Quantum Computation

Quantum phase estimation and the quantum Fourier transform

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Encoding information in phases

In several quantum algorithms information is encoded in the relative phases of a quantum state.

The effect of Hadamard (once again)

$$
\begin{aligned}
H|x\rangle &= \frac{1}{\sqrt{2}}(|0\rangle + (-1)^x|1\rangle) = \frac{1}{\sqrt{2}}\sum_{y\in 2}(-1)^{xy}|y\rangle \\
H^{\otimes n}|x\rangle &= \frac{1}{\sqrt{2^n}}\sum_{y\in 2^n}(-1)^{x\cdot y}|y\rangle
\end{aligned}
$$

is to encode information about the value of x into the phases $(-1)^{x\cdot y}$ of basis states $|y\rangle$.

Encoding information in phases

Of course, as a reversible gate, the Hadamard gate also decodes information from phases:

$$
\begin{aligned}
H^{\otimes n} \frac{1}{\sqrt{2^n}} \sum_{y \in 2^n} (-1)^{x \cdot y} |y\rangle &= H^{\otimes n}(H^{\otimes n}|x\rangle) \\
&= (H^{\otimes n} H^{\otimes n})|x\rangle \\
&= I|x\rangle \\
&= |x\rangle
\end{aligned}
$$

Encoding information in phases

In general, phases are complex numbers

$$e^{2\pi i w}$$

for any real $w \in [0, 1]$.

Of course, $H^{\otimes n}$ cannot encode/decode information over such generic phases. The general situation can be described as follows:

The phase estimation problem

Determine a good estimation of the phase parameter w given a general quantum state

$$\frac{1}{\sqrt{2^n}} \sum_{y \in 2^n} e^{2\pi i w y} |y\rangle$$

An algorithm for phase estimation

Notation

$$w = 0 \,.\, x_1 x_2 \cdots$$

is written in base 2 (i.e. $w = x_1 2^{-1} + x_2 2^{-2} + \cdots$); thus

$$2^k w = x_1 x_2 \cdots x_k \,.\, x_{k+1} x_{k+2} \cdots$$

and

$$\begin{aligned} e^{2\pi i (2^k w)} &= e^{2\pi i (x_1 x_2 \cdots x_k \,.\, x_{k+1} x_{k+2} \cdots)} \\ &= e^{2\pi i (x_1 x_2 \cdots x_k)} \, e^{2\pi i (0 \,.\, x_{k+1} x_{k+2} \cdots)} \\ &= e^{2\pi i (0 \,.\, x_{k+1} x_{k+2} \cdots)} \end{aligned}$$

because $e^{2\pi i z} = 1$ for any integer z.

Case A: 1-qubit state and $w = 0.x_1$

$$\begin{aligned}
\frac{1}{\sqrt{2}}\sum_{y\in 2} e^{2\pi i(0.x_1)y}|y\rangle &= \frac{1}{\sqrt{2}}\sum_{y\in 2} e^{2\pi i(\frac{x_1}{2})y}|y\rangle \\
&= \frac{1}{\sqrt{2}}\sum_{y\in 2} e^{\pi i(x_1 y)}|y\rangle \\
&= \frac{1}{\sqrt{2}}\sum_{y\in 2} (-1)^{x_1 y}|y\rangle \\
&= \frac{1}{\sqrt{2}}(|0\rangle + (-1)^{x_1}|1\rangle)
\end{aligned}$$

Clearly H will decode and retrieve x_1 because

$$H\left(\frac{1}{\sqrt{2}}(|0\rangle + (-1)^{x_1}|1\rangle)\right) = |x_1\rangle$$

Case B: 2-qubit state and $w = 0.x_1x_2$

Observe that

$$\frac{1}{\sqrt{2^2}} \sum_{y \in 2^2} e^{2\pi i (0.x_1x_2) y} |y\rangle = \left(\frac{|0\rangle + e^{2\pi i (0.x_2)} |1\rangle}{\sqrt{2}} \right) \otimes \left(\frac{|0\rangle + e^{2\pi i (0.x_1x_2)} |1\rangle}{\sqrt{2}} \right)$$

which means that x_2, but not x_1, can be retrieved from the first qubit through an application of H.

