

# Some quantum algorithms

Nielsen and Chuang, Chapter 5

We know that a quantum computer can efficiently simulate quantum dynamics

We know that it can efficiently simulate a classical computer

But what else can it do?

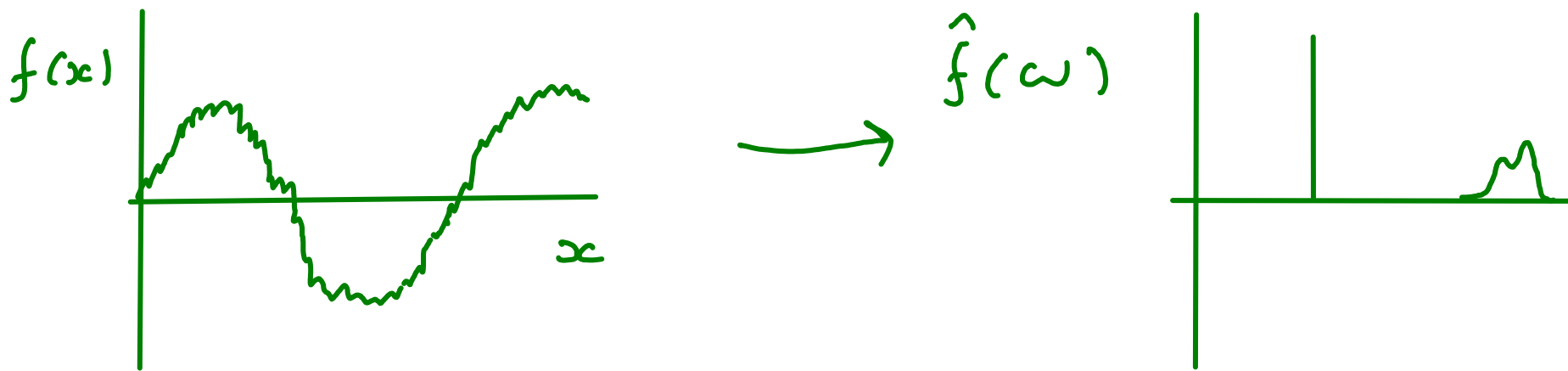
Today we will start to look at some algorithms that are unrelated to physics, all based on the quantum Fourier transform

# Quantum Fourier Transform

We know about the Fourier transform

$$\hat{f}(\omega) = \int_{-\infty}^{\infty} f(x) e^{-i2\pi x\omega} dx$$

This takes a function and outputs its spectrum



Useful in many applications, such as when a function has some periodicity that must be found and analyzed

A discrete version (DFT) can also be defined, where instead of a function we have a list of values (in a vector)

$$|f\rangle = \sum_{j=0}^{N-1} f_j |j\rangle$$

The transform acts on basis states according to

$$\rightarrow |\hat{j}\rangle = \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} e^{i2\pi jk/N} |k\rangle$$

And so acts on a general vector as

$$|f\rangle = \sum_{j=0}^{N-1} f_j |j\rangle \rightarrow |\hat{f}\rangle = \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} \sum_{j=0}^{N-1} f_j e^{i2\pi jk/N} |k\rangle$$

We consider the case that  $N=2^n$  and express the basis

$$|0\rangle, |1\rangle, |2\rangle, \dots, |N-1\rangle$$

In binary (and so as n qubits)

$$|0\dots 00\rangle, |0\dots 01\rangle, |0\dots 10\rangle, \dots, |1\dots 11\rangle$$

So for a general Z basis state

$$|j\rangle = |j_1 j_2 j_3 \dots j_n\rangle, \quad j = \sum_{l=1}^n j_l 2^{n-l}$$

Let's also consider the following notation for binary fractions (numbers less than 1 expressed in binary)

$$0.j_1 j_2 \dots j_m = \sum_{l=1}^m j_l 2^{-l}$$

Now let's see if we can simplify the Fourier transform basis states a bit

$$|\hat{j}\rangle = \frac{1}{2^{n/2}} \sum_{k=0}^{2^n-1} e^{i2\pi jk/2^n} |k\rangle = \frac{1}{2^{n/2}} \sum_{k_1=0}^1 \dots \sum_{k_n=0}^1 e^{i2\pi j \left( \sum_{l=1}^n k_l 2^{-l} \right)} |k_1 k_2 \dots k_n\rangle$$

