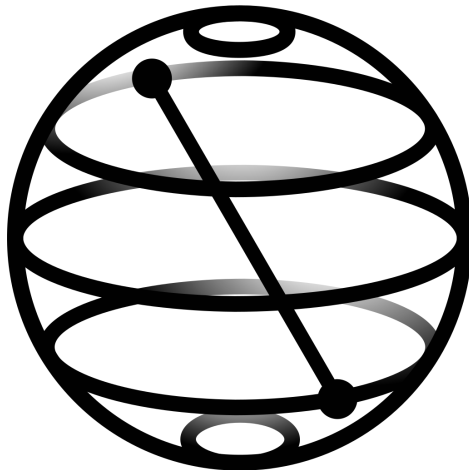


# Introduction to Quantum Programming

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# 1 Introduction

## 1.1 Libraries

Quantum programming can be done using several programming libraries, we are going to use **Qiskit** which is a Python library for quantum computing. With Qiskit we can run our code on IBM real quantum computers. It can be installed with the command: `pip install qiskit`. Note that you may want to install matplotlib to visualize your quantum circuits. You may also need to run the command: `pip install pylatexenc`.

First we need to do the necessary imports:

```
import numpy as np
import random
import matplotlib.pyplot as plt
from qiskit import *
from qiskit.circuit import *
from qiskit.extensions import *
from qiskit.circuit.library import *
from qiskit.extensions.simulator.snapshot import snapshot
from qiskit.quantum_info import Statevector
from qiskit.quantum_info.operators import Operator
from qiskit.extensions.simulator.snapshot import snapshot
from qiskit.visualization import plot_state_city, plot_bloch_multivector
from qiskit.visualization import plot_state_paulivec, plot_state_hinton
from qiskit.visualization import plot_state_qsphere
from qiskit.tools.visualization import plot_histogram
from qiskit.quantum_info import random_unitary
from qiskit.tools.monitor import job_monitor
import qiskit.quantum_info as qi
```

## 1.2 Introduction to Qiskit

### 1.2.1 Quantum circuit

A circuit in Qiskit is composed of quantum and classical registers. The quantum registers are used to perform the quantum computations, and the classical ones to store the classical outcome of the measurements. In Qiskit a register is always initialized in state  $|0\rangle$ .

```
n = 3
m = n
# Quantum register of n qubits
q = QuantumRegister(n, name='q')
# Classical register of m bits
c = ClassicalRegister(m, name='c')
# Quantum circuit
qc = QuantumCircuit(q,c)
# Apply X on qubit 0
qc.x(q[0])
# Apply H on qubit 0
qc.h(q[0])
# Apply a CNOT on qubit 0 controlled by qubit 2
qc.cnot(q[2],q[0])
# Put a barrier between the operations
qc.barrier()
# Measure of the register q, the results are stored in c
qc.measure(q,c)
# Drawing the circuit
qc.draw('mpl')
# Note that the display style of the circuit can be modified with:
# qc.draw('mpl',style='bw'), qc.draw('mpl',style='iqx') or qc.draw()
```

Once a quantum circuit is created, one can run it to get some outcomes.

```
backend = BasicAer.get_backend('qasm_simulator')
job = execute(qc, backend, shots=1024)
res = dict(job.result().get_counts(qc))
```

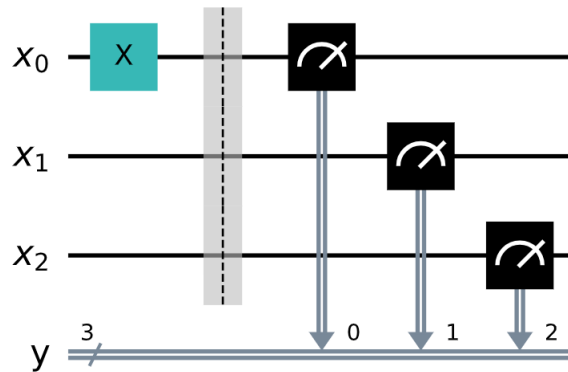
```
# Plot the results of the measurements
plot_histogram(res)
```

The variable `res` is a dictionary that contains the results of the measurements. The keys are the bit string measured and the values corresponds to the number of times this string has been measured. Hence, the dictionary `{'00':134, '01':237, '10':511, '11':118}` indicates that we measured the state  $|00\rangle$  134 times,  $|01\rangle$  237 times,  $|10\rangle$  511 times and  $|11\rangle$  118 times.

It is also possible to only measure some qubits of the system:

```
# Measure of the qubit 0 of the register q
qc.measure(q[0], c[0])
```

The top wire corresponds to the qubit encoding the least significant bit. Let  $|x_2x_1x_0\rangle = |000\rangle$  be a quantum state, we want to apply a NOT gate on  $|x_0\rangle$ , its Qiskit circuit is:



The outcome of the measurement is 001 with probability 1.

### 1.2.2 Sub-circuit

It is possible to transform a quantum circuit into a unitary gate, one can see it as a subroutine. Let's create a simple circuit:

```
q = QuantumRegister(2, name='q')
qc = QuantumCircuit(q)
qc.cnot(q[0], q[1])
qc.h(q)
```

We transform it into a unitary gate:

```
gate = qc.to_gate(label='gate')
```

We can now add it as a subroutine in other quantum circuits:

```
q = QuantumRegister(4, name='q')
qc = QuantumCircuit(q)
qc.x(q)
# We apply the gate on the qubits 1 and 2 of the new circuit
qc.append(gate, [q[1], q[2]])
# We apply the same gate on the qubits 1 and 2 of the new circuit but controlled by qubit 0
qc.append(gate.control(), [q[0], q[1], q[2]])
qc.draw('mpl')
```

Then, we can decompose the boxes we just created:

```
qc.decompose().draw('mpl')
```

We can compute the inverse, power or control gate easily:

```
q = QuantumRegister(5,name='q')
qc = QuantumCircuit(q)
# We apply the gate^2 on the qubits 1 and 2
qc.append(gate.power(2),[q[1],q[2]])
# We apply the inverse of the gate on the qubits 1 and 2
qc.append(gate.inverse(),[q[1],q[2]])
# We apply the gate on the qubits 2 and 3 controlled by qubits 0 and 1
qc.append(gate.control().control(),[q[0],q[1],q[2],q[3]])
qc.draw('mpl')
```

