



# INTRODUCTION TO QUANTUM ALGORITHMS

## Short course report

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## Index

<b>1. Computational model</b>	<b>3</b>
1.1. Tensor product	3
1.2. Primitive constructs	3
1.3. Quantum states	3
1.4. Operations on qubits	3
1.4.1. Basic operations	4
<b>2. Amplitude amplification</b>	<b>4</b>
2.1. Grover's black-box search algorithm	4
2.1.1. Complexity	4
<b>3. Quantum Fourier transform</b>	<b>4</b>
3.1.1. Complexity	5
<b>4. Gradient estimation</b>	<b>5</b>
4.1. Jordan's quantum gradient algorithm	5
4.2. Nonlinear functions	5
<b>5. Linear systems solving</b>	<b>5</b>
<b>6. Shor's factorization algorithm</b>	<b>6</b>
6.1. Classical integer factorization	6
6.2. Quantum factorization	6
6.3. Implications	6

# 1. Computational model

In order to understand quantum algorithms it is necessary to define the tensor product operation between matrices and to present some primitive constructs.

## 1.1. Tensor product

Let  $A \in \mathbb{C}^{m \times n}$ ,  $B \in \mathbb{C}^{p \times q}$ ,  $D \in \mathbb{C}^{mp \times nq}$

The tensor product  $\otimes$  is defined as a bilinear operator between vector spaces:

$$\otimes : A \times B \rightarrow D$$

In the course the  $\otimes$  operator always refers to the Kronecker product, a special case of the more generic tensor product.

### Kronecker product

$$D := A \otimes B = \begin{pmatrix} a_{11}B & \dots & a_{1n}B \\ a_{21}B & \dots & a_{2n}B \\ \vdots & & \vdots \\ a_{m1}B & \dots & a_{mn}B \end{pmatrix}$$

## 1.2. Primitive constructs

**Bra-kets** Ket  $|\psi\rangle$  denotes a column vector in  $\mathbb{C}^n$ , while bra  $\langle\psi|$  denotes a row vector in  $(\mathbb{C}^n)^\dagger$ , such that  $\langle\psi| = |\psi\rangle^\dagger$

Properties:

- $|\psi\rangle\langle\phi| = |\psi\rangle \otimes \langle\phi|$
- $\langle\psi|\phi\rangle = \langle\psi, \phi\rangle \Rightarrow$  inner product

**Binary strings**  $\vec{j} \in \{0, 1\}^q$  denotes a binary string of  $q$  digits, such that  $j = \sum_{k=1}^q j_k \cdot 2^{q-k}$

$|\vec{j}\rangle = |j_1 j_2 \dots j_q\rangle = |j_1\rangle |j_2\rangle \dots |j_q\rangle =$  basis vector of  $(\mathbb{C}^2)^{\otimes q}$  with a 1 in position  $j$

## 1.3. Quantum states

A state in a quantum computer coincides with a single (possibly composite) register of qubits.

A register of  $q$  qubits can be represented with a vector  $|\psi\rangle$  in  $(\mathbb{C}^2)^{\otimes q}$  (which is  $\subseteq \mathbb{C}^{2^q}$ ).

A state  $|\psi\rangle$  can be represented as  $\sum_{\vec{j} \in \{0,1\}^q} \alpha_j |\vec{j}\rangle$ ,  $\alpha_i \in \mathbb{C}$ , which is equivalent to  $\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_{2^q} \end{pmatrix}$

**Basis states / superpositions** A  $q$ -qubits register  $|\psi\rangle$  represents a basis state if  $|\psi\rangle = |\vec{j}\rangle$ , for some  $\vec{j} \in \{0, 1\}^q$ . Otherwise, it is a superposition.

**Product states / entanglements** A  $q$ -qubits quantum state  $|\psi\rangle$  is a product state if  $|\psi\rangle = |\psi_1\rangle |\psi_2\rangle \dots |\psi_q\rangle$ , where  $q_i$  is a 1-qubit state. Otherwise, it is an entangled state.

## 1.4. Operations on qubits

**Operations** An operation, or gate, is a unitary matrix  $U \in \mathbb{C}^{2^q \times 2^q}$

$\Rightarrow$  Operations on quantum states are always:

- Linear
- Reversible

**Measurement** Gate allowing to gather information on the state. Applying a measurement to a  $q$ -qubits state  $|\psi\rangle = \sum_{\vec{j} \in \{0,1\}^q} \alpha_j |\vec{j}\rangle$  returns the binary string  $\vec{k} \in \{0,1\}^q$  with probability  $|\alpha_k|^2$

After applying a measurement gate on a state, the original state is not recoverable.

