## CMPE 598 - Lecture Notes

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## 1. Shor's Algorithm for Factorization

Given a positive integer, find its factors.

There exists a fast classical algorithm for detecting whether the number is prime. If so, problem solved.

There exists a fast classical algorithm for detecting whether the number is a power (i.e. of the form  $a^b$  for b > 1).

 $x=a^b$ 

log x = b \* log a so b cannot be bigger than logx.

**Binary Search:** If you can find a factor, you can find all other factors as well using the same method repeatedly.

Assume you are given a number pq where p and q are primes.

We want to factor a large integer N.

FACTORING is reduced to finding a non-trivial square root of 1 modulo N.
y<sup>2</sup> ≡ 1 (modN) (y ∈ 1, 2, ..., N − 1)
if N = 15
trivial square roots of 1 (mod 15)

 $1^2 \equiv 1 \pmod{15}$  not exciting  $(-1)^2 \equiv 1 \pmod{15}$  not exciting  $14^2 \equiv 1 \pmod{15}$  not exciting  $-1 \equiv 14 \pmod{15}$ 

non-trivial square root of 1 (mod 15)

 $4^2 \equiv 1 \ (mod15)$ 

If y is a non-trivial square root of 1 mod N,

Then N divides  $y^2 - 1 = (y + 1)(y - 1)$ , but N does **not** divide neither y - 1

nor y + 1. So this means that gcd(y - 1, N) > 1 because if y - 1 and N were relatively prime, then since N divides (y-1)(y+1), it would have to divide (y+1).

 Finding such a root is reduced to computing the order of a random integer modulo N.

Pick a random number X (mod N).

The order of X (mod N) is the smallest number r such that  $X^r \equiv 1 \pmod{N}$ .

**Example:** The order of 2 mod 15 is 4.  $2^1 = 2, 2^2 = 4, 2^3 = 8, 2^4 = 1, 2^5 = 2, 2^6 = 4, 2^7 = 8, 2^8 = 1, \dots$ 

• The order of an integer is precisely the period of a particular periodic superposition.

"Period = Order"

• And, periods of superpositions can be found by the quantum FFT.

Classical FFT's input is an M-dimensional, complex valued vector  $\alpha$  (where M is a power of 2, say  $2^m$ ) and its output is an M-dimensional, complex valued vector  $\beta$ ;

$$\begin{bmatrix} \beta_{0} \\ \beta_{1} \\ \vdots \\ \beta_{M-1} \end{bmatrix} = \frac{1}{\sqrt{M}} \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & W & W^{2} & \dots & W^{M-1} \\ 1 & W^{2} & W^{4} & \dots & W^{2(M-1)} \\ 1 & W^{3} & W^{6} & \dots & W^{3(M-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & W^{(M-1)} & W^{2(M-1)} & \dots & W^{(M-1)(M-1)} \end{bmatrix} \begin{bmatrix} \alpha_{0} \\ \alpha_{1} \\ \vdots \\ \alpha_{M-1} \end{bmatrix}$$
(1)

where W is a complex  $M^{th}$  root of unity. FFT runs in O(M \* log M) steps. **Input:** A superposition of m = log M qubits  $|\alpha\rangle = \sum_{j=0}^{M-1} \alpha_j |j\rangle$ .

$$\begin{bmatrix} \alpha_{0} \\ \alpha_{1} \\ \vdots \\ \alpha_{M-1} \end{bmatrix} = \alpha_{0} \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + \alpha_{1} \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + \alpha_{2} \begin{bmatrix} 0 \\ 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} + \dots + \alpha_{M-1} \begin{bmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$
(2)

Method: We'll use  $O(m^2)$  quantum operations to obtain the superposition.  $|\beta{>}=\sum_{j=0}^{M-1}\beta_j\ |j{>}$ 

**Output:** A random m-bit number j (i.e.  $0 \le j \le M - 1$ ), from the probability distribution  $Pr[j] = |\beta_j|^2$ .

Suppose the input to quantum Fourier sampling is periodic with period k, for some k that divides M. Then the output will be a multiple of  $\frac{M}{k}$ , and it is equally likely to be any of the k multiples of  $\frac{M}{k}$ .

$$|\alpha\rangle = \begin{pmatrix} \alpha_0 \\ \alpha_1 \\ \vdots \\ \alpha_{M-1} \end{pmatrix}$$
(3)

 $|\alpha\rangle$  is such that  $\alpha_j = \alpha_i$  whenever  $i = j \mod k$  where k is a particular integer that divides M. So there are  $\frac{M}{k}$  repetitions of same sequence  $(\alpha_0, \alpha_1, ..., \alpha_{k-1})$  of length k.

And suppose exactly one of these k numbers is non-zero, say  $\alpha_i$ .

Suppose the vector  $|\alpha\rangle = (\alpha_0, \alpha_1, ..., \alpha_{M-1})^T$  is periodic with period k with no offset (that is, the non-zero terms are  $\alpha_0, \alpha_k, \alpha_{2k}, ...$ ). Thus  $|\alpha\rangle = \sum_{j=0}^{\frac{M}{k}-1} \sqrt{\frac{k}{M}} |jk\rangle$ .