### 1.2.3 From unitary to gate

We can convert a unitary matrix into a quantum gate, let's create the gate  $U = \begin{bmatrix} 1 & 0 \\ 0 & e^{i\pi/7} \end{bmatrix}$ :

```
q = QuantumRegister(2,name='q')
qc = QuantumCircuit(q)
U = np.array([[1,0],[0,np.exp(1j*np.pi/7)]])
op = Operator(U)
qc.unitary(op,[q[0]],label='U')
qc.draw('mpl')
```

### 1.2.4 Run on IBM quantum computer

To run your code on IBM real quantum computer, you first have to create an account on [IBM Quantum Experience](#) to get your API Key. Once it is done, you can run the following code:

```
# List the simulators and the Quantum Computers available : provider.backends()
IBMQ.save_account('API KEY', overwrite=True)
IBMQ.load_account()
provider = IBMQ.get_provider(hub = 'ibm-q')
device = provider.get_backend('ibmq_manila')
job = execute(qc ,backend = device,shots = 1024)
# Entering the queue
job_monitor(job)
device_result = job.result()
real_counts = device_result.get_counts(qc)
plot_histogram(real_counts)
```

### 1.2.5 Usefull functions

One is encouraged to use the following function to be more efficient:

```
# Function to run a quantum circuit
def run(qc, n_shots=1024, plot=True):
    backend = BasicAer.get_backend('qasm_simulator')
    job = execute(qc,backend,shots=n_shots)
    res = dict(job.result().get_counts(qc))
    if plot:
        return plot_histogram(res)
    else:
        return res
```

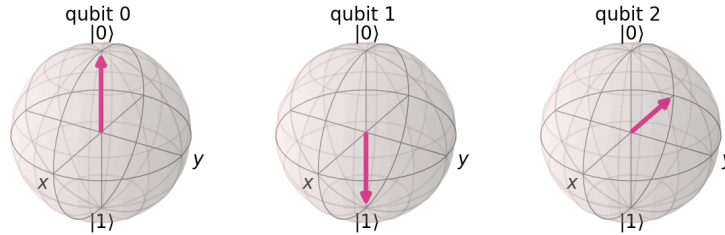
```
# Function to plot the Bloch Sphere
def bloch(qc):
    backend = BasicAer.get_backend('statevector_simulator')
    result = backend.run(transpile(qc, backend)).result()
    psi = result.get_statevector(qc)
    return plot_bloch_multivector(psi)
```

## 2 First quantum circuits

1. Plot the Bloch Sphere representation of the states  $|0\rangle, |1\rangle, \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$  and  $\frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$ . Here is the code for  $|0\rangle$ :

```
q = QuantumRegister(1)
qc = QuantumCircuit(q)
bloch(qc)
```

2. Do the same with the state  $|\psi\rangle = \frac{1}{\sqrt{2}}(|00\rangle + |01\rangle)$ .
3. Lastly, do it for the state  $|\Psi\rangle = \frac{1}{\sqrt{2}}(|00\rangle + |11\rangle)$ . Can you explain what is happening ?
4. Design a quantum circuit that creates the state  $|\psi\rangle = \frac{1}{2}(|00\rangle + |01\rangle + |10\rangle + |11\rangle)$ , run it and plot the measurement results.
5. Design a quantum circuit that creates the state  $|\psi\rangle = \frac{1}{\sqrt{2}}(|000\rangle + |111\rangle)$ , run it and plot the measurement results.
6. Design a quantum circuit that creates the state  $|\psi\rangle = \frac{i}{\sqrt{2}}(|010\rangle - |110\rangle)$ . Plot its Bloch Sphere representation, you should obtain this:



7. Using the function `random_unitary`, create a random state  $|\psi\rangle$ . Design a quantum circuit working on the state  $|\psi\rangle \otimes |0\rangle$  that swaps these two qubits into:

$$|\psi\rangle \otimes |0\rangle \mapsto |0\rangle \otimes |\psi\rangle$$

Plot its Bloch Sphere representation before and after to check the result of the computation.

8. One wishes to design a quantum circuit that performs the following transformation on  $n$  qubits:

$$\sum_{k=0}^{2^n-1} \alpha_k |k\rangle \mapsto \sum_{k=0}^{2^n-1} \alpha_k |2^n - 1 - k\rangle$$

Note that it should work for any  $(\alpha_0, \dots, \alpha_{2^n-1}) \in \mathbb{C}^{2^n}$ , not only for  $\alpha_k = \frac{1}{\sqrt{2^n}}$ . This transformation is just a permutation of the amplitudes  $\alpha_k$  over the basis states, creating some kind of “mirror quantum state”. Let’s take an example with  $n = 2$  qubits, in such a case this mapping acts as:

$$\begin{aligned} \alpha_0 |00\rangle &\mapsto \alpha_0 |11\rangle \\ \alpha_1 |01\rangle &\mapsto \alpha_1 |10\rangle \\ \alpha_2 |10\rangle &\mapsto \alpha_2 |01\rangle \\ \alpha_3 |11\rangle &\mapsto \alpha_3 |00\rangle \end{aligned}$$

- (a) Make a list `unitaries` of length  $n$  that contains random  $2 \times 2$  unitaries.
- (b) Design a quantum circuit `qc1` that creates the state  $\sum_{k=0}^{2^n-1} \alpha_k |k\rangle$ , where the  $\alpha_k$  are that of `unitaries`.
- (c) Design a quantum circuit `qc2` that creates the state  $\sum_{k=0}^{2^n-1} \alpha_k |2^n - 1 - k\rangle$ , where the  $\alpha_k$  are that of `unitaries`.
- (d) Run the quantum circuits and plot their results, both plots should mirror one another.  
It is possible to plot results from 2 quantum circuit at once with:  
`plot_histogram([run(qc1, plot=False), run(qc2, plot=False)])`

### 3 Cloning qubits ?

In quantum computing it is not possible to copy an arbitrary qubit state  $\alpha|0\rangle + \beta|1\rangle$  due to the No-Cloning theorem. Hence, such a transformation is impossible:

$$|\psi\rangle|0\rangle \mapsto |\psi\rangle|\psi\rangle$$

However, it does not mean that any kind of copy is impossible, let's try some experiments. Unknown quantum states can be prepared using the function `random_unitary`.

