

# LECTURE 6

## PHASE ESTIMATION, SHOR'S ALGORITHM AND HIDDEN SUBGROUP PROBLEM

Quantum Information and Computing

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Presentation of Shor's algorithm and hidden Abelian subgroup problem!

It will rely (partly) on:

- ▶ phase estimation and consequences: **QFT** over finite Abelian groups and order finding

1. Phase Estimation
2. Application 1: Quantum Fourier Transform on  $\mathbb{Z}/N\mathbb{Z}$  and any Finite Abelian Group
3. Application 2: Order Finding
4. Shor's Algorithm
5. Hidden Subgroup Problem (HSP)

# PHASE ESTIMATION

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## Phase estimation:

- **Input:** a unitary  $U$  and an **eigenstate**  $|u\rangle$ :

$$U |u\rangle = e^{2i\pi\varphi} |u\rangle$$

- **Output:**  $\varphi \in [0, 1)$ , i.e., the knowledge of the associate eigenvalue of  $|u\rangle$

→ Essential for computing  $\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}$  and Shor's algorithm!

## Proposition:

We can determine (by using  $\text{QFT}_{\mathbb{Z}/2^t\mathbb{Z}}$ ) the first  $n$  bits of  $\varphi$  with probability  $1 - \epsilon$  using

$$O(t^2) \text{ elementary gates where } t = n + \left\lceil \log \left( 2 + \frac{1}{2\epsilon} \right) \right\rceil$$

→  $n$  bits of  $\varphi$  with probability  $1 - e^{-cn}$  but working in the space of  $t$ -qubits with  $t = O(n)$

## Notation:

Given  $j_1, j_2, \dots, j_m \in \{0, 1\}$ :

$$0.j_1j_2 \dots j_m \stackrel{\text{def}}{=} \frac{j_1}{2} + \frac{j_2}{4} + \dots + \frac{j_m}{2^m} = \sum_{i=1}^m \frac{j_i}{2^i}$$

## Example:

$$0.101 = \frac{1}{2} + \frac{1}{8} = 0.625, \quad 0.111 = \frac{1}{2} + \frac{1}{4} + \frac{1}{8} = 0.875 \quad \text{and} \quad 0.011 = \frac{1}{4} + \frac{1}{8} = 0.325$$

$$2^m 0.j_1j_2 \dots j_m = 2^{m-1}j_1 + 2^{m-2}j_2 + \dots + j_m = j_1 \dots j_m \in \llbracket 0, 2^m - 1 \rrbracket$$

(binary representation with  $m$  bits)

$$2^\ell 0.j_1j_2 \dots j_m = \underbrace{2^{\ell-1}j_1 + \dots + j_\ell}_{\in \mathbb{N}} + 0.j_{\ell+1} \dots j_m$$

$$\longrightarrow e^{2i\pi 2^\ell \cdot 0.j_1j_2 \dots j_m} = e^{2i\pi 0.j_{\ell+1} \dots j_m}$$

## PHASE ESTIMATION ALGORITHM

The quantum algorithm to determine the phase starts from ( $|u\rangle$  being the eigenstate)

$$|0^t\rangle |u\rangle$$

→  $t$  function of: (i) accuracy and (ii) probability we wish to be successful

Phase estimation, **two stages** algorithm:

1. Build the following quantum state:

$$\frac{1}{2^{t/2}} \left( |0\rangle + e^{2i\pi 2^{t-1}\varphi} |1\rangle \right) \otimes \left( |0\rangle + e^{2i\pi 2^{t-2}\varphi} |1\rangle \right) \otimes \dots \otimes \left( |0\rangle + e^{2i\pi 2^0\varphi} |1\rangle \right) \otimes |u\rangle$$

2. Apply the  $\text{QFT}_{z/2^t z}^{-1}$  to reach:

$$\approx \left| [2^t \varphi] \right\rangle \otimes |u\rangle = |\varphi_1 \dots \varphi_t\rangle \otimes |u\rangle$$

Does the first step remind you of something?

The controlled  $U^{2^j}$ -unitary:

$$|1\rangle |u\rangle \mapsto |1\rangle U^{2^j} |u\rangle = e^{2i\pi\varphi 2^j} |1\rangle |u\rangle$$

$$|0\rangle |u\rangle \mapsto |0\rangle |u\rangle$$

**Be careful:**  $U^{2^j} = \underbrace{U \cdots U}_{2^j \text{ iterates}}$ , in particular  $U^{2^j} |u\rangle \neq (U |u\rangle)^{2^j}$

The algorithm:

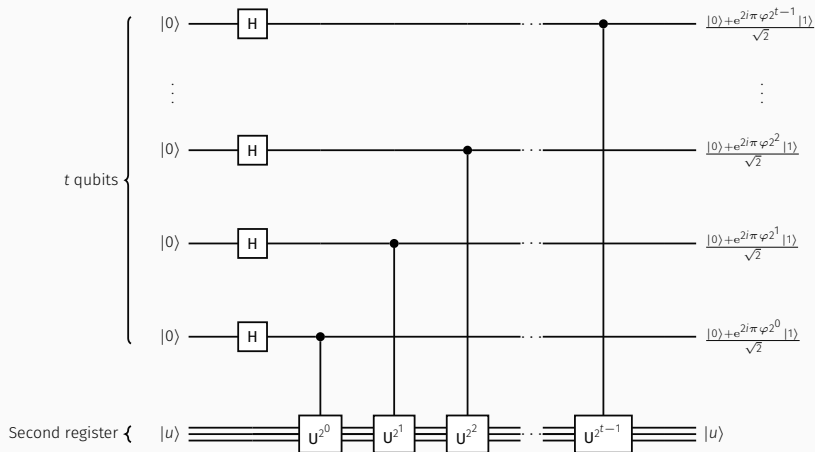
1. Start with  $|0^t\rangle |u\rangle$
2. Apply  $H^{\otimes t} \otimes I$
3. For  $i = 1$  to  $n$ :  
 apply the controlled  $U^{2^i}$ -gate to the  $i$ -th register

Resulting quantum state:

$$\frac{1}{2^{t/2}} \left( |0\rangle + e^{2i\pi 2^{t-1}\varphi} |1\rangle \right) \otimes \left( |0\rangle + e^{2i\pi 2^{t-2}\varphi} |1\rangle \right) \otimes \cdots \otimes \left( |0\rangle + e^{2i\pi 2^0\varphi} |1\rangle \right) \otimes |u\rangle$$



# THE QUANTUM CIRCUIT



But what is the cost for computing  $U^{2^j}$ ? Is it  $2^j$ ?