Convert  $k$  into binary

$$k = \sum_{l=1}^n k_l 2^{n-l} \quad \therefore \frac{k}{2^n} = \sum_{l=1}^n k_l 2^{-l}$$

$$= \frac{1}{2^{n/2}} \sum_{k_1=0}^1 \dots \sum_{k_n=0}^1 \bigotimes_{l=1}^n e^{i2\pi j k_l 2^{-l}} |k_l\rangle$$

$$= \frac{1}{2^{n/2}} \bigotimes_{l=1}^n \left[ \sum_{K=0}^1 e^{i2\pi j K 2^{-l}} |K\rangle \right]$$

$$= \bigotimes_{l=1}^n \frac{|0\rangle + e^{i2\pi j 2^{-l}} |1\rangle}{\sqrt{2}}$$

Using

$$\prod_{x=1}^n \sum_y f(x,y) = \sum_{y_1 \dots y_n} \prod_x f(x,y)$$

So it turns out to be a product state

$$|\hat{j}\rangle = \bigotimes_{l=1}^n \frac{|0\rangle + e^{i2\pi j 2^{-l}} |1\rangle}{\sqrt{2}}$$

Let's also convert  $j$  to binary

$$j = \sum_{k=1}^n j_k 2^{n-k} \quad \therefore \frac{j}{2^l} = \sum_{k=1}^n j_k 2^{n-k-l} = \sum_{k=1}^{n-l} j_k 2^{(n-l)-k} + \sum_{k=n-l+1}^n j_k 2^{n-k-l}$$

$$\left( \begin{array}{l} -k' = n-k-l \\ \therefore k' = k - (n-l) \\ k = k' + (n-l) \\ = n - (l-k') \end{array} \right)$$

$$\begin{aligned} &= \text{integer} + \sum_{k'=1}^l j_{n-(l-k')} 2^{-k'} \\ &= \text{integer} + 0 \cdot j_{n-(l-1)} j_{n-(l-2)} \dots j_n \end{aligned}$$

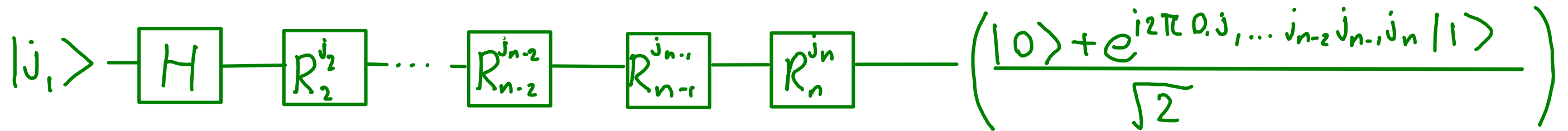
$$\therefore e^{i2\pi j 2^{-l}} = e^{i2\pi 0 \cdot j_{n-(l-1)} j_{n-(l-2)} \dots j_n}$$

$$|\hat{j}\rangle = \left( \frac{|0\rangle + e^{i2\pi 0 \cdot j_n} |1\rangle}{\sqrt{2}} \right) \otimes \left( \frac{|0\rangle + e^{i2\pi 0 \cdot j_{n-1} j_n} |1\rangle}{\sqrt{2}} \right) \otimes \dots \otimes \left( \frac{|0\rangle + e^{i2\pi 0 \cdot j_1 \dots j_n} |1\rangle}{\sqrt{2}} \right)$$

This product representation allows us to see how to perform the DFT on a quantum computer