1. Given a quantum state  $|a\rangle$  with  $a = 0, 1$  not being in any superposition ( $|0\rangle$  or  $|1\rangle$ ), design a quantum circuit that performs the following computation:

$$|0\rangle \otimes |0\rangle \mapsto |0\rangle \otimes |0\rangle$$

$$|1\rangle \otimes |0\rangle \mapsto |1\rangle \otimes |1\rangle$$

That is:

$$|a\rangle \otimes |0\rangle \mapsto |a\rangle \otimes |a\rangle$$

2. Can you still find a circuit to do it when the second qubit is initialized to  $|1\rangle$  ?

$$|a\rangle \otimes |1\rangle \mapsto |a\rangle \otimes |a\rangle$$

3. Given a quantum state  $|a = a_n \dots a_1\rangle = |a_n\rangle \otimes \dots \otimes |a_1\rangle$  with  $|a_i\rangle = |0\rangle$  or  $|1\rangle$ , design a quantum circuit that performs the following computation:

$$|a_n \dots a_1\rangle \otimes |0\rangle^{\otimes n} \mapsto |a_n \dots a_1\rangle \otimes |a_n \dots a_1\rangle$$

4. Given two arbitrary unknown quantum states  $|\psi\rangle$  and  $|\phi\rangle$ , design a quantum circuit that does the following transformation:

$$|0\rangle \otimes |\psi\rangle \otimes |\phi\rangle \mapsto \alpha|0\rangle \otimes |\psi\rangle \otimes |\phi\rangle + \beta|1\rangle \otimes |\phi\rangle \otimes |\psi\rangle$$

for  $(\alpha, \beta) \in \mathbb{C}^2$ .

5. One wishes to apply the quantum gates  $Y$  and  $Z$  on an arbitrary state  $|\psi\rangle$  in the following way:

$$|\psi\rangle \otimes |0\rangle \otimes |0\rangle \mapsto \frac{1}{\sqrt{2}}(|0\rangle \otimes |0\rangle \otimes Y|\psi\rangle + |1\rangle \otimes Z|\psi\rangle \otimes |0\rangle)$$

Design a quantum circuit that implements this computation.

*Hint: the control-Z and control-Y gates can be used with*

`qc.cz(qubit_control, qubit_target)` and `qc.cy(qubit_control, qubit_target)`.

### 4 Single qubit interference

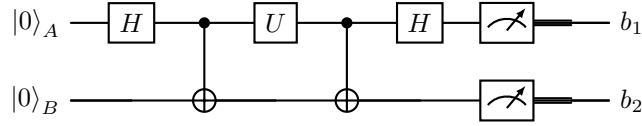
Interferences play a major role in quantum computation, by going into the Fourier basis (applying a Hadamard on a single qubit) and applying a phase it is possible to create an interesting interference scheme as depending on the value of the phase, the outcome of the measurement will change.



1. Implement the single qubit interference circuit with the phase gate  $P_\varphi = \begin{bmatrix} 1 & 0 \\ 0 & e^{i\varphi} \end{bmatrix}$ .
2. Plot the probability of measuring  $|0\rangle$  as a function of  $\varphi$ .

## 5 Superdense coding

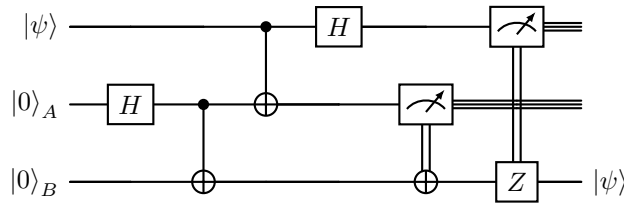
The Superdense coding protocol corresponds to the following quantum circuit:



Note that the black box  $U$  corresponds to the operations performed by Alice depending on the informations she wants to send. Write a function `Superdense_Coding(b2,b1)` that takes the bits of information Alice wants to send to Bob as inputs, and implements the corresponding Superdense coding protocol.

## 6 Quantum Teleportation

The Quantum Teleportation protocol is described with the quantum circuit:



1. Implement the Quantum Teleportation protocol.
2. Plot the Bloch Sphere of the 3 qubits system before and after teleportation. Do not forget to generate a random quantum state for the state  $|\psi\rangle$  to teleport. Note that the state before teleportation is  $|\psi\rangle \otimes \frac{1}{\sqrt{2}}(|0_A 0_B\rangle + |1_A 1_B\rangle)$ .

## 7 Quantum Fourier Transform

The Quantum Fourier Transform (QFT) is the quantum analog of the Fourier Transform, it is used as a subroutine in many quantum algorithms. It acts on a quantum state of  $n$  qubits ( $N = 2^n$ )

$$|x\rangle = \sum_{k=0}^{N-1} x_k |k\rangle$$

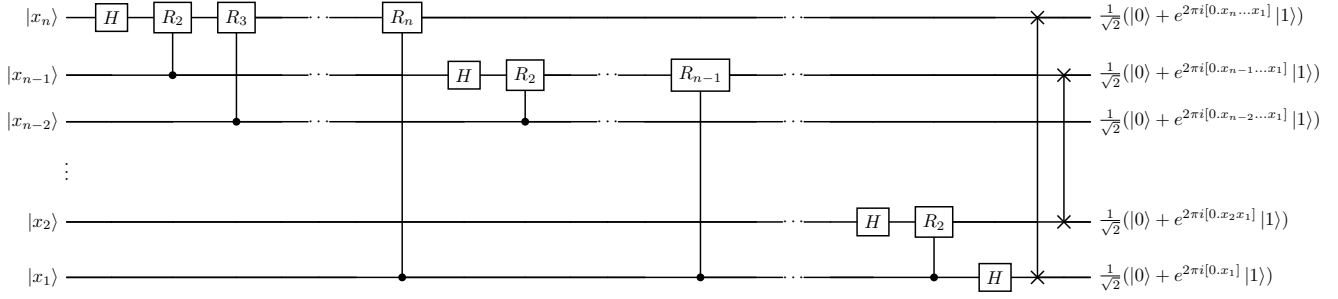
and maps it to

$$QFT |x\rangle = \sum_{k=0}^{N-1} y_k |k\rangle \text{ with } y_k = \frac{1}{\sqrt{N}} \sum_{j=0}^{N-1} x_j e^{2\pi i \frac{jk}{N}}$$

