

This is a Linear Equation

$$Ax = b$$

Unknown ↴

Useful, READ VERY USEFUL
from communications to reunions

Solving a Non-Hermitian Linear System by Simulation

$$A |x\rangle = |b\rangle$$

of HHL Algorithm (in 2-dim)

Spoiler Alert: It is faster than classical in some ways

Introduction to the Problem Statement

The problem of solving differential equations has its applications in almost every science field known to man. As these systems of equations become large, solving them pushes the capabilities of classical computers.

In this project, we explored quantum computation and how it is used to solve systems of linear equations.

We take the Matrix Form of the Differential Equations in the following way:

$$A |x\rangle = |b\rangle$$

Where:-

A: a $N \times N$ Hermitian matrix

B: some N -dimensional vector with unit length

Problem Statement:- Given A and B, we need to solve the above equation for the vector $|x\rangle$.

Now, we have the the problem in the Quantum Linear System form.



The Theory Spectral Decomposition

$$A = \sum_i \lambda_i |v_i\rangle\langle v_i|$$

$$|b\rangle = \sum_i \beta_i |v_i\rangle$$

$$A|x\rangle = |b\rangle \Rightarrow$$

$$\{ |x\rangle = \sum_i \frac{\beta_i}{\lambda_i} |v_i\rangle \}$$

But life is more complicated

On the face of it we do not need eigenvalues or eigenvectors -----

eigenvalues
eigenvectors

Birds Eye View

1

Conversion of the Differential Equation into the Quantum Linear System Equation (QLSE) form.

2

Getting the Hermitian Operator and State

3

Phase Estimation

4

Inversion

5

Uncomputing to get $|x\rangle$

$$Ax = b$$

Hermitian Operator and State

$$\hat{A}|u_j\rangle = \lambda_j|u_j\rangle, \\ |b\rangle = \sum_j \beta_j|u_j\rangle$$

Phase Estimation

$$\text{Using } \hat{U}|b\rangle = e^{i\hat{A}t}|b\rangle, \\ |0\rangle^{\otimes n}|b\rangle \rightarrow \sum_j |\lambda_j\rangle \beta_j|u_j\rangle$$