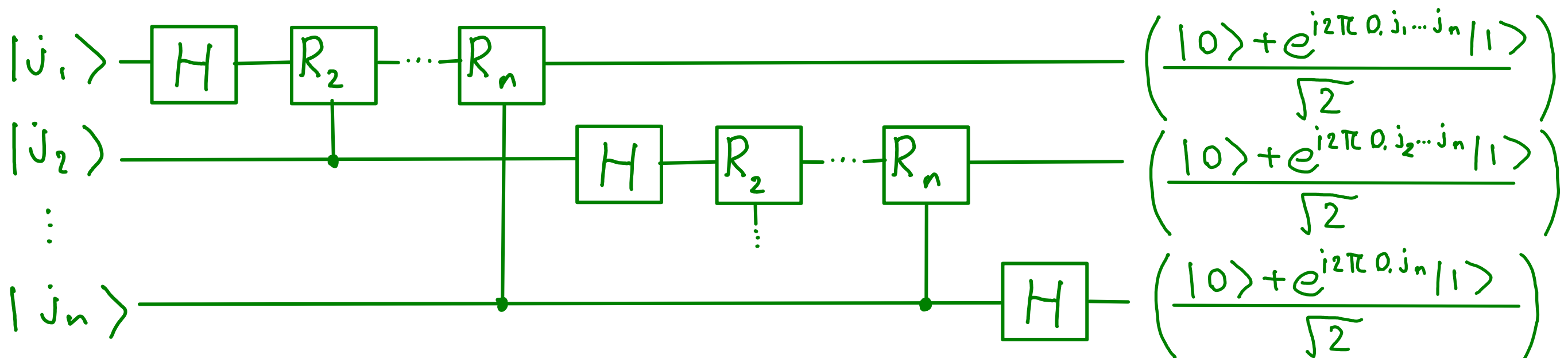
For the last qubit we could use

$$R_k = \begin{pmatrix} 1 & 0 \\ 0 & e^{i2\pi/2^k} \end{pmatrix}$$



Better to use controlled ops so we can deal with a superposition of different  $j$ 's

We then find that the circuit



Performs the FT (and reverses qubit order)

This clearly requires  $O(n^2)$  gates

So the DFT (and its inverse) can be implemented on a quantum state efficiently by a quantum computer

The fastest known classical algorithm requires  $O(n2^n)$

So can we use quantum computers to do fast DFTs?

Yes and No

'No' because preparing a general state to be transformed is inefficient, even if the transformation itself is efficient

So we cannot use it to do a DFT on any vector that we be interested from a real-world problem

'Yes' because it can be used as a component in larger quantum algorithms that do have efficient read-in and read-out

# Phase Estimation

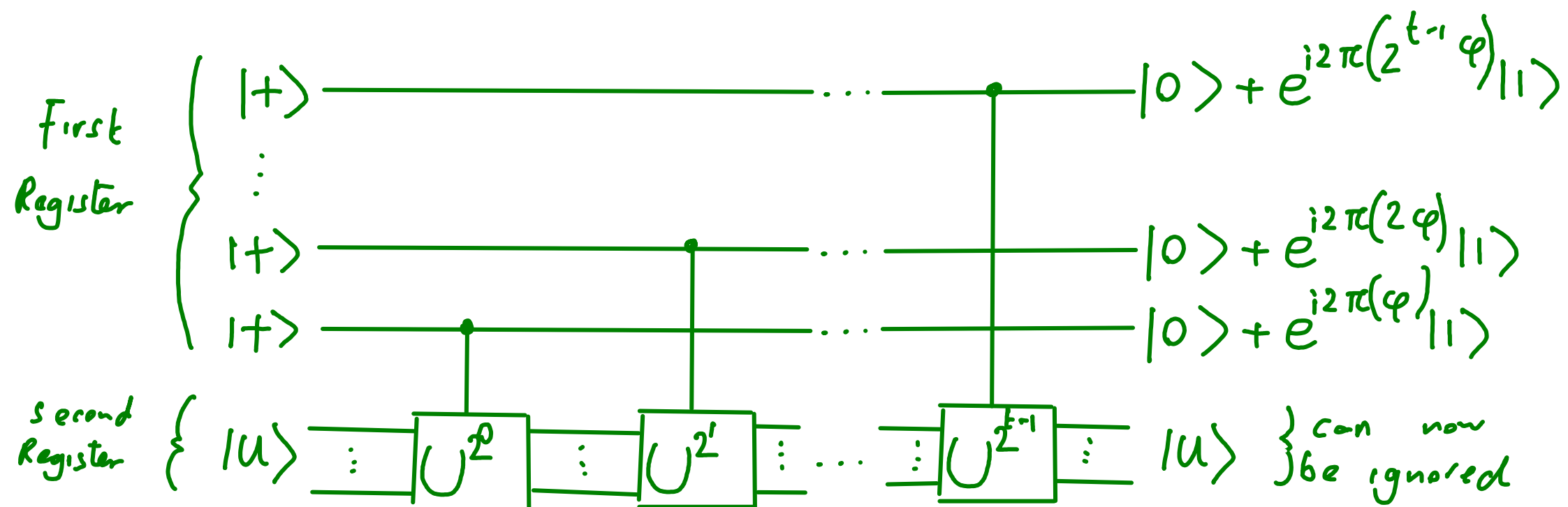
Consider a unitary operation for which we know an eigenstate, and wish to find out the corresponding eigenvalue

$$U|u\rangle = e^{i2\pi\varphi}|u\rangle \quad \varphi = 0.\varphi_1\varphi_2\dots\varphi_t$$