The QFT is a transform between the computational basis and the Fourier basis. The computational basis encodes information in the  $|k\rangle$  states unlike the Fourier basis that does it in the phases. For more informations, one is encouraged to click [here](#).

The quantum circuit implementing the QFT is shown bellow with  $R_k = \begin{bmatrix} 1 & 0 \\ 0 & e^{2\pi i/2^k} \end{bmatrix}$ :

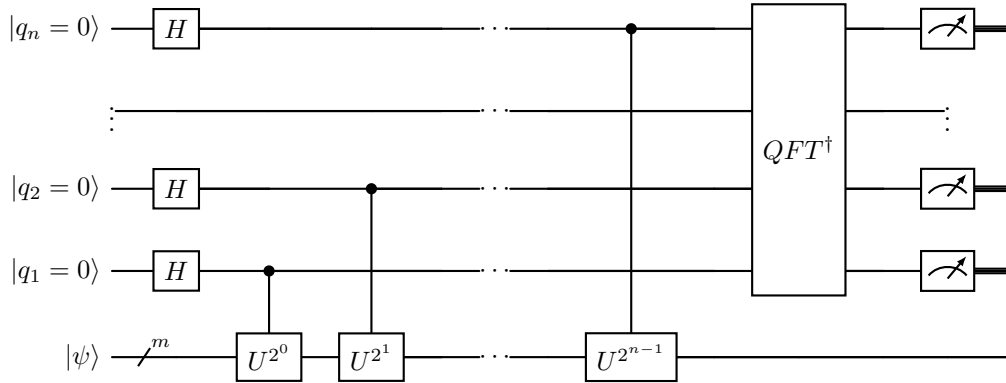




1. Write a function `QFT(n,swaps=True)` that takes the number of qubits  $n$  and a boolean `swaps` as inputs, which implements the QFT. The boolean `swaps` indicates if one should include the swap step at the end of the circuit. The controlled-rotations can be implemented with `U1Gate` or by creating the gate with a numpy array.

## 8 Quantum Phase Estimation

Quantum Phase Estimation (QPE) is a very useful subroutine for finding the eigenvalues of a unitary  $U$ . As it is unitary, its eigenvalues are of the form  $e^{2i\pi\varphi}$ . Using QPE, one can find the phase  $\varphi$  as the outcome of the measurements is  $\varphi 2^n$  with high probability. We recall that if  $|\psi\rangle$  is an eigenvector of  $U$ , then  $U|\psi\rangle = e^{2i\pi\varphi}|\psi\rangle$  for  $\varphi \in \mathbb{R}$ . It uses two quantum registers, the eigenvector  $|\psi\rangle$  (acting on  $m$  qubits) and a working register of  $n$  qubits initialized to  $|0\rangle$ . Here is the quantum circuit implementation of QPE:



1. Implement the QPE circuit with  $m = 1$  by writing a function `QPE(n,psi,U)` that takes the number of qubits  $n$  initialized to  $|0\rangle$ , `psi` a unitary to initialize the state of the bottom qubit, and a unitary  $U$  as inputs.
2. To check if your implementation is correct, we are going to estimate the value of  $\pi$  using QPE. Let  $|\psi\rangle$  be an eigenvector of  $U$ , therefore  $U|\psi\rangle = e^{2i\pi\varphi}|\psi\rangle$ . Using QPE we can easily get  $\varphi$  since  $\varphi = \frac{\delta}{2^n}$ , where  $\delta$  is the outcome of the measurement. Let's pick  $U$  such that  $U = \begin{bmatrix} 1 & 0 \\ 0 & e^i \end{bmatrix}$ . Hence, we have  $U|1\rangle = e^i|1\rangle$ , with  $|1\rangle$  and  $e^i$  being eigenvector and eigenvalue of  $U$ . We recall that any eigenvalue of a unitary matrix can be written as  $e^{2i\pi\varphi}$ , thus  $2\pi\varphi = 1$ , i.e.  $\pi = \frac{1}{2\varphi}$ . The protocol is summarized below:

- Compute the QPE circuit with  $|\psi\rangle = |1\rangle$  and  $U = \begin{bmatrix} 1 & 0 \\ 0 & e^i \end{bmatrix}$
- Get the outcome  $\delta$  by measuring the  $n$  qubit register
- Compute  $\pi = \frac{1}{2^{\frac{\delta}{2^n}}}$

- (a) Write a function `estimate_pi(n)` that uses QPE to return an estimation of  $\pi$ .
- (b) Plot the quantum estimation of  $\pi$  as a function of the number of qubits  $n$  with  $n \in [2, \dots, 15]$ .