The phase rotator

$$R_2 = \begin{bmatrix} 1 & 0 \\ 0 & e^{\frac{2\pi i}{4}} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & e^{2\pi i (0.01)} \end{bmatrix}$$

where 0.01 is in base 2 (thus, equal to 2^{-2}).

Case B: 2-qubit state and $w = 0.x_1x_2$

Taking $x_2 = 1$ and applying the inverse of the phase rotator to the second qubit, yields

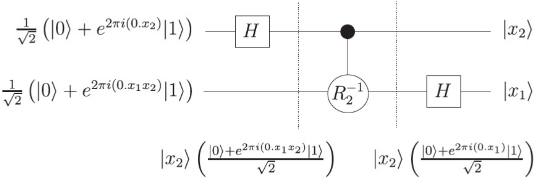
$$
\begin{aligned}
R_2^{-1}\left(\frac{|0\rangle + e^{2\pi i(0.x_1 1)}|1\rangle}{\sqrt{2}}\right) &= \begin{bmatrix} 1 & 0 \\ 0 & e^{-2\pi i(0.01)} \end{bmatrix}\left(\frac{|0\rangle + e^{2\pi i(0.x_1 1)}|1\rangle}{\sqrt{2}}\right) \\
&= \frac{|0\rangle + e^{2\pi i(0.x_1 1 - 0.01)}|1\rangle}{\sqrt{2}} \\
&= \frac{|0\rangle + e^{2\pi i(0.x_1)}|1\rangle}{\sqrt{2}}
\end{aligned}
$$

Concluding

- x_1 can now be determined by an application of H, as before.
- Moerevoer, the decision to apply R before the application of H depends on x_2 being 1 or 0, respectively.
- Thus, to find $w = 0.x_1x_2$ it is enough to apply a controlled version of R, precisely controlled by the state of the first qubit.

Case B: 2-qubit state and $w = 0.x_1 x_2$

The circuit



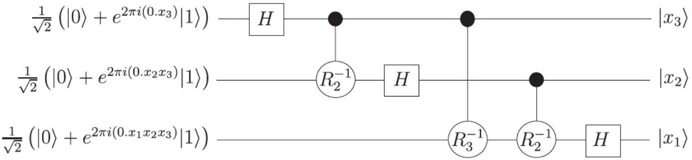
Case C: 3-qubit state and $w = 0.x_1x_2x_3$

The state is now

$$\frac{1}{\sqrt{2^3}} \sum_{y \in 2^3} e^{2\pi i (0.x_1x_2x_3) y} |y\rangle =$$

$$= \left(\frac{|0\rangle + e^{2\pi i (0.x_3)}|1\rangle}{\sqrt{2}} \right) \otimes \left(\frac{|0\rangle + e^{2\pi i (0.x_2x_3)}|1\rangle}{\sqrt{2}} \right) \otimes \left(\frac{|0\rangle + e^{2\pi i (0.x_1x_2x_3)}|1\rangle}{\sqrt{2}} \right)$$

In this case the third qubit has to conditionally rotate both x_2 and x_3, leading to the following circuit



Going generic

Gate R_3 in the circuit is an instance of a 1-qubit phase rotator

$$R_k = \begin{bmatrix} 1 & 0 \\ 0 & e^{\frac{2\pi i}{2^k}} \end{bmatrix}$$

whose inverse acts as

$$\begin{aligned} R_k^{-1}|0\rangle &= |0\rangle \\ R_k^{-1}|1\rangle &= e^{-2\pi i (0.0\cdots 1)}|1\rangle \end{aligned}$$

with 1 in $0.0 \cdots 1$ appearing in position k.

Going generic

The output state of the circuit is

$$|x_3 x_2 x_1\rangle$$

Thus, relabelling the qubits in reverse order, this provides an efficient circuit to estimate the phase (actually, to give a totally accurate estimation ...), by computing

$$\frac{1}{\sqrt{2^n}} \sum_{y \in 2^n} e^{2\pi i \left(\frac{x}{2^n}\right) y} |y\rangle \quad \rightsquigarrow \quad |x\rangle$$

Inverting ...