$t$  is # bits required to express  $\varphi$

Assume that we have the ability to prepare the eigenstate and apply a controlled-U

This means we can apply the circuit





Outcome for the first register is

$$\bigotimes_{l=t}^1 \frac{|0\rangle + e^{i2\pi\varphi 2^{l-1}} |1\rangle}{\sqrt{2}}$$

Let's change our variable a little

$$\varphi = 0.\varphi_1\varphi_2\dots\varphi_t \quad \therefore \quad \varphi_1\varphi_2\dots\varphi_t = 2^t \varphi = \phi \quad \therefore \quad \varphi = \phi 2^{-t}$$

$$\bigotimes_{l=t}^1 \frac{|0\rangle + e^{i2\pi\phi 2^{l-t}} |1\rangle}{\sqrt{2}} = \bigotimes_{l'=1}^t \frac{|0\rangle + e^{i2\pi\phi 2^{-l'}} |1\rangle}{\sqrt{2}} = |\hat{\phi}\rangle \quad (l' = 1+t-l)$$

So the outcome is the FT of the state  $|\phi\rangle = |\varphi_1\varphi_2\dots\varphi_t\rangle$

$$\left( \text{recall } |\hat{j}\rangle = \bigotimes_{l=1}^n \frac{|0\rangle + e^{i2\pi j 2^{-l}} |1\rangle}{\sqrt{2}} \right)$$

Performing the inverse FT and measuring the state in the Z basis then gives the binary representation of the phase

$$\text{Final state: } |\phi\rangle \otimes |u\rangle$$

Note that this method assumes

the phase can be written using a finite number of bits,  $t$   
we know what  $t$  is (or at least an upper bound)

In general, this is not the case

However, even if the  $t$  we use is too small, it will give a  
good approximation

To get the phase accurate to  $n$  bits with high probability, we  
need to use

$$t = n + \left\lceil \log \left( 2 + \frac{1}{2\varepsilon} \right) \right\rceil, \text{ accurate with probability } 1 - \varepsilon$$

Which is efficient

But can phase estimation be used for anything useful?

## Order Finding

Consider the positive integers  $x$  and  $N$  for which  $x < N$  and there are no common factors

What is the smallest possible integer  $r$  such that

$$x^r = 1 \pmod{N}$$

This is called the order of  $x$  modulo  $N$

It is believed that no  $\text{poly}(L)$  algorithm exists to compute this on a classical computer, where  $L$  is the number of bits needed to specify  $N$

$$L = \lceil \log N \rceil \therefore 2^L \geq N$$

To compute it with a quantum computer, consider the operator

$$U = \sum_{j=0}^{2^L-1} |x^j \pmod{N} \rangle \langle j|$$

Where we use the convention

$$x^j \pmod{N} = j \quad \text{for} \quad N \leq j \leq 2^L-1$$

The eigenstates of this are

$$|U_s\rangle = \frac{1}{\sqrt{r}} \sum_{k=0}^{r-1} \exp\left[-\frac{i2\pi sk}{r}\right] |x^k \bmod N\rangle$$

With eigenvalues

$$e^{i2\pi\varphi(s)} = \exp\left[\frac{i2\pi s}{r}\right] \quad \therefore \varphi(s) = \frac{s}{r}$$

If we can use phase estimation to find these, we can find  $r$

For that we need to efficiently perform the controlled-U's

Efficient methods exist for this

We also need to prepare eigenvalues of U

This cannot be done efficiently, so is there another option?

Consider the superposition of the first r eigenstates

$$\begin{aligned}\frac{1}{\sqrt{r}} \sum_{s=0}^{r-1} |u_s\rangle &= \frac{1}{r} \sum_{s=0}^{r-1} \sum_{k=0}^{r-1} \exp\left[-\frac{i2\pi sk}{r}\right] |x^k \bmod N\rangle \\ &= \frac{1}{r} \sum_{k=0}^{r-1} \left( \sum_{s=0}^{r-1} \exp\left[-\frac{i2\pi sk}{r}\right] \right) |x^k \bmod N\rangle\end{aligned}$$

$$\sum_{s=0}^{r-1} \exp\left[-\frac{i2\pi sk}{r}\right] = \delta_k r, \text{ due to the properties of sums of roots of unity (next slide)}$$

$$\therefore \frac{1}{\sqrt{r}} \sum_{s=0}^{r-1} |u_s\rangle = |x^0\rangle = |1\rangle = |00\dots 01\rangle$$

