1}} \cdot \frac{1}{\sin^2 \frac{\pi}{2^{n+1}}}$$

$$(\sin x \leqslant x) \geqslant \frac{1}{2^{2n-1}} \cdot \frac{1}{\left(\frac{\pi}{2^{n+1}} \right)^2} = \frac{8}{\pi^2}.$$

Summary: Given $|\psi\rangle$ with $U|\psi\rangle = e^{i\varphi}|\psi\rangle$, we can produce an estimate of φ that differs from the true value by at most $\varepsilon\left(\frac{2\pi}{2^n}\right)$ with probability $\geqslant \frac{8}{\pi^2}$. This use QFT^{-1} with gate complexity $O\left((\log\frac{1}{\epsilon})^2\right)(n^2)$ and $O(1/\varepsilon)$ controlled $-U_s\left(2^n\right)$.