## 9 Quantum Addition: Draper Adder

With a quantum computer it is possible to perform additions by using the QFT. Given two quantum register of  $n$  qubits each,  $|a\rangle$  and  $|b\rangle$ , we wish to compute the sum of them and to store the result in  $|b\rangle$ . Let's see how this algorithm works. Initially the state of the system is:

$$|\psi\rangle = |a\rangle \otimes |b\rangle$$

After applying the QFT on the second register it becomes:

$$\begin{aligned} |\psi\rangle &= |a\rangle \otimes \frac{1}{\sqrt{2}}(|0\rangle + e^{2\pi i(0.b_n \dots b_1)} |1\rangle) \otimes \frac{1}{\sqrt{2}}(|0\rangle + e^{2\pi i(0.b_{n-1} \dots b_1)} |1\rangle) \otimes \dots \otimes \frac{1}{\sqrt{2}}(|0\rangle + e^{2\pi i(0.b_1)} |1\rangle) \\ &= |a\rangle \otimes |\phi_n(b)\rangle \otimes |\phi_{n-1}(b)\rangle \otimes \dots \otimes |\phi_1(b)\rangle \end{aligned}$$

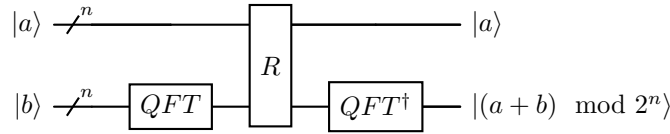
Then, we perform some rotations on  $|b\rangle$  controlled by  $|a\rangle$ :

$$\begin{aligned} |\psi\rangle &= |a\rangle \otimes \frac{1}{\sqrt{2}}(|0\rangle + e^{2\pi i(0.a_n \dots a_1 + 0.b_n \dots b_1)} |1\rangle) \otimes \frac{1}{\sqrt{2}}(|0\rangle + e^{2\pi i(0.a_{n-1} \dots a_1 + 0.b_{n-1} \dots b_1)} |1\rangle) \otimes \dots \otimes \frac{1}{\sqrt{2}}(|0\rangle + e^{2\pi i(0.a_1 + 0.b_1)} |1\rangle) \\ &= |a\rangle \otimes |\phi_n(a+b)\rangle \otimes |\phi_{n-1}(a+b)\rangle \otimes \dots \otimes |\phi_1(a+b)\rangle \end{aligned}$$

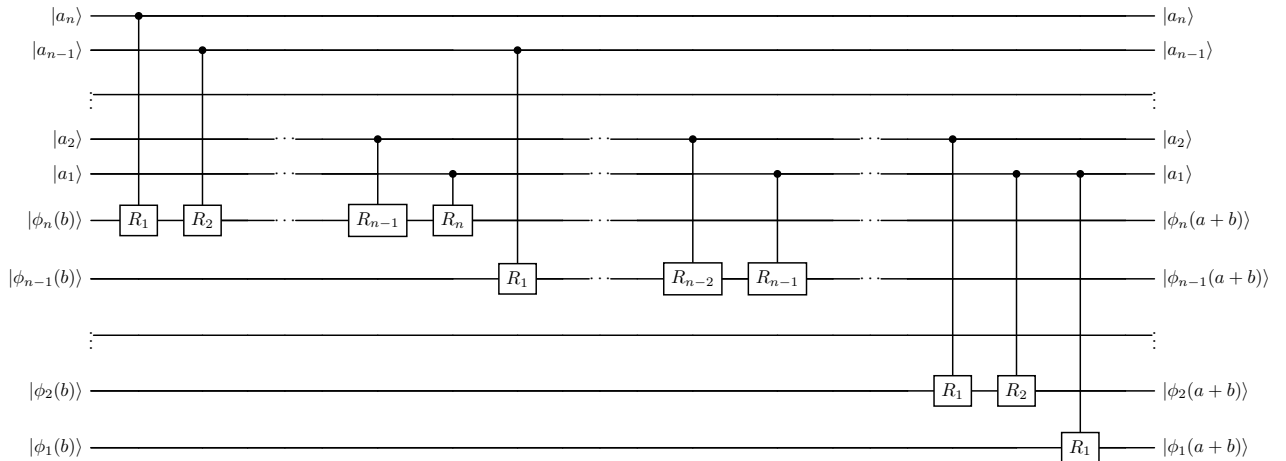
Finally, as the information is now encoded in the phases, we go back to the computational basis by applying the inverse QFT on  $|b\rangle$ :

$$\begin{aligned} |\psi\rangle &= |a\rangle \otimes |(a+b)_n\rangle \otimes |(a+b)_{n-1}\rangle \otimes \dots \otimes |(a+b)_1\rangle \\ &= |a\rangle \otimes |(a+b) \bmod 2^n\rangle \end{aligned}$$

The quantum circuit representation of the algorithm is shown bellow:



The controlled-rotations  $R$  step is shown bellow with  $R_k = \begin{bmatrix} 1 & 0 \\ 0 & e^{2\pi i/2^k} \end{bmatrix}$ :



Implement the Draper Adder. The controlled-rotations can be implemented with **U1Gate** or by creating the gate with a numpy array. **To make it work you must not do the swaps in the QFT.** Therefore, you have to set **swaps=False** when calling the QFT function. We will do the implementation in two steps;

- The initialization of the registers  $|a\rangle$  and  $|b\rangle$ .

- The implementation of the Draper Adder.

1. Write a function `Draper_Adder(x,y)` that takes two integers  $x$  and  $y$  as inputs and compute their sum using the Draper Adder. The first quantum register  $|a\rangle$  is going to store the value of  $x$  and  $|b\rangle$  the value of  $y$ . Start by initializing the two quantum registers correctly with the corresponding NOT gates. We recall that  $|x = a_n \dots a_1\rangle$  and  $|y = b_n \dots b_1\rangle$  where the index 1 corresponds to the least significant qubit. Then, implement the Draper Adder with no swaps for the QFT steps. Note that your function should work for any arbitrary non negative integers values of  $x$  and  $y$ .

You can use the line `bin(i)[2:].zfill(n)` to get the binary string representation of integer  $i$  on  $n$  bits, setting  $n$  to 0 gives the binary representation of  $i$  with the minimum number of bits. Recall that in Python the first element `s[0]` of a bit string  $s$  is the leftmost bit, you can reverse a string  $s$  with `reversed(s)` or `s = s[::-1]`.