Given an arbitrary  $U$ , computing the controlled- $U^{2^j}$  costs  $2^j \times \text{Cost}(U) \dots$

### An example:

If  $f : \{0, 1\}^n \rightarrow \{0, 1\}^n$  is a **bijection** efficiently computable, then the unitary

$$U : |x\rangle \mapsto |f(x)\rangle$$

is efficiently computable. But, is

$$U^{2^j} : |x\rangle \mapsto |f^{2^j}(x)\rangle \quad (f^{2^j} \text{ composition, not exponentiation})$$

efficiently computable? It depends of the particular shape of  $f \dots$

→ Does it imply that phase estimation has an exponential cost?

## BE CAREFUL (I)

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→ Does it imply that phase estimation has an exponential cost?

**No... or Yes... It depends!**

As in the classical case: computing  $f^{2^j}$  is expensive ( $2^j \times \text{Cost}(f)$ ) except for some functions...

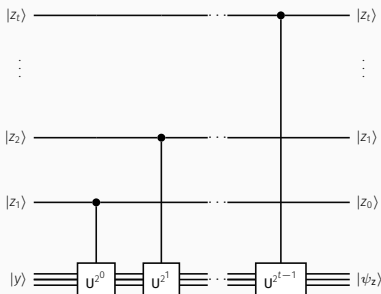
**Phase estimation: be careful, in the general case**

Computing  $U^{2^j}$  costs  $2^j \times \text{Cost}(U)$  unless one succeeds to use the particular shape of  $U \dots$

All the game in phase estimation lies in computing efficiently (designing an efficient circuit)

$$z_1, \dots, z_t \in \{0, 1\}^t, \quad \mathbf{V} : |z_1 \dots z_t\rangle |u\rangle \mapsto |z_1 \dots z_t\rangle |\psi_z\rangle$$

where  $\mathbf{V}$  is the following unitary



### Phase estimation: be careful

Computing  $U^{2^j}$  costs  $2^j \times \text{Cost}(U)$  unless one succeeds to use the particular shape of  $U \dots$

→ Let us take a look at the classical case!

## CLASSICAL EXPONENTIATION: FAST OR TERRIBLY SLOW, CHOOSE!

What is the cost to compute  $x^{2^j}$ ? Is it  $2^j$ ?

## CLASSICAL EXPONENTIATION: FAST OR TERRIBLY SLOW, CHOOSE!

What is the cost to compute  $x^{2^j}$ ? Is it  $2^j$ ?

Of course not. . . **fast exponentiation**

- Stupid algorithm:  $y = 1$  and then  $2^j$  times:  $y \leftarrow yx$ ; output  $y$
- Clever algorithm: if  $j$  even,  $y \leftarrow 2^{2^{j/2}}$ ; outputs  $y^2$ ; otherwise  $y \leftarrow 2^{2^{(j-1)/2}}$  then outputs  $2y^2$ .

→ To compute  $2^{2^{j/2}}$  or  $2^{2^{(j-1)/2}}$ : **recursive call**

Cost?

- Stupid algorithm:  $2^j$  multiplications!
- Clever algorithm:  $\log 2^j = j$  recursive calls and 1 or 2 multiplications for each call

→ It costs  $j \times$   $\underbrace{j^2}_{\text{cost of squaring}}$

→ The “clever” algorithm is exponentially faster. . .

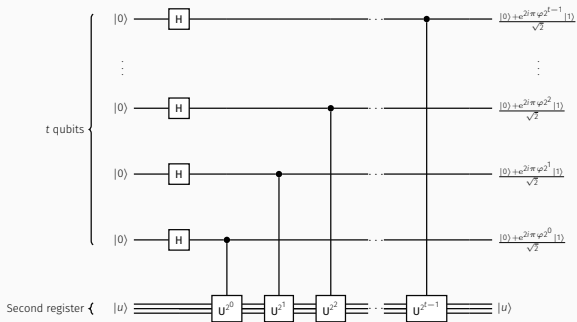
**Be careful:** we have used the particular shape of  $x \mapsto x^{2^j}$

Usually  $f^{2^j}(x) \neq f^{2^{j/2}}(x)^2$  but  $f^{2^j}(x) = f^{2^{j/2}}(f^{2^{j/2}}(x))$

# REBOOT: ANALYSIS OF THE FIRST STEP IN PHASE ESTIMATION

$$U|u\rangle = e^{2i\pi\varphi}|u\rangle \implies U^{2^j}|u\rangle = e^{2i\pi 2^j\varphi}|u\rangle$$

$$C-U^{2^j}|0\rangle|u\rangle = |0\rangle|u\rangle \quad \text{and} \quad C-U^{2^j}|1\rangle|u\rangle = e^{2i\pi 2^j\varphi}|1\rangle|u\rangle$$



- First Step:

$$\frac{1}{\sqrt{2^t}} (|0\rangle + |1\rangle)^{\otimes t} \otimes |u\rangle$$

- Second Step:

$$\frac{1}{\sqrt{2^t}} \left( |0\rangle + e^{2i\pi 2^{t-1}\varphi} |1\rangle \right) \otimes \left( |0\rangle + e^{2i\pi 2^{t-2}\varphi} |1\rangle \right) \otimes \dots \otimes \left( |0\rangle + e^{2i\pi 2^0\varphi} |1\rangle \right) \otimes |u\rangle$$

Suppose that

$$\varphi = 0.\varphi_1 \dots \varphi_t$$

See Lecture 5:

$$\begin{aligned} & \frac{1}{2^{t/2}} \left( |0\rangle + e^{2i\pi 2^{t-1}\varphi} |1\rangle \right) \otimes \left( |0\rangle + e^{2i\pi 2^{t-2}\varphi} |1\rangle \right) \otimes \dots \otimes \left( |0\rangle + e^{2i\pi 2^0\varphi} |1\rangle \right) \otimes |u\rangle \\ &= \frac{1}{2^{t/2}} \left( |0\rangle + e^{2i\pi 0.\varphi_t} |1\rangle \right) \otimes \left( |0\rangle + e^{2i\pi 0.\varphi_{t-1}\varphi_t} |1\rangle \right) \otimes \dots \otimes \left( |0\rangle + e^{2i\pi 0.\varphi_1\varphi_2 \dots \varphi_t} |1\rangle \right) \otimes |u\rangle \\ &= \text{QFT}_{\mathbb{Z}/2^t\mathbb{Z}} |\varphi_1 \dots \varphi_t\rangle \end{aligned}$$

Applying  $\text{QFT}_{\mathbb{Z}/2^t\mathbb{Z}}^{-1}$  leads to:

$$|\varphi_1 \dots \varphi_t\rangle \longrightarrow \text{we have recovered } \varphi!$$

→ But what does happen if  $\varphi = 0.\varphi_1 \dots \varphi_t \varphi_{t+1} \varphi_{t+2} \dots \varphi_\ell \dots$  ?