The inverse of the phase estimation transformation computes

$$|x\rangle \quad \rightsquigarrow \quad \frac{1}{\sqrt{2^n}} \sum_{y \in 2^n} e^{2\pi i \left(\frac{x}{2^n}\right) y} |y\rangle$$

which is obtained by taking the inverses of each gate and building the circuit in reverse order.

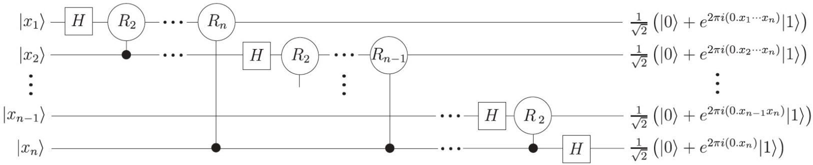
The result is formally identical to the discrete Fourier transform.

The quantum Fourier transform

QFT on basis states $|0\rangle, |1\rangle \cdots |k-1\rangle$

$$QFT_k(|x\rangle) \;=\; \frac{1}{\sqrt{k}} \sum_{y=0}^{k-1} e^{2\pi i \left(\frac{x}{k}\right) y} |y\rangle$$

The circuit



The quantum Fourier transform

Complexity (number of gates)

- one H plus $n-1$ conditional rotations on the first qubit
- one H plus $n-2$ conditional rotations on the second qubit
- ...

$$n + (n-1) + (n-2) + \cdots + 1 \; = \; \frac{n(n-1)}{2}$$

- plus $\frac{n}{2}$ swaps (each implemented by 3 CNOT gates)

Thus

$$\frac{n(n-1)}{2} + 3x\frac{n}{2} \; = \; \frac{n^2+2n}{2} \; \approx \; \mathcal{O}(n^2)$$

The quantum Fourier transform

Complexity (number of gates)

$$\frac{n(n-1)}{2} + 3x\frac{n}{2} = \frac{n^2+2n}{2} \approx \mathcal{O}(n^2)$$

which compares to the classical case for the Fast FT: $\mathcal{O}(n2^n)$

The result is impressive: the quantum version requires exponentially less operations to compute the Fourier transform than the (best) classical one.

- However, typical uses (e.g. in speech recognition) are limited by the impossibility of directly measuring the Fourier transformed amplitudes of the original state.
- This requires a subtler use of QFT in practice: the phase estimation procedure, underlying many quantum algorithms (e.g. Shor and the determination of the number of solutions in an unstructured search), is one of them.

Are we done?

- The circuit for QFT_k computes the QFT for k a power of 2, i.e $k = 2^n$
- The phase estimation algorithm works only when the phase is of the form $w = 0.x_1 x_2 \cdots x_n$, i.e. $\frac{x}{2^n}$ for some integer x

However, it can be shown that, for an arbitrary w, the algorithm will compute x such that $\frac{x}{2^n}$ is closest to w with high probability.

The question

What is the error emerging when w is not an integer multiple of $\frac{1}{2^n}$?

Are we done?

QFT^{-1} computes some superposition

$$\sum_x \alpha_x(w)|x\rangle$$

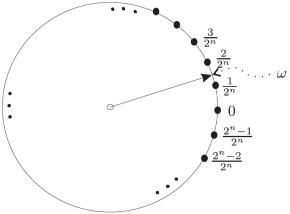
which represents the values of x that once measured gives a good estimate of w, outputing x with probability $|\alpha_x(w)|^2$.

This output x corresponds to an estimate

$$\tilde{w} = \frac{x}{2^n}$$

Are we done?

Consider w an integer **not** multiple of $\frac{1}{2^n}$, and let $\hat{w}$ be the nearest integer multiple of $\frac{1}{2^n}$ to w, i.e. $\hat{w} = \frac{\hat{x}}{2^n}$ is the closest number of this form to w.



Theorem

The phase estimation algorithm returns $\hat{x}$ with probability at least $\frac{4}{\pi^2}$, i.e. the algorithm outputs an estimate $\hat{x}$ with the given probability such that

$$\left| \frac{\hat{x}}{2^n} - w \right| \leq \frac{1}{2^{n+1}}$$

Are we done?