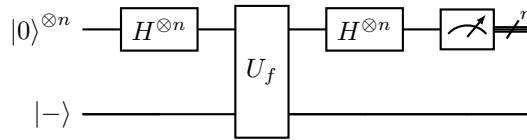
You can verify the correctness of your implementation with the function:

```
# Function to verify if the Draper Adder is correctly implemented
def verification(x,y):
    n = max(len(bin(x)[2:].zfill(0)),len(bin(y)[2:].zfill(0)))
    outcome = run(Draper_Adder(x,y),plot=False)
    return bin((x+y)%2**n)[2:].zfill(n) == list(outcome.keys())[0]
```

## 10 Bernstein-Vazirani Algorithm

In the Bernstein-Vazirani problem, one has to guess a secret number encoded on  $n$  bits. Classically the most efficient method would require  $n$  trials. Indeed, one would first do the logical operation AND between 1 and the first bit, if this bit is 1 the result would be 1, otherwise 0. Repeating this procedure for the  $n$  bits, one is able to guess the secret number with a time complexity  $O(n)$ . However, with a quantum computer one can solve this problem with only 1 trial using Bernstein-Vazirani Algorithm, hence with a constant complexity  $O(1)$ . Let's formalize the problem.

Given a function  $f : \{0,1\}^n \rightarrow \{0,1\}$  acting as an oracle where  $f(x)$  is a scalar product between  $x$  and the secret bit string  $s \in \{0,1\}^n$  modulo 2, with  $f(x) = x \cdot s = x_1 s_1 \oplus \dots \oplus x_n s_n$ . The problem is to find the value of  $s$ . The quantum circuit implementing Bernstein-Vazirani Algorithm is shown below, where  $|-\rangle = \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$ :



Let's compute the effect of the algorithm on an input state. We start with the initial state and we only focus on the  $n$  qubits of input:

$$|\psi\rangle = |0\rangle^{\otimes n}$$

After applying the Hadamard tower on the first  $n$  qubits, we obtain:

$$|\psi\rangle = \frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n-1} |x\rangle$$

The oracle  $U_f$  acts on  $|x\rangle$  as  $U_f |x\rangle = (-1)^{f(x)} |x\rangle$ . The superposition becomes:

$$|\psi\rangle = \frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n-1} (-1)^{f(x)} |x\rangle$$

Reapplying the Hadamard tower, we get:

$$\begin{aligned} |\psi\rangle &= \frac{1}{2^n} \sum_{x=0}^{2^n-1} (-1)^{f(x)+x \cdot y} |y\rangle \\ &= |s\rangle \end{aligned}$$

The final state is  $|s\rangle$  because for a given value of  $y$  we have:

$$\begin{aligned} |\psi\rangle &= \frac{1}{2^n} \sum_{x \in \{0,1\}^n} (-1)^{f(x)+x \cdot y} \\ &= \frac{1}{2^n} \sum_{x \in \{0,1\}^n} (-1)^{x \cdot s + x \cdot y} \\ &= \frac{1}{2^n} \sum_{x \in \{0,1\}^n} (-1)^{x \cdot (s \oplus y)} \\ &= 1 \text{ if } s \oplus y = 0, 0 \text{ otherwise} \end{aligned}$$

As  $s \oplus y = 0$  only if  $s = y$ , the only non zero amplitude is that of  $|s\rangle$ .

1. Write a function that generates a random bit string of length  $n$ .
2. Let's implement the oracle  $U_f$ . To do so we are going to use sequence of CNOT gates. Given a secret bit string  $s = s_n \dots s_1$ , one aims to apply a CNOT gate on the last qubit  $|-\rangle$  controlled by  $|x_i\rangle$  if  $s_i = 1$ . Recall that in Python the first element  $s[0]$  of a bit string  $s$  is the leftmost bit, you can reverse a string  $s$  with `reversed(s)` or `s[::-1]`. As we do not use this convention in our encoding, do not forget to reverse the bit string  $s$  before entering the loop.
3. Merging everything together, write a function `Bernstein_Vazirani(n)` that takes a number  $n$  as input and implements the Bernstein-Vazirani Algorithm.
4. Run your code on IBM real quantum device.

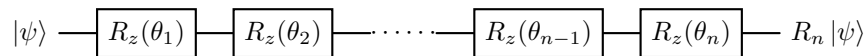
## 11 Quantum circuit complexity

Quantum circuits have 3 important characteristics: depth, width and size. These 3 notions are used to evaluate the complexity of a quantum circuit. The depth corresponds to a circuit's execution time, the width to the number of qubits it uses, and the size to the number of 1- and 2-qubit gates it employs. The biggest problem in quantum computing today is the appearance of errors in calculations. A phenomenon known as "decoherence" reduces the lifetime of qubits, introducing huge errors into the final result. One way of circumventing this problem is to reduce the execution time of a circuit by increasing the number of qubits it uses, which is a trade-off between time and resources complexity. We will investigate a way of transferring depth to width for a specific quantum circuit and see how this impacts scaling in depth and width. We will ignore the scaling of the size. We will restrict ourselves to studying this compromise for diagonal unitary matrices.

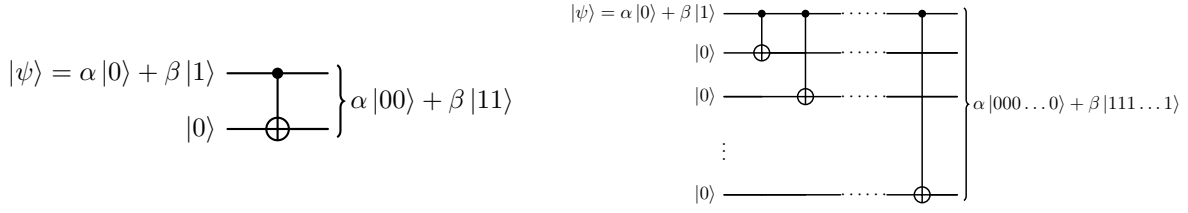
Starting from an arbitrary quantum state  $|\psi\rangle = \alpha|0\rangle + \beta|1\rangle$ , one wishes to implement the operation  $R_n|\psi\rangle$  where

$$R_n = \prod_{k=1}^n R_z(\theta_k) \text{ with } R_z(\theta_k) = \begin{pmatrix} e^{-i\frac{\theta_k}{2}} & 0 \\ 0 & e^{i\frac{\theta_k}{2}} \end{pmatrix}. \quad (1)$$