This can be efficiently prepared

Roots of unity can be written

$$\omega = \exp\left[-\frac{i2\pi}{r}\right] \quad \therefore \omega^r = \omega^0 = 1$$

Summing all powers of roots of unity gives zero

$$\sum_{s=0}^{r-1} \omega^s = 0$$

For example

$$r=2: \quad \omega = -1, \quad \omega^0 + \omega = 1 - 1 = 0$$

$$r=4: \quad \omega = i, \quad \omega^0 + \omega + \omega^2 + \omega^3 = 1 + i - 1 - i = 0$$

The same is true integer powers of roots of unity

$$\sum_{s=0}^{r-1} \omega^{ks} = 0 \quad \text{for } k \in \{1, \dots, r-1\}$$

$$\text{z.B. } r=4, k=3: \quad \omega^k = -i, \quad (\omega^k)^0 + \omega^k + \omega^{2k} + \omega^{3k} = 1 - i - 1 + i = 0$$

But things are obviously different if the power is zero (or r)

$$\sum_{s=0}^{r-1} \omega^0 = r$$

Putting it all together

$$\sum_{s=0}^{r-1} \omega^{ks} = \delta_{kr}$$

If this is used as the input state of the second register and the phase estimation algorithm is applied, the final state is

$$\sum_{s=0}^{r-1} |\varphi(s)\rangle \otimes |u_s\rangle, \quad \varphi(s) \approx \frac{s}{r}$$

By applying the method  $O(r) = O(L)$  times, we can find (approximations of) all the phases  $\varphi(s)$ ,  $s = 0, \dots, r-1$

But since  $r = 2^{O(L)}$  this would be inefficient

Fortunately we need only one (randomly chosen) phase

$$\varphi \approx \frac{s}{r}$$

For unknown  $s$  and  $r$

These unknowns can be determined by the continued fractions algorithm if the phase is sufficiently accurate

## The relevant theorem

If  $\left| \frac{s'}{r'} - \varphi \right| \leq \frac{1}{2r'^2}$ , for 2 bit integers  $s'$  and  $r'$   
with no common factors

then the continued fractions algorithm can compute  $s'$  and  $r'$  from  $\varphi$  in  $O(L^3)$  time

Since the phase is accurate to  $n$  bits with, we have

$$\left| \frac{s}{r} - \varphi \right| \leq 2^{-n}$$

So for the theorem to apply we require

$$2^{-n} \leq \frac{1}{2r^2} \quad \therefore n \geq 1 + 2 \log r \geq 2L + 1 \quad \text{since } r \leq N \leq 2^L$$

For a good enough approximation, we need to use

$$t = 2L + 1 + \lceil \log(2 + \frac{1}{\epsilon}) \rceil = O(L)$$

bits on the first register, which is efficient



Problem:  $s$  and  $r$  may have common factors, so the  $s'$  and  $r'$  output by continued fractions may not be the numbers we want

$$\varphi = \frac{s}{r} = \frac{s'}{r'} \quad s' < s, r' < r$$

But recall the definition of  $r$ . It is the smallest integer such that  $x^r = 1 \pmod{N}$

We can efficiently (and classically) check if  $x^{r'} = 1 \pmod{N}$

If it is, we know that  $r' = r$

If not, we can try again until we get it right

This will certainly occur if  $s$  is prime, which occurs with probability  $O\left(\frac{1}{\log r}\right) = O\left(\frac{1}{\log N}\right)$

So only  $O(\log N)$  repetitions are required until  $r$  is found

Better methods with only  $O(1)$  repetitions also exist

Another problem: Approximation of the phase is bad with probability  $\varepsilon$

This probability is efficiently suppressed by using a large enough register

$$t = O\left(\log\left(2 + \frac{1}{\varepsilon}\right)\right) = O(\log(M))$$

where  $M = \frac{1}{\varepsilon}$  is the expected # successful runs before an error

So we can find  $r$  efficiently with a quantum computer using

modular exponentiation  $O(L^3)$  (or less)

Fourier transform  $O(L)$

continued fractions  $O(L^3)$

repetitions  $O(1)$

Total complexity is  $(O(L^3) + O(L) + O(L^3))O(1) = O(L^3)$