# Quantum Computing Lecture 7

Quantum Factoring

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# Quantum Factoring

A *polynomial time* quantum algorithm for factoring numbers was published by *Peter Shor* in 1994.

polynomial time here means that the number of gates is bounded by a polynomial in the number of bits n of the number being factored.

The best known classical algorithms are exponential (in  $n^{1/3}$ ).

Fast factoring would undermine public-key cryptographic systems such as RSA.

# Period Finding

Suppose we are given a function  $f: \mathbb{N} \to \{0, \dots, N-1\}$  which we know is periodic, i.e.

$$f(x+r) = f(x)$$
 for some fixed  $r$  and all  $x$ .

Can we find the least value of r?

If we can find the period of a function efficiently, we can factor integers quickly.

# Order Finding

Suppose we are given an integer N and an a with a < N and

$$gcd(a, N) = 1.$$

Consider the function  $f_a:\mathbb{N} \to \{0,\dots,N-1\}$  given by

$$f_a(x) \equiv a^x \pmod{N}$$

Then,  $f_a$  is periodic, and if we can find the period r, we can factor N.

## Factoring

Suppose (for simplicity) N=pq, where p and q are prime. And, for some a < N, we know the period r of the function  $f_a$ . Then,  $a^{r+1} \equiv a \pmod{N}$ , so  $a^r \equiv 1 \pmod{N}$ .

If r is even and  $a^{r/2} + 1 \not\equiv 0 \pmod{N}$ , then take  $x^2 = a^r$ .

$$\begin{array}{ccc} x^2 - 1 & \equiv & 0 \pmod{N} \\ (x - 1)(x + 1) & \equiv & 0 \pmod{N} \end{array}$$

But,

$$x-1 \not\equiv 0 \pmod{N}$$
 (by minimality of  $r$ )  
 $x+1 \not\equiv 0 \pmod{N}$  (by assumption)

## Factoring—contd.

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So, (x-1)(x+1) = kpq for some k.
Now, finding gcd(N, x - 1) and gcd(N, x + 1) will find p and q.
If we randomly choose a < N
    (and check that gcd(a, N) = 1—if not, we've already found a
    factor of N)
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then, there is a probability  $> \frac{1}{2}$  that

- the period of  $f_a$  is even; and
- $a^{r/2} + 1 \not\equiv 0 \pmod{N}$ .

# Using a Fourier Transform

A fast period-finding algorithm allows us to factor numbers quickly.

The idea is to use a *Fourier Transform* to find the period of a function f.

Note, classically, we can use the *fast Fourier transform* algorithm for this purpose, but it can be shown that this would require time  $N \log N$ , which is exponential in the number of bits of N.

#### Discrete Fourier Transform

The discrete Fourier transform of a sequence of complex numbers

$$x_0, \ldots, x_{M-1}$$

is another sequence of numbers

$$y_0, \ldots, y_{M-1}$$

where

$$y_k = \frac{1}{\sqrt{M}} \sum_{i=0}^{M-1} x_i e^{2\pi i j k/M}$$

or

$$y_k = \frac{1}{\sqrt{M}} \sum_{i=0}^{M-1} x_j \omega^{jk}.$$

where  $\omega = e^{2\pi i/M}$ .

## **DFT** is Unitary

The discrete Fourier transform is a unitary operation on  $\mathbb{C}^M$ . Writing  $\omega$  for  $e^{2\pi i/M}$ ,

$$\omega, \omega^2, \dots, \omega^{M-1}, \omega^M = 1$$

are the Mth roots of 1.

$$D = \frac{1}{\sqrt{M}} \begin{bmatrix} 1 & 1 & 1 & \cdots & 1\\ 1 & \omega & \omega^2 & \cdots & \omega^{-1}\\ 1 & \omega^2 & \omega^4 & \cdots & \omega^{-2}\\ 1 & \omega^3 & \omega^6 & \cdots & \omega^{-3}\\ \vdots & \vdots & \vdots & \ddots & \vdots\\ 1 & \omega^{M-1} & \omega^{2M-2} & \cdots & \omega \end{bmatrix}$$

#### Inverse DFT

The inverse of the discrete Fourier Transform is given by:

$$D^{-1} = \frac{1}{\sqrt{M}} \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & \omega^{-1} & \omega^{-2} & \cdots & \omega \\ 1 & \omega^{-2} & \omega^{-4} & \cdots & \omega^{2} \\ 1 & \omega^{-3} & \omega^{-6} & \cdots & \omega^{3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{1-M} & \omega^{2-2M} & \cdots & \omega^{M-1} \end{bmatrix}$$

*Exercise:* Verify that D is unitary. Verify that  $D^{-1}$  as given above is the inverse of D.

## Quantum Fourier Transform

Computing the discrete Fourier transform classically takes time polynomial in M.