## Important convention:

When working in  $\mathbb{Z}/2^t\mathbb{Z}$  the considered Hilbert space is  $\underbrace{\mathbb{C}^2 \otimes \dots \otimes \mathbb{C}^2}_{t \text{ times}}$  and for all  $x \in \mathbb{Z}/2^t\mathbb{Z}$ ,

$$|x\rangle \stackrel{\text{def}}{=} |x_1 \dots x_t\rangle$$

where  $x_1 \dots x_t$  being the binary decomposition of  $x$ , i.e.,  $x = \sum_{k=1}^t x_k 2^{t-k}$

$$\begin{aligned} & \frac{1}{2^{t/2}} \left( |0\rangle + e^{2i\pi 2^{t-1}\varphi} |1\rangle \right) \otimes \left( |0\rangle + e^{2i\pi 2^{t-2}\varphi} |1\rangle \right) \otimes \dots \otimes \left( |0\rangle + e^{2i\pi 2^0\varphi} |1\rangle \right) \otimes |u\rangle \\ &= \frac{1}{2^{t/2}} \sum_{\ell=0}^{2^t-1} e^{2i\pi \ell \varphi} |\ell\rangle \otimes |u\rangle \end{aligned}$$

## Important convention:

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Applying  $\text{QFT}_{\mathbb{Z}/2^t\mathbb{Z}}^{-1} \otimes \text{Id}$  leads to:

$$\text{QFT}_{\mathbb{Z}/2^t\mathbb{Z}}^{-1} \otimes \text{Id} \left( \frac{1}{2^{t/2}} \sum_{\ell=0}^{2^t-1} e^{2i\pi \ell \varphi} |\ell\rangle \otimes |u\rangle \right) = \frac{1}{2^t} \sum_{k,\ell=0}^{2^t} e^{2i\pi \ell (\varphi - \frac{k}{2^t})} |k\rangle \otimes |u\rangle$$

Best approximation of  $\varphi$  for the first  $t$  bits:

Let  $b \in \llbracket 0, 2^t - 1 \rrbracket$  be such that  $b/2^t = 0.b_1 \dots b_t$  and

$$0 \leq \varphi - \frac{b}{2^t} \leq 2^{-t} \quad : b/2^t \text{ best } t \text{ bits approximation of } \varphi$$

Up to now we have the following quantum state:

$$\frac{1}{2^t} \sum_{k, \ell=0}^{2^t-1} e^{2i\pi\ell(\varphi - \frac{k}{2^t})} |k\rangle |u\rangle$$

Let  $\alpha_j$  be the amplitude of  $(b + j \bmod 2^t)$  in the first register:

$$|\alpha_j|^2 = \frac{1}{2^{2t}} \left| \sum_{\ell=0}^{2^t-1} \left( e^{2i\pi\left(\varphi - \frac{b+j}{2^t}\right)\ell} \right) \right|^2$$

Measure (see Exercise Session):

Let  $m \in \{0, 1\}^t$  be the outcome after measuring the first register in the computational basis (defining an integer in  $\llbracket 0, 2^t - 1 \rrbracket$ ). We have

$$\mathbb{P}(|b - m| > \alpha) \leq \frac{1}{2(\alpha - 1)}$$

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Measure:

Let  $m$  be the outcome after measuring the first register in the computational basis. We have

$$\mathbb{P}(|b - m| > \alpha) \leq \frac{1}{2(\alpha - 1)}$$

→ Determining  $\varphi$  with  $n$  bits of accuracy thanks to the **output of the measure  $m$**  ( $t > n$ ):

$$\left| \frac{b}{2^t} - \frac{m}{2^t} \right| < 2^{-n}$$

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$$\left| \frac{b}{2^t} - \frac{m}{2^t} \right| < 2^{-n}$$

→ Therefore: choosing  $\alpha = 2^{t-n} - 1$  in the above probability. . .

But to reach a **probability of success  $\geq 1 - \epsilon$** :

$$\frac{1}{2(\alpha - 1)} = \frac{1}{2(2^{t-n} - 2)} \leq \epsilon \iff t = n + \left\lceil \log \left( 2 + \frac{1}{2\epsilon} \right) \right\rceil$$

## Phase estimation:

- **Input:** a unitary  $U$  and an **eigenstate**  $|u\rangle$ :

$$U |u\rangle = e^{2i\pi\varphi} |u\rangle$$

- **Output:**  $\varphi \in [0, 1)$ , i.e., the knowledge of the associate eigenvalue of  $|u\rangle$

## Proposition:

The phase estimation (before the last step measuring in the computational basis) computes,

$$|0^t\rangle |u\rangle \mapsto |\psi_u\rangle |u\rangle$$

such that  $|\psi_u\rangle$  is an **approximation of  $\varphi$** , i.e. when measuring the first register we obtain  $\tilde{\varphi} \in \{0, 1\}^t$  admitting the same first  $n$  bits than  $\varphi$  with probability  $\geq 1 - \varepsilon$  if  $t$  is chosen as

$$t = n + \left\lceil \log \left( 2 + \frac{1}{2\varepsilon} \right) \right\rceil$$

Furthermore, the algorithm uses  $O(t^2)$  elementary gates and  $t$  calls to controlled- $U^{2^j}$  for  $0 \leq j < t$

- ▶ **Be careful:** we need to compute  $\left(\mathbf{U}^{2^j}\right)_{0 \leq j \leq t}$  which has a cost  $\geq 2^t$  **unless one uses the particular shape of  $\mathbf{U}$**  . . .
- ▶ **Accuracy with a probability exponentially close to 1 at the cost of a “constant” overhead:**  $n$  bits of  $\varphi$  with probability  $1 - e^{-cn}$  but with  $t = O(n)$
- ▶ **Be careful:** to run phase estimation we also need to be able to compute the eigenvector  $|u\rangle$  . . .