Claim:

$$|\beta\rangle = \frac{1}{\sqrt{k}} \sum_{j=0}^{k-1} |\frac{jM}{k}\rangle \tag{4}$$

In the input vector, the coefficient of  $\alpha_l$  is  $\sqrt{\frac{k}{M}}$  if k divides l, and zero otherwise. The  $j^{th}$  coefficient  $|\beta\rangle$  is

$$\beta_j = \frac{1}{\sqrt{M}} \sum_{l=0}^{M-1} w^{jl} \alpha_l = \frac{\sqrt{k}}{M} \sum_{i=0}^{\frac{M}{k}-1} w^{jik}$$
(5)

So this sum is the geometric series  $1 + w^{jk} + w^{2jk} + \dots$  containing  $\frac{M}{k}$  terms and with ratio  $w^{jk}$ . There are two cases. If the ratio is exactly 1, which happens if  $jk \equiv 0 \mod M$ , then the sum of the series is just the number of terms. If the ratio isn't 1, apply the usual formula for geometric series to find that the sum is

$$\frac{1 - w^{jk(\frac{M}{k})}}{1 - w^{jk}} = \frac{1 - w^{jM}}{1 - w^{jk}} = 0$$
(6)

So  $\beta_j = \frac{1}{\sqrt{k}}$  is M divides jk, and is zero otherwise. Also works (with little modification) for the case where the offset is non-zero.

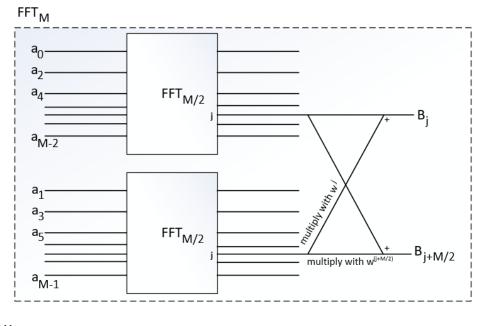
Suppose s independent samples are drawn uniformly from  $\{0, \frac{M}{k}, \frac{2M}{k}, ..., \frac{(k-1)M}{k}\}$ . Then, with probability at least  $1 - \frac{k}{2^s}$ , the greatest common divisor of these samples is  $\frac{M}{k}$ .

**Proof:** The only way this can fail is if all samples are multiples of  $j\frac{M}{k}$ , for some j > 1. So, fix any integer  $j \ge 2$ . The chance that a particular sample is a multiple of  $j\frac{M}{k}$  is at most  $\frac{1}{j} \le \frac{1}{2}$ , so the chance that **all samples** are multiples of  $j\frac{M}{k}$  is at most  $\frac{1}{2^s}$ . The probability that this bad thing will happen for some  $j \le k$  is at most  $k\frac{1}{2^s}$ , since these are k candidates for the number j.

How does the classical FFT work?

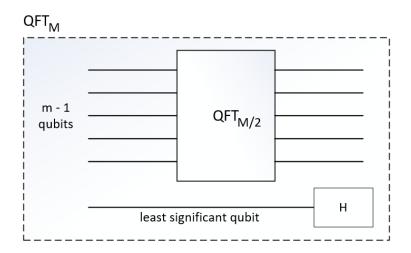
"Divide & Conquer"

from input  $(\alpha_0, \alpha_1, ..., \alpha_{M-1})^T$  to output  $(\beta_0, \beta_1, ..., \beta_{M-1})^T$ 



 $w^{\frac{M}{2}} = -1.$ 

In the quantum version, the input is now encoded in the  $2^m$  amplitudes of m = logM qubits. So the decomposition of the inputs to evens and odds is determined by the least significant qubit. We will design a quantum circuit (subroutine)  $QFT_M$ .  $QFT_{\frac{M}{2}}$  will be applied to the remaining m-1 qubits.



$$H = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$
(7)

$$Other\ lines = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{8}$$

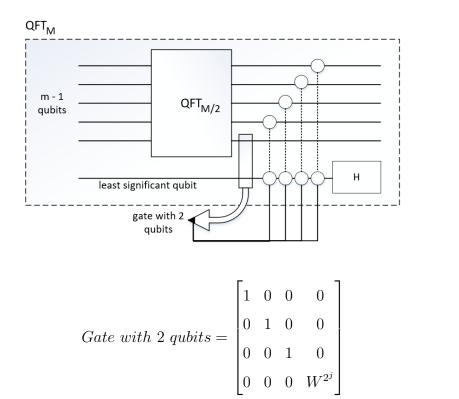
$$A = \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & & & & \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix} \otimes \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 & 0 & \dots & 0 & 0 \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \dots & 0 & 0 \\ 0 & 0 & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & \dots & 0 & 0 \\ \vdots & & & & & \\ 0 & 0 & 0 & 0 & 0 & \dots & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ 0 & 0 & 0 & 0 & 0 & \dots & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$
(9)

$$\begin{array}{l} 01100 \ 0 \rightarrow {\rm even} \\ 01100 \ 1 \rightarrow {\rm odd} \end{array}$$

$$\begin{pmatrix}
\alpha_{y_0} : \text{even} \\
\alpha_{y_1} : \text{odd}
\end{pmatrix}$$
(10)

$$A\begin{pmatrix}\alpha_{y_0}\\\alpha_{y_1}\end{pmatrix} = \begin{pmatrix}\frac{\alpha_{y_0} + \alpha_{y_1}}{\sqrt{2}}\\\frac{\alpha_{y_0} - \alpha_{y_1}}{\sqrt{2}}\end{pmatrix}$$
(11)

For each j, an operation is done in the classical FFT on the  $(\frac{M}{2} + j)^{th}$  wire. If j is represented by the m-1 bits  $j_1, j_2, ..., j_{m-1}$ , then  $w_j = \prod_{l=1}^{m-1} w^{2^{j_l}}$ . Ex: m = 3, m - 1 = 2,  $j_2 j_1$ .



(12)