### 1.4.1. Basic operations

There are some basic gates that are used to create more complex operations:

- $X$  gate: performs a bit flip  $\Rightarrow X|0\rangle = |1\rangle, X|1\rangle = |0\rangle$

$$X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

- $H$  (Hadamard) gate: returns uniform superposition of basis states (when run in parallel on all qubits):

$$H^{\otimes q}|0\rangle = \frac{1}{\sqrt{2^q}} \sum_{\vec{j} \in \{0,1\}^q} |\vec{j}\rangle$$

- $CX$  (Controlled  $X$ /NOT) gate: two-qubit gate that can create or destroy entanglement.

$CX_{12}$  flips target qubit (qubit 2), when control qubit (qubit 1) is  $|1\rangle$ , leaves qubit 2 as it is otherwise:  $CX_{21}|1\rangle = |01\rangle, CX_{21}|11\rangle = |10\rangle$

$CX_{21}$  is the same, but with target qubit 1 and control qubit 2:  $CX_{21}|10\rangle = |10\rangle, CX_{21}|11\rangle = |01\rangle$

$$CX_{12} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix} \quad CX_{21} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

## 2. Amplitude amplification

Amplitude amplification is a technique used to distinguish a good state  $|\psi_G\rangle$  from a bad state  $|\psi_B\rangle$  by flipping the sign of  $|\psi_G\rangle$ , in order to find a state that has a large overlap with  $|\psi_G\rangle$ .

### 2.1. Grover's black-box search algorithm

A special case of amplitude amplification is Grover's black-box search algorithm.

Let  $f : \{0,1\}^n \rightarrow \{0,1\}$  be an unknown function, that takes as input an  $n$ -qubits binary string and outputs 1 for a unique string  $\vec{l}$ , 0 otherwise.

$$f(\vec{x}) = \begin{cases} 1 & \text{if } \vec{x} = \vec{l} \\ 0 & \text{otherwise} \end{cases}$$

The good state  $|\psi_G\rangle$  searched by the algorithm is the one in which  $f(\vec{x})$  returns 1. It does it by repeatedly brute forcing every possible string and amplifying the good state at every step, until the probability of obtaining  $\vec{l}$  from a measurement is almost 1.

#### 2.1.1. Complexity

The optimal number of iterations  $k$  is around  $\frac{\pi}{4}\sqrt{2^n} \Rightarrow O(2^{\frac{n}{2}})$ . Compared to a classic algorithm with complexity  $O(2^n)$ , the quantum version provides a quadratic speedup.

## 3. Quantum Fourier transform

The quantum Fourier transform algorithm (QFT) aims to compute the DFT of a vector  $x \in \mathbb{C}^{2^n}$ .

The classic DFT is defined as

$$y_j = \sum_{k=0}^{2^n-1} x_k \cdot e^{\frac{2\pi i j k}{2^n}} \quad \forall j = 0, 1, \dots, 2^n - 1$$

The matrix  $Q_n$  that implements the  $n$ -qubit QFT is

$$(Q_n)_{jk} = \frac{1}{\sqrt{2^n}} \omega_n^{jk} \quad \forall \vec{j}, \vec{k} \in \{0, 1\}^n \quad \omega_n = e^{\frac{2\pi i}{2^n}}$$

### 3.1.1. Complexity

It is not easy to declare a precise complexity for the algorithm that implements  $Q_n$ , but it uses a polynomial amount of resources. By using the *elementary gate complexity* measure the complexity can be declared as polynomial, which-in any case at all-is a huge improvement over the classical DFT algorithm (fast Fourier transform), that has a complexity of  $O(n \cdot e^n)$ .

This leads to an exponential speedup of the QFT over the classic algorithm.

## 4. Gradient estimation

Classical computers are able to estimate the gradient of a  $d$ -dimension function  $f : [0, 1]^d \rightarrow \mathbb{R}_+$  at the origin by evaluating  $f$  in  $O(d)$  time.

Quantum computers are able to do it by evaluating  $f$  once, a constant amount of time.

### 4.1. Jordan's quantum gradient algorithm

Given a linear  $d$ -dimension function  $f(x) = a^\top x + b$ ,  $a \in \mathbb{R}^d$  the algorithm returns  $a = \nabla f(0)$ .