Inversion

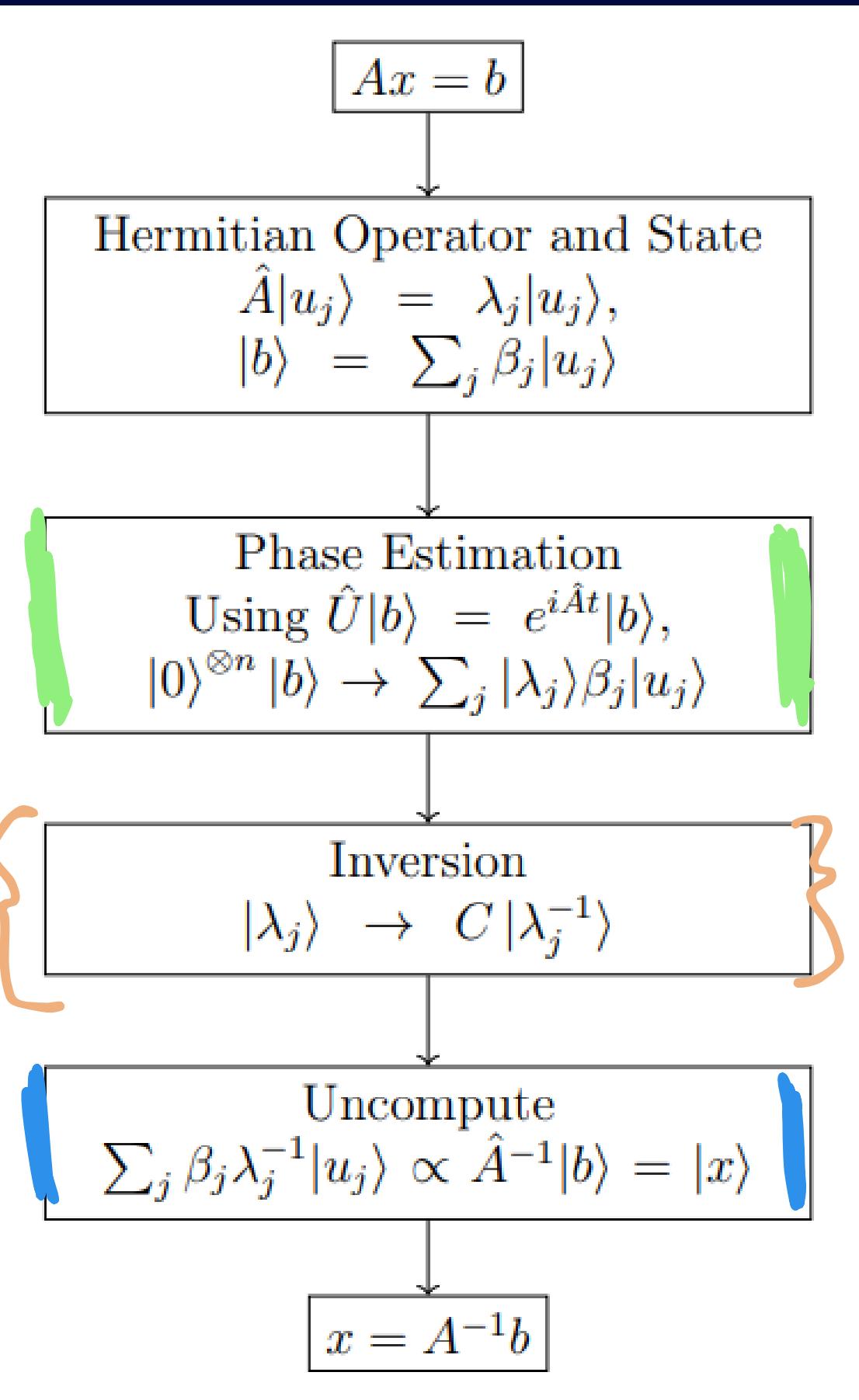
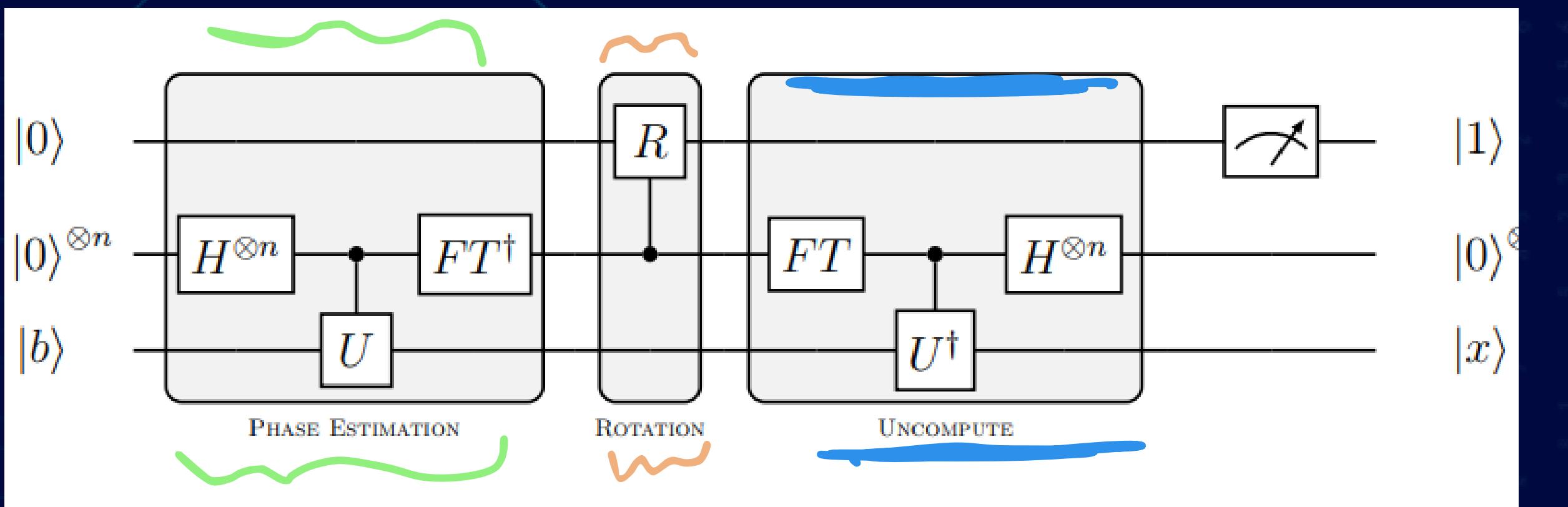
$$|\lambda_j\rangle \rightarrow C|\lambda_j^{-1}\rangle$$