## APPLICATION 1: QFT OVER $\mathbb{Z}/N\mathbb{Z}$

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Recall that characters of

- ▶  $\mathbb{F}_2^n$ :  $\chi_x(\mathbf{y}) = (-1)^{\mathbf{x} \cdot \mathbf{y}}$
- ▶  $\mathbb{Z}/2^n\mathbb{Z}$ :  $\chi_x(y) = e^{\frac{2i\pi xy}{2^n}}$
- ▶  $\mathbb{Z}/N\mathbb{Z}$ :  $\chi_x(y) = e^{\frac{2i\pi xy}{N}}$

Lecture 5: computing **efficiently** ( $O(n)$  and  $O(n^2)$ )

$$\text{QFT}_{\mathbb{F}_2^n} = \text{H}^{\otimes n} : |x\rangle \mapsto \frac{1}{\sqrt{2^n}} \sum_{y \in \{0,1\}^n} (-1)^{\mathbf{x} \cdot \mathbf{y}} |y\rangle \quad \text{and} \quad \text{QFT}_{\mathbb{Z}/2^n\mathbb{Z}} : |x\rangle \mapsto \frac{1}{\sqrt{2^n}} \sum_{y \in \mathbb{Z}/2^n\mathbb{Z}} e^{\frac{2i\pi xy}{2^n}} |y\rangle$$

**Aim: computing efficiently  $\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}$  (when  $N$  not a power of 2)**

$$\text{QFT}_{\mathbb{Z}/N\mathbb{Z}} : |x\rangle \mapsto \frac{1}{\sqrt{N}} \sum_{y \in \mathbb{Z}/N\mathbb{Z}} e^{\frac{2i\pi xy}{N}} |y\rangle$$

Computing  $\text{QFT}_{Z/NZ}$ : use **phase estimation!**

$$U_1(|k\rangle|0\rangle) \mapsto |k\rangle \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle \quad \text{and} \quad U_2(\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle|0\rangle) \mapsto \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle|k\rangle$$

→ These two unitaries are enough to compute  $\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle$ !

We can perform  $\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}$  as:

$$|k\rangle|0\rangle \xrightarrow{U_1} |k\rangle \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle \xrightarrow{\text{SWAP}} \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle|k\rangle \xrightarrow{U_2^{-1}} \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle|0\rangle$$

$$U_1(|k\rangle|0\rangle) \mapsto |k\rangle \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle \quad \text{and} \quad U_2(\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle|0\rangle) \mapsto \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}|k\rangle|k\rangle$$

Be careful:  $|k\rangle$  here is such that  $k \in \mathbb{Z}/N\mathbb{Z}$  and  $N$  may not be a power of two... In particular  $|k\rangle$  cannot be written as  $|0010\dots 1\rangle$

$(|k\rangle)_{k \in \mathbb{Z}/N\mathbb{Z}}$  is an orthonormal basis of an Hilbert space of dimension  $N$

→ This quantum space is called the space of **qudits**!

Two possibilities to perform computation with qudits: (i) encode qudits in qubits or (ii) implement your quantum device directly with Hilbert spaces of dimension  $> 2$

It is the same issue with classical computer! How to implement trits, namely  $\mathbb{Z}/3\mathbb{Z}$ ?

## COMPUTING THE FIRST UNITARY $U_1$

To build the unitary  $|k\rangle |0\rangle \mapsto |k\rangle \mathbf{QFT}_{\mathbb{Z}/N\mathbb{Z}} |k\rangle$  (admitting we can perform efficiently the different unitaries **over qudits**)

1. Start from  $|k\rangle |0\rangle |0\rangle$

2. Apply the “uniform superposition” over the second register

$$|k\rangle \frac{1}{\sqrt{N}} \sum_{j \in \mathbb{Z}/N\mathbb{Z}} |j\rangle |0\rangle$$

3. Apply the multiplication operator ( $|x\rangle |y\rangle |0\rangle \mapsto |x\rangle |y\rangle |xy \bmod N\rangle$ )

$$|k\rangle \frac{1}{\sqrt{N}} \sum_{j \in \mathbb{Z}/N\mathbb{Z}} |j\rangle |kj \bmod N\rangle$$

4. Apply the “phase flip in  $\mathbb{Z}/N\mathbb{Z}$ ” ( $|x\rangle \mapsto e^{2i\pi \frac{x^2}{N}} |x\rangle$ ) on the third register

$$|k\rangle \frac{1}{\sqrt{N}} \sum_{j \in \mathbb{Z}/N\mathbb{Z}} e^{2i\pi \frac{kj^2}{N}} |j\rangle |kj \bmod N\rangle$$

5. Apply the inverse of the multiplication operation:

$$|k\rangle \frac{1}{\sqrt{N}} \sum_{j \in \mathbb{Z}/N\mathbb{Z}} e^{2i\pi \frac{kj^2}{N}} |j\rangle |0\rangle = |k\rangle \mathbf{QFT}_{\mathbb{Z}/N\mathbb{Z}} |k\rangle |0\rangle$$

$$U : |k\rangle \mapsto |k + 1 \bmod N\rangle$$

$\rightarrow U^{2^j} : |k\rangle \mapsto |k + 2^j \bmod N\rangle$  can be built in time  $O(\log N)$

(  $x \mapsto x + 2^j \bmod N$  can be classically computed in time  $O(\log N)$  )

We have the following computation:

$$\begin{aligned} U(\text{QFT}_{\mathbb{Z}/N\mathbb{Z}} |k\rangle) &= \frac{1}{\sqrt{N}} \sum_{y \in \mathbb{Z}/N\mathbb{Z}} e^{\frac{2i\pi ky}{N}} U|y\rangle \\ &= \frac{1}{\sqrt{N}} \sum_{y \in \mathbb{Z}/N\mathbb{Z}} e^{\frac{2i\pi ky}{N}} |y+1\rangle \\ &= e^{\frac{-2i\pi k}{N}} \frac{1}{\sqrt{N}} \sum_{y \in \mathbb{Z}/N\mathbb{Z}} e^{\frac{2i\pi ky}{N}} |y\rangle \\ &= e^{2i\pi \frac{N-k}{N}} \text{QFT}_{\mathbb{Z}/N\mathbb{Z}} |k\rangle \end{aligned}$$

$\rightarrow \text{QFT}_{\mathbb{Z}/N\mathbb{Z}} |k\rangle$  is an eigenvector of  $U$  with eigenvalue  $e^{2i\pi\varphi}$  where  $\varphi \stackrel{\text{def}}{=} \frac{N-k}{N}$

(remember: Fourier basis is the basis where translation operator is diagonal)

## COMPUTING THE SECOND UNITARY $U_2$ : USE PHASE AMPLIFICATION

The translation operator  $U$  is diagonal in the Fourier basis

$\text{QFT}_{\mathbb{Z}/N\mathbb{Z}} |k\rangle$  is an eigenvector of  $U$  with eigenvalue  $e^{2i\pi\varphi}$  where

$$\varphi \stackrel{\text{def}}{=} \frac{N-k}{N}$$

Applying phase estimation with  $n = \lceil \log N \rceil$  (bits of precision) enables to compute:

$$\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}(|k\rangle) |0\rangle \mapsto \text{QFT}_{\mathbb{Z}/N\mathbb{Z}} |k\rangle |N-k\rangle$$

→ **Be careful:** phase estimation gives only an approximation of the transform!