A binary oracle  $U_f$  of the function is needed: a matrix such that

$$U_f |\vec{x}\rangle |\vec{y}\rangle = |\vec{x}\rangle |\vec{y} \boxplus f(\vec{x})\rangle$$

Where  $|\vec{x}\rangle$  is a register composed of  $d$   $q$ -qubit registers, representing the arguments  $x_1, x_2, \dots, x_d$  of  $f \Rightarrow |\vec{x}\rangle = |\vec{x}_1\rangle |\vec{x}_2\rangle \dots |\vec{x}_d\rangle$

In particular, when  $|\vec{y}\rangle$  is the initial quantum state

$$U_f |\vec{x}\rangle |\vec{0}\rangle = |\vec{x}\rangle |f(\vec{x})\rangle$$

The algorithm returns  $a$ , which is  $\nabla f(0)$ , the gradient of  $f$  at the origin, by applying the  $H$  gate on  $|\vec{x}\rangle$ , QFT on  $|\vec{y}\rangle$ , then  $U_f$  on the result. In order to measure  $a$ ,  $\text{QFT}^{-1}$  is applied.

### 4.2. Nonlinear functions

If  $f$  is nonlinear, an approximation can be constructed as  $g(x) = \nabla f(0)^\top x$ . Then Jordan's algorithm can be applied to  $g$ , in order to estimate  $\nabla f(0)$ .

$g(x)$  can be expressed (for  $m$  large enough) as

$$\sum_{k=-m}^m a_n f(kx)$$

The complexity is  $O(m)$ .

## 5. Linear systems solving

Linear systems can be solved by a quantum algorithm.

Given a linear system  $Ax = b$ ,  $A \in \mathbb{C}^{2^n \times 2^n}$ ,  $b \in \mathbb{C}^{2^n}$

The solution  $x$  of the system is encoded in a quantum state  $|\psi\rangle$  such that

$$\| |\psi\rangle - |\text{amp}(x)\rangle \| \leq \varepsilon$$

For some precision parameter  $\varepsilon > 0$ .

$|\text{amp}(x)\rangle$  is the amplitude encoding of the solution as a quantum state, equivalent to

$$|\text{amp}(x)\rangle := \sum_{\vec{j} \in \{0,1\}^n} \frac{x_j}{\|x\|} |\vec{j}\rangle = \begin{pmatrix} x_1/\|x\| \\ x_2/\|x\| \\ \vdots \\ x_{2^n-1}/\|x\| \end{pmatrix}$$

By encoding the result in a quantum state it cannot be immediately measured. The algorithm is supposed to be used as a stepping stone for other operations, as there is no performance improvement over classical methods if the aim is to measure the solution.

## 6. Shor's factorization algorithm

An interesting quantum algorithm, that brings great performance improvements and has real-world implications is Shor's integer factorization algorithm.

### 6.1. Classical integer factorization

Factorizing an integer  $n$  is generally intractable on classical computational models, since the fastest known algorithm has a sub-exponential complexity of  $O\left(e^{(\log n)^{\frac{1}{3}}(\log \log n)^{\frac{2}{3}}}\right)$ .

Many cryptography-related algorithms and protocols exploit this fact to ensure (algorithmically)-unbreakable encryption.

### 6.2. Quantum factorization

Shor's algorithm provides an exponential speedup over classical approaches: it uses a number of quantum gates of order  $O((\log n)^2(\log \log n)(\log \log \log n))$ .

The high-level steps of Shor's algorithm are:

1. The problem of finding  $n$ 's integer factors is reduced to finding the period of a function  $f(x) = a^x \bmod n$ . The objective is finding the smallest positive integer  $r$ , such that  $a^r \equiv 1 \bmod n$ ,  $r$  being the period of  $f$
2. Quantum phase estimation process: applies modular exponentiation and the inverse QFT to a uniform superposition of all possible states (initialized with a H gate)
3. Post processing process: uses the period  $r$  to find the factorization of  $n$

### 6.3. Implications

Implementing this quantum algorithm would lead to severe consequences in a big section of current cryptography and cyber security in general. It could be used to break various cryptography mechanisms, like private/public-key schemes, most notably:

- The RSA algorithm
- Diffie-Hellman key exchange

However, current quantum computers seem to lack a sufficient number of qubits and results are not so stable due to noise and errors, for Shor's algorithm to pose a serious threat in real-world scenarios.