Shor showed that D can be implemented using a number of one and two-qubit gates that is only polynomial in the number of qubits  $O((\log M)^2)$ .

*Note:* This *does not* give a fast way to compute the DFT on a quantum computer.

There is no way to extract all the complex components from the transformed state.

# Fourier Transform on Binary Strings

Suppose  $M=2^n$ , and let  $|x\rangle$  be a computational basis state in  $\mathbb{C}^M$  with binary representation  $b_1\cdots b_n$ .

Let

$$\eta_j = e^{2\pi i (0 \cdot b_j b_{j+1} \cdots b_n)}.$$

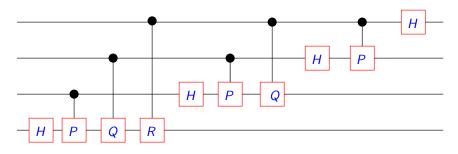
Then

$$D|x\rangle = (|0\rangle + \eta_n|1\rangle)(|0\rangle + \eta_{n-1}|1\rangle)\cdots(|0\rangle + \eta_1|1\rangle).$$

Exercise: Verify.

## Quantum Fourier Transform Circuit

We can use this form to implement the *quantum Fourier transform* using Hadamard gates and conditional phase-shift gates.



In the input, the least significant bit is at the top, in the output, it is at the bottom.

#### Conditional Phase Shifts

Here

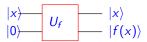
$$P = \begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix} \qquad Q = \begin{bmatrix} 1 & 0 \\ 0 & e^{i\pi/4} \end{bmatrix} \qquad R = \begin{bmatrix} 1 & 0 \\ 0 & e^{i\pi/8} \end{bmatrix}$$

Two-qubit conditional phase shift gates are actually symmetric between the two bits, despite the asymmetry in the drawn circuit.

It seems that for large n, an n-bit quantum Fourier transform circuit would require conditional phase shifts of *arbitrary precision*. It can be shown that this can be avoided with some (but not significant) loss in the probability of success *for the factoring algorithm*.

# Preparing the State

We are given an implementation of the function f as a unitary operator  $U_f$ 



Where, now, each of the two input wires represents n distinct qubits

Writing  $|\Psi\rangle$  for the state

$$H^{\otimes n}|0^n\rangle = \frac{1}{2^{n/2}}\sum_{x=0}^{2^n-1}|x\rangle.$$

We have,

$$|U_f|\Psi\rangle|0^n\rangle=\frac{1}{2^{n/2}}\sum_{x=0}^{2^n-1}|x\rangle|f(x)\rangle.$$

#### First Measurement

We measure the second n qubits of the state  $U_f |\Psi\rangle |0^n\rangle$  and get a value  $f_0$ . The state after measurement is:

$$\big(\frac{1}{\sqrt{m}}\sum_{k=0}^{m-1}|x_0+kr\rangle\big)|f_0\rangle.$$

where:

 $x_0$  is the least value such that  $f(x_0) = f_0$  r is the period of the function f $m = \lfloor \frac{2^n}{r} \rfloor$ .

# Applying the QFT

We apply the n-qubit quantum Fourier transform to the first n bits of the transformed state.

$$D\left(\frac{1}{\sqrt{m}}\sum_{k=0}^{m-1}|x_0+kr\rangle\right)$$

$$=\frac{1}{2^{n/2}}\sum_{y=0}^{2^n-1}\frac{1}{\sqrt{m}}\sum_{k=0}^{m-1}\omega^{(x_0+kr)y}|y\rangle$$

$$=\sum_{y=0}^{2^n-1}\omega^{x_0y}\frac{1}{2^{n/2}\sqrt{m}}\left(\sum_{k=0}^{m-1}\omega^{kry}\right)|y\rangle.$$

where  $\omega = e^{2\pi i/2^n}$ .

#### Second Measurement

The probability of observing a given state  $|y\rangle$  is:

$$\frac{1}{2^n m} \left| \sum_{k=0}^{m-1} \omega^{kry} \right|^2.$$

This probability function has peaks when  $ry/2^n$  is close to an integer. Indeed, if  $ry/2^n$  is an integer, then with probability 1 we measure a y that is a multiple of  $r/2^n$ .

Given an integer multiple of  $2^n/r$ , it is not difficult to find r.

### Exponentiation

To complete the factoring algorithm, we need to check that we can also implement the unitary transform  $U_f$  for the particular function

$$f_a(x) = a^x \mod N$$
.

with a number of quantum gates that is polynomial in  $\log N$ .

This is achieved through repeated squaring.

#### Some Points to Note

The two measurement steps can be combined at the end, with the Fourier transform applied before the measurement of f(x).

The probability of successfully finding the period in any run of the algorithm is only about 0.4.

However, this means a small number of repetitions will suffice to find the period with high probability.

Putting a *lower bound* on the conditional phase shift we are allowed to perform affects the probability of success, but not the rest of the algorithm.