Therefore: after applying the unitary  $|x\rangle \mapsto |N-x\rangle$  we obtain an approximation of

$$\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}(|k\rangle) |0\rangle \mapsto \text{QFT}_{\mathbb{Z}/N\mathbb{Z}}(|k\rangle) |k\rangle$$

**Cost:**

Given  $t = O(\log N)$ , we have a cost of  $O(t^2)$  plus the cost to run  $U^{2^j} : |k\rangle \mapsto |k + 2^j \bmod N\rangle$  for

$0 \leq j < t$  which can be done in time  $O(t^3)$  (clever combination of the  $U^{2^j}$ -controlled)

→ Final cost to compute  $\text{QFT}_{\mathbb{Z}/N\mathbb{Z}}$ :  $O(\log^3 N)$

## WHAT ABOUT THE GENERAL CASE?

Is it possible to efficiently build  $\text{QFT}_G$  where  $G$  is any arbitrary finite abelian group?



## WHAT ABOUT THE GENERAL CASE?

Is it possible to efficiently build  $\text{QFT}_G$  where  $G$  is any arbitrary finite abelian group?

→ Yes!

How to proceed (rough explanation):

Any finite abelian group  $G$  of size  $N$  is isomorphic to the product of cyclic groups:

$$\mathbb{Z}/n_1\mathbb{Z} \times \cdots \times \mathbb{Z}/n_k\mathbb{Z}$$

Then (admitted),

$\text{QFT}_G$  can be written as  $\text{QFT}_{\mathbb{Z}/n_1\mathbb{Z}} \otimes \cdots \otimes \text{QFT}_{\mathbb{Z}/n_k\mathbb{Z}}$

→ We deduce that  $\text{QFT}_G$  can be computed in time  $O(\log^3 \#G)$

Be careful, given a finite Abelian group it is **classically** hard to compute its decomposition as cyclic groups. . . Quantum case: end of the lecture

## APPLICATION 2: ORDER FINDING

---

## Order finding problem:

- **Input:** integers  $x, N$  where  $\gcd(x, N) = 1$
- **Output:** least positive integer  $r$  such that  $x^r = 1 \pmod N$

Solving the factorization reduces to this problem  
(solving order finding  $\implies$  solving factorization)

## Proposition:

We can quantumly determine the order  $r$  (with high probability) in time  
 $O(\log^3 N)$

→ Best classical algorithms are sub-exponential in  $N$ :

$$\exp\left((c + o(1)) \log^\alpha(N) \log^{1-\alpha}(\log N)\right)$$

where  $c, \alpha$  are constants

Suppose that we work in the space of  $L$  qubits

Given  $y \in \llbracket 0, 2^L - 1 \rrbracket$ , we will naturally identify  $|y\rangle$  to  $|y_1 \dots y_L\rangle$

where  $y_1 \dots y_L$  **binary decomposition of  $y$**

**For instance:**

Given  $3, 5 \in \llbracket 0, 2^3 - 1 \rrbracket$ ,

$$|3\rangle = |011\rangle \quad \text{and} \quad |5\rangle = |101\rangle$$

## IT REDUCES TO PHASE ESTIMATION

$x$  integer :  $\gcd(x, N) = 1$  and  $r$  its order, smallest positive integer such that  $x^r = 1 \pmod N$

Phase estimation applied to the following unitary and eigenvector:

Let,

$L \stackrel{\text{def}}{=} \lceil \log N \rceil$  (work in the space of  $L$ -qubits)

$$\forall y \in \{0, 1\}^L = \llbracket 0, 2^L - 1 \rrbracket, \quad \mathbf{U} |y\rangle \stackrel{\text{def}}{=} \begin{cases} |xy \pmod N\rangle & \text{if } 0 \leq y \leq N - 1 \\ |y\rangle & \text{otherwise } (N - 1 < y < 2^{\lceil \log N \rceil}) \end{cases}$$

$$\forall s \in \llbracket 0, r \rrbracket, \quad |u_s\rangle \stackrel{\text{def}}{=} \frac{1}{\sqrt{r}} \sum_{k=0}^{r-1} e^{-\frac{2i\pi sk}{r}} |x^k \pmod N\rangle \text{ eigenvector of } \mathbf{U} \text{ with eigenvalue } e^{2i\pi \frac{s}{r}}$$

→ We work here in the **space of qubits** (natural trick, identity if integers  $\geq N - 1$ )

$$\begin{aligned} \mathbf{U} |u_s\rangle &= \frac{1}{\sqrt{r}} \sum_{k=0}^{r-1} e^{-\frac{2i\pi sk}{r}} \mathbf{U} |x^k \pmod N\rangle \\ &= \frac{1}{\sqrt{r}} \sum_{k=0}^{r-1} e^{-\frac{2i\pi sk}{r}} |x^{k+1} \pmod N\rangle \\ &= e^{2i\pi \frac{s}{r}} |u_s\rangle \end{aligned}$$

→ Be careful: in the last equality we used:  $x$  has order  $r$  modulo  $N$ , thus  $x^r = 1 \pmod N$

## BUT TWO QUESTIONS

For the eigenvalue  $\frac{s}{r}$ : we work in  $\mathbb{Z}/N\mathbb{Z}$  and with  $L = \lceil \log N \rceil$  qubits

To perform efficiently phase estimation, two issues:

- ▶ How to compute efficiently the  $U^{2^j}$ 's?
- ▶ How to compute the eigenvector  $|u_s\rangle$ ?

→ We will be able to recover approximations of  $\frac{s}{r}$ , **not**  $r$  . . .

For the eigenvalue  $\frac{\xi}{r}$ : we work in  $\mathbb{Z}/N\mathbb{Z}$  and with  $L = \lceil \log N \rceil$  qubits

To perform efficiently phase estimation, two issues:

- ▶ How to compute efficiently the  $U^{2^j}$ 's?
- ▶ How to compute the eigenvector  $|u_s\rangle$ ?

→ We will be able to recover approximations of  $\frac{\xi}{r}$ , **not**  $r$  . . .

Be patient!