Theorem

$$\text{If} \quad \frac{x}{2^n} \leq w \leq \frac{x+1}{2^n}$$

The phase estimation algorithm returns either x or $x+1$ with probability at least $\frac{8}{\pi^2}$ i.e. the algorithm outputs an estimate $\hat{x}$ with the given probability such that

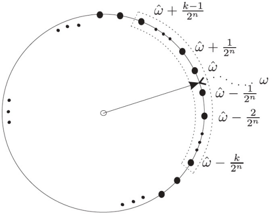
$$\left| \frac{\hat{x}}{2^n} - w \right| = \frac{1}{2^n}$$

The reverse question

How many qubits are required to get w accurate to n bits, with a probability p below a certain level?

Actually, the crucial choice is the value of n (number of qubits used) to ensure the estimation is close enough.

For $p = 1 - \frac{1}{2(k-1)}$, the algorithm returns one of the 2k closest integer multiples of $\frac{1}{2^n}$, i.e.



which means that $|w - \hat{w}| \leq \frac{k}{2^n}$.

The reverse question

Thus, to estimate $\hat{w}$ such that $|w - \hat{w}| \leq \frac{1}{2^r}$ with probability at least

$$1 - \frac{1}{2^m}$$

the maximum number of qubits required is

$$n \; = \; r + m + 1$$

- In practice a much smaller error is obtained: for example, with probability at least $\frac{8}{\pi^2}$, the error will be at most

$$\frac{1}{2^{r+m}}$$

Exercises

Recall the definition of QFT on K basis states:

$$QFT_K(|x\rangle) \;=\; \frac{1}{\sqrt{K}} \sum_{y=0}^{K-1} e^{2\pi i (\frac{x}{K}) y} |y\rangle$$

Exercise 1

Compute $QFT_K(|00\cdots 0\rangle)$.

Exercise 2

Verify the following equality, used in the slides but not proved.

$$QFT_K(|x_1 \cdots x_n\rangle) \;=\; \left(\frac{|0\rangle + e^{2\pi i (0.x_n)}|1\rangle}{\sqrt{2}}\right) \otimes \left(\frac{|0\rangle + e^{2\pi i (0.x_n x_{n-1})}|1\rangle}{\sqrt{2}}\right) \cdots \otimes \cdots \left(\frac{|0\rangle + e^{2\pi i (0.x_1 x_2 \cdots x_n)}|1\rangle}{\sqrt{2}}\right)$$

Exercises

Hint to Exercise 2: The case of QFT_4 applied to $|x\rangle = |x_1 x_2\rangle$

$$\begin{aligned} QFT_4(|x\rangle) &= \frac{1}{2}\sum_{y=0}^{3} e^{2\pi i x y 2^{-2}}|y\rangle \\ &= \frac{1}{2}\sum_{y_1, y_2=0}^{1} e^{2\pi i x (y_1 2^{-1} + y_2 2^{-2})}\,|y_1 y_2\rangle \end{aligned}$$

because, for $|y\rangle = |y_1 y_2\rangle$,

$$\frac{y}{2^n} = \sum_{j=1}^{n} y_j 2^{-j}$$

Exercises

Hint to Exercise 2: The case of QFT_4 applied to $|x\rangle = |x_1 x_2\rangle$

$$\cdots = \frac{1}{2} \sum_{y_1, y_2 = 0}^{1} (e^{2\pi i x y_1 2^{-1}} |y_1\rangle \otimes e^{2\pi i x y_2 2^{-2}} |y_2\rangle)$$

$$= \frac{1}{2} \sum_{y_1 = 0}^{1} (e^{2\pi i x y_1 2^{-1}} |y_1\rangle \otimes \sum_{y_2 = 0}^{1} e^{2\pi i x y_2 2^{-2}} |y_2\rangle)$$

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

$$= \frac{(|0\rangle + e^{2\pi i (x_1 . x_2)|1\rangle})}{\sqrt{2}} \otimes \frac{(|0\rangle + e^{2\pi i (0.x_1 x_2)|2\rangle})}{\sqrt{2}}$$

$$= \frac{(|0\rangle + e^{2\pi i (0.x_2)|1\rangle})}{\sqrt{2}} \otimes \frac{(|0\rangle + e^{2\pi i (0.x_1 x_2)|2\rangle})}{\sqrt{2}}$$

because, $e^{2\pi i (a.b)} = e^{2\pi i a} e^{2\pi i (0.b)} = e^{2\pi i (0.b)}$