The quantum circuit representation of such an operation is:

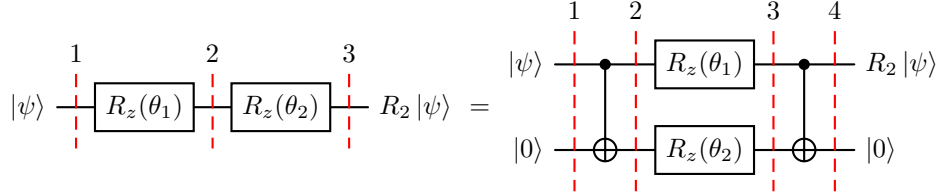


The width is 1 as we only use one qubit and the depth is  $n$ . One way to reduce its depth would be to add some ancilla qubits to execute the  $R_z$  gates simultaneously. However, those gates must act on the quantum state  $|\psi\rangle$  to perform the right computation. It turns out that one can use entanglement to "swap" the state of  $|\psi\rangle$  onto the ancillas. One can entangled an arbitrary quantum state with other qubits in state  $|0\rangle$  as:



In general, one can entangle an arbitrary quantum state  $|\psi\rangle$  with  $p$  ancillary qubits in state  $|0\rangle$  by doing  $k$  CNOTs controlled by  $|\psi\rangle$  on each ancilla. We start by computing the outcome of the first circuit:

Before going any further, let's see why it's possible to parallelize diagonal unitaries using entanglement. We want to show that:



We start by computing the outcome of the first circuit:

1.

$$|\psi\rangle = \alpha|0\rangle + \beta|1\rangle$$

2.

$$R_z(\theta_1)|\psi\rangle = e^{-i\frac{\theta_1}{2}}\alpha|0\rangle + e^{i\frac{\theta_1}{2}}\beta|1\rangle$$

3.

$$\begin{aligned} R_z(\theta_2)R_z(\theta_1)|\psi\rangle &= e^{-i\frac{\theta_2}{2}}e^{-i\frac{\theta_1}{2}}\alpha|0\rangle + e^{i\frac{\theta_2}{2}}e^{i\frac{\theta_1}{2}}\beta|1\rangle \\ &= e^{-i\frac{\theta_1+\theta_2}{2}}\alpha|0\rangle + e^{i\frac{\theta_1+\theta_2}{2}}\beta|1\rangle \\ &= R_2|\psi\rangle \end{aligned}$$

Let's see what's happening for the second circuit:

1.

$$|\psi\rangle \otimes |0\rangle = (\alpha|0\rangle + \beta|1\rangle) \otimes |0\rangle = \alpha|00\rangle + \beta|10\rangle$$

2.

$$CNOT|\psi\rangle = \alpha|00\rangle + \beta|11\rangle$$

3.

$$\begin{aligned} (R_z(\theta_1) \otimes R_z(\theta_2))CNOT|\psi\rangle + \beta|10\rangle &= \alpha e^{-i\frac{\theta_1}{2}}|0\rangle \otimes e^{-i\frac{\theta_2}{2}}|0\rangle + \beta e^{i\frac{\theta_1}{2}}|1\rangle \otimes e^{i\frac{\theta_2}{2}}|1\rangle \\ &= \alpha e^{-i\frac{\theta_1}{2}}e^{-i\frac{\theta_2}{2}}|00\rangle + \beta e^{i\frac{\theta_1}{2}}e^{i\frac{\theta_2}{2}}|11\rangle \\ &= e^{-i\frac{\theta_1+\theta_2}{2}}\alpha|00\rangle + e^{i\frac{\theta_1+\theta_2}{2}}\beta|11\rangle \end{aligned}$$

4.

$$\begin{aligned} CNOT(R_z(\theta_1) \otimes R_z(\theta_2))CNOT|\psi\rangle &= e^{-i\frac{\theta_1+\theta_2}{2}}\alpha|00\rangle + e^{i\frac{\theta_1+\theta_2}{2}}\beta|10\rangle \\ &= (e^{-i\frac{\theta_1+\theta_2}{2}}\alpha|0\rangle + e^{i\frac{\theta_1+\theta_2}{2}}\beta|1\rangle) \otimes |0\rangle \\ &= R_2|\psi\rangle \otimes |0\rangle \end{aligned}$$

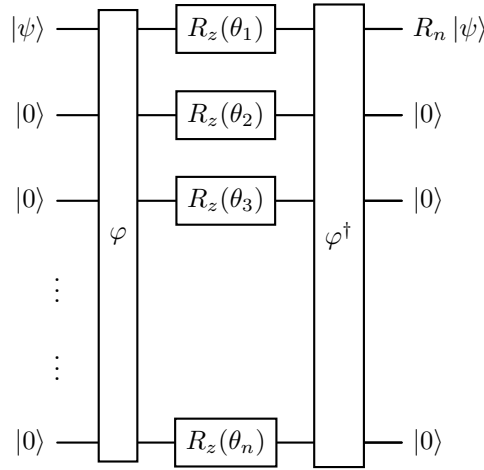
In this exercise you can compare your quantum circuit with the following function that takes two quantum circuits as inputs and outputs 1 if they both do the same computation:

```

# Function to verify if your circuits are correctly implemented
def fidelity(qc1,qc2):
    state1 = qi.Statevector.from_instruction(qc1).data[:2]
    state2 = qi.Statevector.from_instruction(qc2).data[:2]
    scalar_prod = 0 + 0j
    for i in range(len(state1)):
        scalar_prod += np.conj(state1[i]) * state2[i]
    return abs(scalar_prod)**2

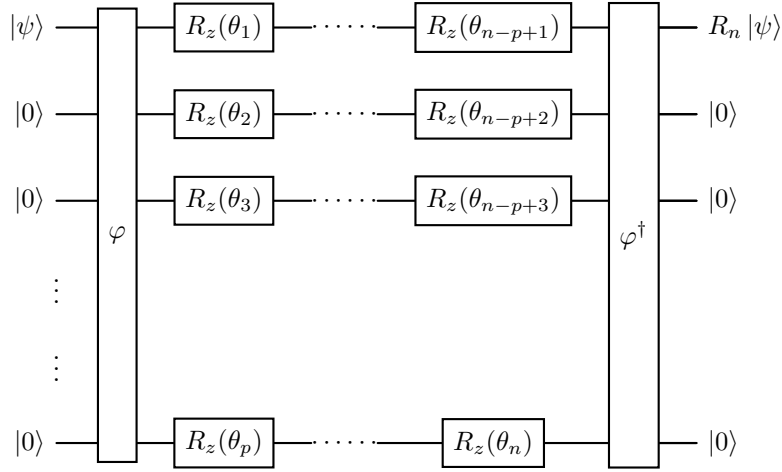
```