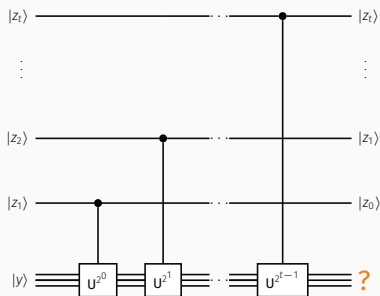
### Parameter of phase estimation:

We will determine the first  $2L + 1$  bits of  $\frac{\xi}{r}$  with probability  $1 - \varepsilon$

→ Choose in phase estimation  $t = 2L + 1 + \lceil \log(2 + \frac{1}{2\varepsilon}) \rceil$  qubits

In particular:  $t = O(L)$  even if  $\varepsilon = e^{-cL}$  with  $C > 0$  (constant)

$$U|y\rangle = |xy \bmod N\rangle \quad (0 \leq y \leq N-1)$$



The above circuit (used in the phase estimate) performs the following computation:

$$\begin{aligned} |z_t \dots z_1\rangle |y\rangle &\longrightarrow |z\rangle U^{z_t 2^{t-1}} \dots U^{z_1 2^0} |y\rangle \\ &= |z\rangle \left| x^{z_t 2^{t-1}} \times \dots \times x^{z_1 2^0} y \bmod N \right\rangle \\ &= |z\rangle |yx^z \bmod N\rangle \end{aligned}$$

→ To perform **efficiently** phase estimation: compute  $|z\rangle |y\rangle \mapsto |z\rangle |yx^z \bmod N\rangle$  efficiently

(**modular exponentiation**)



Aim: computing efficiently

$$|z\rangle |y\rangle \mapsto |z\rangle |yx^z \bmod N\rangle$$

1. Let  $U_{EM} : |z\rangle |y\rangle \mapsto |z\rangle |y \oplus (x^z \bmod N)\rangle$  (be careful  $z \mapsto x^z \bmod N$  not bijective)

$$|z\rangle |y\rangle |0\rangle \xrightarrow{U_{EM}} |z\rangle |y\rangle |x^z \bmod N\rangle \xrightarrow{\text{mult}} |z\rangle |yx^z \bmod N\rangle |x^z \bmod N\rangle \xrightarrow{U_{EM}^{-1}} |z\rangle |yx^z \bmod N\rangle |0\rangle$$

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2. Computing efficiently  $U_{EM}$ : classically

$$x \mapsto x^z \bmod N$$

can be computed in  $O(\log z) = O(\log t) = O(\log N)$  squaring, therefore  $O(\log^3 N)$  operations

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Conclusion: using phase estimation

We determine the first  $2L + 1$  bits of  $\frac{s}{r}$  with probability  $1 - e^{-cL}$  in time  $O(L^3)$  where  $L = \lceil \log N \rceil$

Aim: computing

$$|u_s\rangle = \frac{1}{\sqrt{r}} \sum_{k=0}^{r-1} e^{-\frac{2i\pi sk}{r}} |x^k \bmod N\rangle$$

But we do not know  $r$  . . . Our aim is to find it!

The trick:

$$\frac{1}{\sqrt{r}} \sum_{s=0}^{r-1} |u_s\rangle = |1\rangle$$

→ Plugging  $|1\rangle$  in the phase estimation algorithm will give the first  $2L + 1$  bits of  $\frac{s}{r}$  for some  
 (uniform and **unknown**)  $s \in \llbracket 0, r - 1 \rrbracket$  with probability  $1 - \epsilon$

Exercise Session:

Proof of this statement

Up to now we have recovered (with high probability) in quantum time  $O(L^3)$  the first  $2L + 1$  bits of

$$\frac{s}{r} \text{ where } 0 \leq s < r \text{ and } s \in \llbracket 1, N - 1 \rrbracket$$

→ It does not give  $r$ , even  $\frac{s}{r} \dots$

Theorem (admitted) about continued fractions:

Let  $\tilde{\varphi}$  be a rational given as input, let  $s$  and  $r$  be  $L$  bits integers such that

$$\left| \frac{s}{r} - \tilde{\varphi} \right| < \frac{1}{2^{2L}}$$

Then, there exists an algorithm (using “continued fractions”) that outputs  $(s', r')$  which verifies

$$\gcd(s', r') = 1 \quad \text{and} \quad \frac{s'}{r'} = \frac{s}{r}$$

using  $O(L^3)$  classical operations

In our case:

With probability  $1 - \epsilon$ : phase estimation outputs  $\tilde{\varphi}$  an approximation of  $\frac{s}{r}$  accurate to  $2L + 1$  bits, therefore:

$$\left| \frac{s}{r} - \tilde{\varphi} \right| \leq \frac{1}{2^{2L+1}} \leq \frac{1}{2^{2L}} \quad \left( \text{as } r \leq N - 1 \leq L = \lceil \log N \rceil \right)$$

→ In time  $O(L^3)$  we compute  $s', r'$  co-prime such that  $\frac{s'}{r'} = \frac{s}{r}$

$s', r'$  are co-prime such that  $\frac{s'}{r'} = \frac{s}{r}$

→ If  $\text{gcd}(s, r) > 1$  then  $r' \neq r$ , only  $r' \mid r \dots$

A solution (but inefficient...):

The number of prime numbers  $< r$  is  $\approx r / \log(r)$

→  $\mathbb{P}(\text{gcd}(s, r) = 1) \approx \log(r)/r$  as  $s$  is uniformly picked in  $\llbracket 0, r - 1 \rrbracket$

Therefore we need to repeat  $\approx r = O(L)$  number of times the algorithm before reaching  $\text{gcd}(s, r) = 1$ . It will increase the cost from  $O(L^3)$  to  $O(L^4)$ .

Fundamental remark:

$$\frac{s'_1}{r'_1} = \frac{s_1}{r} \implies r s'_1 = s_1 r'_1 \quad \text{and} \quad \frac{s'_2}{r'_2} = \frac{s_2}{r} \implies r s'_2 = s_2 r'_2$$

We have  $\gcd(s'_1, r'_1) = \gcd(s'_2, r'_2) = 1$ , supposing that  $\gcd(s'_1, s'_2) = 1$  implies that  $r = \text{lcm}(r'_1, r'_2)$

→ Therefore: obtaining two estimations  $(s'_1, r'_1)$  and  $(s'_2, r'_2)$  and supposing that  $\gcd(s'_1, s'_2) = 1$   
we can recover  $r = \text{lcm}(r'_1, r'_2)$ .

What is the probability that  $\gcd(s'_1, s'_2) = 1$  given that  $s'_1$  and that  $s'_2$  are uniformly distributed in  $\llbracket 0, r - 1 \rrbracket$ ?



## REPEAT JUST A CONSTANT NUMBER OF TIMES!

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→ It is  $\geq \frac{1}{4}$  (see Exercise Session)

**In conclusion:**

Repeating the algorithm a constant number of times enables to recover  $r$  with probability exponentially close to one (times  $(1 - \varepsilon)$ )!

## ORDER FINDING ALGORITHM

To compute the order  $r$  of  $x \bmod N$  (where  $\gcd(x, N) = 1$ ) we first run a constant number of times the **phase estimation** with  $t = 2\lceil \log N \rceil + 1 + \log\left(2 + \frac{1}{2\varepsilon}\right)$ . It has been necessary to compute:

- ▶  $\text{QFT}_{\mathbb{Z}/2^t\mathbb{Z}}$ : done in time  $O(t^2) = O(\log^2 N)$
- ▶ modular exponentiation  $|z\rangle |y\rangle \mapsto |z\rangle |yx^z \bmod N\rangle$ : done in time  $O(\log^3 N)$

It outputs a  $2\lceil \log N \rceil + 1$  approximation of some  $\frac{s}{r}$  where  $s \in \llbracket 0, r - 1 \rrbracket$  is uniform and unknown

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Then after collecting some approximations of  $\frac{s_i}{r}$ , apply **continued fraction algorithm** to obtain  $(s'_i, r'_i)$  in time  $O(\log^3 N)$  with  $\frac{s'_i}{r'_i} = \frac{s_i}{r}$ . It enables to get  $r$  by computing some  $\text{lcm}(r'_i, r'_j)$ .

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→ This procedure works with probability  $(1 - e^{-C \log N})(1 - \varepsilon)$  for constant  $C > 0$  depending on the number of repetitions

**Final cost:**

$$O(\log^3 N)$$

→ This could be done in time  $O(\log^2(N) \text{poly}(\log \log N))$

Order finding algorithm is efficient because we know quantumly how to perform classical computations **and the quantum Fourier transform over  $\mathbb{Z}/2^t\mathbb{Z}$**

# SHOR'S ALGORITHM

---

**Factoring problem:**

- **Input:** an integer  $N$
- **Output:** a non-trivial factor of  $N$

→ Security of public-key encryption scheme RSA relies on the hardness of this problem. . .

Classically best algorithms have a complexity:

$$\exp\left((c + o(1)) \log^\alpha(N) \log^{1-\alpha}(\log N)\right)$$

*Shor's algorithm is basically applying order finding for some random  $x \in \llbracket 0, N - 1 \rrbracket \dots$*

*But why?*



## Theorem 1:

Suppose  $N$  is a  $L$  bits not prime integer and  $1 \leq y \leq N$  be a non-trivial integer such that

$$y^2 = 1 \pmod N$$

Then, at least  $\gcd(y - 1, N)$  or  $\gcd(y + 1, N)$  is a non-trivial factor of  $N$  that can be computed in time  $O(L^3)$

## Theorem 2:

Suppose that  $N = p_1^{\alpha_1} \cdots p_m^{\alpha_m}$  where the  $p_i$ 's are different primes. Let  $x$  be an integer chosen uniformly at random, subject to the requirements that  $1 \leq x \leq N - 1$  and  $\gcd(x, N) = 1$ . Let  $r$  be the order of  $x$ . Then,

$$\mathbb{P} \left( r \text{ is even and } x^{r/2} \neq -1 \pmod N \right) \geq 1 - \frac{1}{2^m}$$

→ Let  $x$  be picked according to Theorem 2, then with (at least) a constant probability

$$x^{r/2} \text{ is a solution } \neq \pm 1 \text{ of } (X^2 = 1 \pmod N)$$

According to Theorem 1:  $\gcd(x^{r/2} - 1, N)$  or  $\gcd(x^{r/2} + 1, N)$  is a  $\neq \pm 1$  factor of  $N$

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According to Theorem 1:  $\gcd(x^{r/2} - 1, N)$  or  $\gcd(x^{r/2} + 1, N)$  is a  $\neq \pm 1$  factor of  $N$

Given  $x$ , we just need to compute its order  $r$  to find a non-trivial factor!

1. Pick  $x$  uniformly at random in  $\llbracket 1, N \rrbracket$
2. Compute  $d = \text{gcd}(x, N)$ . If  $d > 1$ , output  $d$
3. Use the **quantum** order-finding subroutine to find the order  $r$  of  $x \bmod N$
4. If  $r$  is even and  $x^{r/2} \not\equiv -1 \pmod N$  then compute  $\text{gcd}(x^{r/2} - 1, N)$  or  $\text{gcd}(x^{r/2} + 1, N)$  and test if one of these is a non-trivial factor of  $N$ . Otherwise go back to Step 1.

By using the law of total probability:

$$\begin{aligned}
 \mathbb{P}(\text{success}) &\geq \mathbb{P}(\text{success} \mid \text{Step 3 succeeds}) \cdot \mathbb{P}(\text{Step 3 succeeds}) \\
 &= \mathbb{P}(r \text{ is even and } x^{r/2} \not\equiv -1 \pmod N) \cdot \mathbb{P}(\text{order finding succeeds}) \\
 &\geq \left(1 - \frac{1}{2^m}\right) \cdot (1 - e^{-c \log N})(1 - \epsilon) \quad (m \text{ number of prime factors of } N)
 \end{aligned}$$

→ Repeating the algorithm a constant amount of times gives a non-trivial factor

**Final cost:**

$$O(\log^3 N) \text{ cost of phase estimation} + \text{Step 4}$$

# HIDDEN SUBGROUP PROBLEM

---

Shor's algorithm relies on the order-finding which itself crucially used  $\text{QFT}_{\mathbb{Z}/2^t\mathbb{Z}}$   
(in the phase estimation)

→ It turns out that what we did is extremely “general”

Techniques we have presented enable to compute the “period” of a wide class of functions. . .

- ▶ What do we mean by “general”?
- ▶ Computing the “period” of which class of functions and does it imply some interesting statements?

→ Hidden Subgroup Problem!