1. Write a function `horizontal_rz(n,psi,angles)` that takes as inputs: the number  $n$  of  $R_z$  gates, a random unitary `psi` to set the state of  $|\psi\rangle$ , and a list of  $n$  angles  $\theta_k$ . This function should implement the quantum circuit of depth  $n$  and width 1.
2. We now aim to transfer the depth into the width, that is we want to execute all the  $R_z$  gates simultaneously. First, one needs to implement the entangling step. Recall that this step consists in applying  $k$  CNOTs controlled by  $|\psi\rangle$  on each ancillary qubit.
  - (a) A naive method to perform this computation has a time complexity of  $O(n)$ . Implement it.
  - (b) Make an efficient circuit that does the same computation with a time complexity of  $1 + \lfloor \log_2(n) \rfloor = O(\log_2(n))$ .
  - (c) Plot the time complexity evolution for the naive and efficient methods for  $n \in [1 \dots 50]$ .
3. Write a function `vertical_rz(n,psi,angles)` that implement the quantum circuit executing all the  $R_z$  gates simultaneously. Hence, such a circuit uses  $n - 1$  ancilla qubits. Let  $\varphi$  be the entangling step, the quantum circuit to implement is:



Make sure to use the efficient method to perform the entangling step. The depth of this quantum circuit scales in  $O(\log_2(n))$  and its width in  $O(n)$ .

4. Remove the entangling step and compare the outcomes of `horizontal_rz` and `vertical_rz` with the function `fidelity`. You can observe that the entangling step is essential to make the algorithm work.
5. We now know how to transfer depth to width, but our two methods each propose an extreme: do the  $R_z$  rotations one after the other, or all at once. Now let's see how to make a real compromise between execution time and resource requirements. To do this, we're going to execute  $p$  gates simultaneously, so it will take  $\lceil n/p \rceil$  steps to implement them all. This quantum circuit is going to use  $p - 1$  ancilla qubits. Write a function `tradeoff_rz(n,psi,angles,p)` that implement this trade-off circuit, where `p` is the number of ancillary qubits. The quantum circuit representation is:



This quantum circuit has a depth and width scaling of  $O(\log_2(p-1) + n/p)$  and  $O(p)$ .

6. Plot the depth and width scaling for the trade-off quantum circuit with 100  $R_z$  gates. The  $x$ -axis corresponds to the number of auxiliary qubits  $p \in [0 \dots 99]$ , and the  $y$ -axis displays the depth and width. Note that you can obtain the depth and width of a quantum circuit with `qc.depth()` and `qc.width()`.
7. As one can see on Fig. 1, the depth has an interesting scaling. One may note that for a certain number of ancilla qubits, it is no longer advantageous to reduce the depth, as the quantity of ancilla qubits added in relation to the depth reduction is too high. This optimum value is reached when the depth and width of the circuit are equal. To calculate this optimal value  $p^*$ , one needs to solve the equation  $2\log_2(p-1) + n/p = p+1$ . The logarithmic term makes the equation hard to solve, so we can approximate the solution by removing it from the equation, as it is the most negligible term, and simply add a term scaling in  $O(\log_2(p-1))$  to the final solution in order to approximate  $p^*$ .

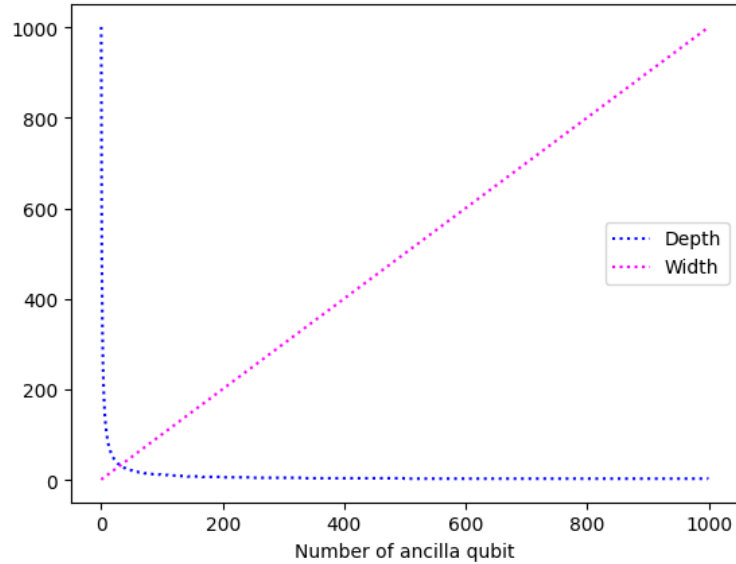


Figure 1: Depth and width scaling for the implementation of  $R_{1000}$ .

Write a function `near_optimal_rz(n,psi,angles)` that returns the trade-off quantum circuit with the approximate optimal value  $p^*$  of ancilla qubits.

8. **Bonus:** It turns out that this method only works for diagonal unitaries. For an arbitrary unitary matrix of the form  $\begin{pmatrix} \cos(\frac{\theta}{2}) & -e^{i\lambda} \sin(\frac{\theta}{2}) \\ e^{i\phi} \sin(\frac{\theta}{2}) & e^{i(\phi+\lambda)} \cos(\frac{\theta}{2}) \end{pmatrix}$  one should find another method. Do you have any idea why?