## Hidden Subgroup Problem (HSP):

- **Input:** a function  $f : G \rightarrow S$  where  $G$  is a known group<sup>a</sup> and  $S$  is a finite set
- **Promise:**  $f$  satisfies

$$f(x) = f(y) \text{ if and only if } y \in xH$$

$$\text{i.e., } y = xh \text{ for some } h \in H$$

for an **unknown subgroup**  $H \subseteq G$

- **Output:**  $H$

---

<sup>a</sup> see later for a precise definition

→ We say that  $f$  hides the subgroup  $H$

## Left-cosets:

The set:

$$xH \stackrel{\text{def}}{=} \{xh : h \in H\}$$

is called a **left-coset** of  $H$

→ A function  $f$  that hides  $H$  is constant on each left-coset of  $H$  and distinct on different left cosets

## WHY IS THIS PROBLEM IMPORTANT?

HSP may be seen as a purely abstract problem. . . **But no!**

Here are **particular instantiations of HSP**

- ▶ Simon's problem:

$$G = \mathbb{F}_2^n, H = \{0, s\} \text{ and } f \text{ being the input in Simon's problem}$$

- ▶ Order finding:

$$G = \mathbb{Z}/\Phi(N)\mathbb{Z} \left( \Phi \text{ be the Euler function} \right), H = \{rx : x \in \mathbb{Z}/\Phi(N)\mathbb{Z}\} \text{ and } f(a) = x^a \bmod N$$

- ▶ Discrete logarithm problem: see Exercise Session!
- ▶ etc. . .

**Be careful:**

In Shor's algorithm, when using a solver for the order finding problem we don't know  $\Phi(N)$  and therefore we don't know  $G$  . . .

We suppose that  $G$  is **Abelian** (we note  $G$  in additive notation)

$f : G \rightarrow S$  that hides some subgroup  $H$

1. Start with  $|0\rangle |0\rangle$ , where the two registers have dimensions  $\#G$  and  $\#S$ , respectively
2. Create a uniform superposition over  $G$  in the first register:  $\frac{1}{\sqrt{\#G}} \sum_{g \in G} |g\rangle |0\rangle$
3. Compute  $f$  in superposition:  $\frac{1}{\sqrt{\#G}} \sum_{g \in G} |g\rangle |f(g)\rangle$
4. Measure the second register. This yields some value  $s \in S$ . The second register collapses to (using the promise over  $f$ )

$$\frac{1}{\sqrt{\#H}} \sum_{h \in H} |s + h\rangle$$

5. **Apply QFT<sub>G</sub>** giving:  $\frac{1}{\sqrt{\#H}} \sum_{h \in H} |\chi_{s+h}\rangle$  for some quantum state  $|\chi_{s+h}\rangle$  ( $\chi_g$  characters of  $G$ )
6. Measure and output the resulting  $g \in G$



## WHY DOES IT WORK?

$G$  is Abelian, be  $(\chi_g)_{g \in G}$  be its characters

$$\begin{aligned} |\chi_{s+h}\rangle &= \text{QFT}_G \sum_{h \in H} |s+h\rangle \\ &= \frac{1}{\sqrt{G}} \sum_{h \in H} \text{QFT}_G |s+h\rangle \\ &= \frac{1}{\sqrt{G}} \sum_{h \in H} \sum_{g \in G} \chi_g(s+h) |g\rangle \\ &= \frac{1}{\sqrt{G}} \sum_{g \in G} \left( \sum_{h \in H} \chi_g(h) \right) \chi_g(s) |g\rangle \quad \text{from Lecture 5: } \sum_{h \in H} \chi_g(h) = \begin{cases} \#H & \text{if } g \in H^\perp \\ 0 & \text{otherwise} \end{cases} \\ &= \frac{1}{\sqrt{G}} \sum_{g \in H^\perp} \#H \chi_g(s) |g\rangle \end{aligned}$$

→ The quantum step before measurement is:  $\sqrt{\frac{\#H}{\#G}} \sum_{g \in H^\perp} \chi_g(s) |g\rangle$

The quantum state before measurement is:

$$\frac{1}{\sqrt{\#H}} \sum_{g \in H^\perp} \chi_g(s) |g\rangle \quad \text{where } H^\perp = \{g \in G : \forall h \in H, \chi_g(h) = 1\}$$

→ Measuring gives a uniform  $g \in H^\perp$  giving some information about  $H \dots$

repeating a **poly(log #G)** times enables to recover  $H$  with high probability!

- For a rigorous proof of this statement: see Chapter 6 in the lecture notes by Andrew Childs

**An example: Simon's problem**

$$G = \mathbb{F}_2^n, \quad \forall \mathbf{x}, \mathbf{y} \in \mathbb{F}_2^n, \quad \chi_{\mathbf{x}}(\mathbf{y}) = (-1)^{\mathbf{x} \cdot \mathbf{y}} \quad \text{and } H = \{\mathbf{0}, \mathbf{s}\}$$

$$\rightarrow H^\perp = \{\mathbf{x} \in \mathbb{F}_2^n : \mathbf{x} \cdot \mathbf{s} = 0\}$$

In other words, we recover Simon's algorithm. . .

## HOW TO COMPUTE QFT OVER $G$ ?

$G$  is an **Abelian** group

Recall that we compute  $\text{QFT}_G$  as  $\text{QFT}_{\mathbb{Z}/n_1\mathbb{Z}} \otimes \cdots \otimes \text{QFT}_{\mathbb{Z}/n_k\mathbb{Z}}$  where **we used the isomorphism:**

$$G \cong \mathbb{Z}/n_1\mathbb{Z} \times \cdots \times \mathbb{Z}/n_k\mathbb{Z} \quad (1)$$

But is it easy to compute this isomorphism/decomposition even if we “know”  $G$ ?

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But is it easy to compute this isomorphism/decomposition even if we “know”  $G$ ?

→ **Yes!** At least quantumly for a “good” definition of knowing  $G$  . . .

**Quantum decomposition of Abelian groups:**

Suppose we have (i) **a unique encoding of each element of  $G$** , (ii) **the ability to perform group efficiently operations on these elements**, and (iii) **a generating set for  $G$** .

Then, there exists an efficient quantum algorithm that decomposes  $G$ , namely outputs the isomorphism given in Equation (1)

→ See Chapter 6 in the lecture notes by Andrew Childs

## GENERALIZATION TO THE NON-ABELIAN CASE?

To solve HSP we crucially used that we restrict ourself to the Abelian case . . .

(in the Abelian case,  $H^\perp$  gives linear relations enabling to recover  $H$ )

→ And the **non-Abelian case**?

No efficient algorithm is known for the non-Abelian case

(even if nothing indicates that it is impossible) . . .

→ Finding such an algorithm would have a huge impact in theoretical computer science,

(post-quantum) cryptography. . .

If you are interested by this topic:

- ▶ Nice reading about Fourier transform (classical & quantum) over non-Abelian group: Chapter 11 in the lectures by Andrew Child <https://www.cs.umd.edu/~amchilds/qa/>
- ▶ The hidden nonabelian subgroup problem and the Kuperberg algorithm, see Chapters 11-13 in the lectures by Andrew Child <https://www.cs.umd.edu/~amchilds/qa/>

# EXERCISE SESSION

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