Uncompute

$$\sum_j \beta_j \lambda_j^{-1}|u_j\rangle \propto \hat{A}^{-1}|b\rangle = |x\rangle$$

$$x = A^{-1}b$$

Birds Eye View

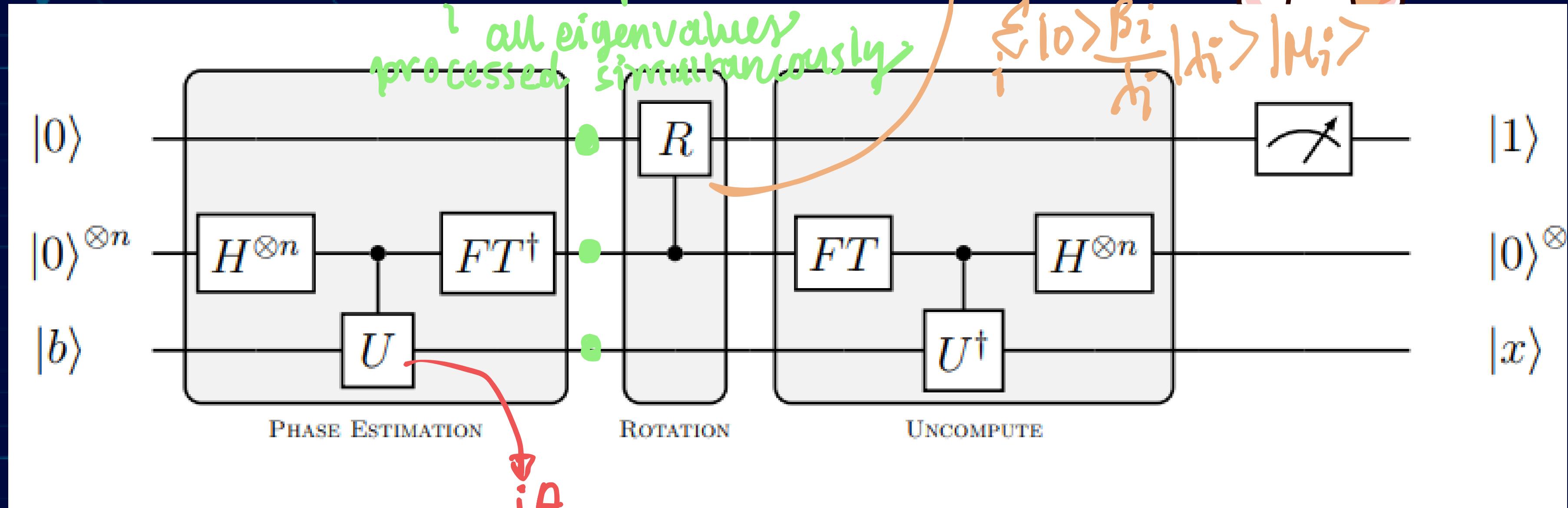


Quantum Speeding

$\{|\alpha_j\rangle\beta_j|\lambda_j\rangle|\mu_j\rangle\}$

all eigenvalues
processed simultaneously

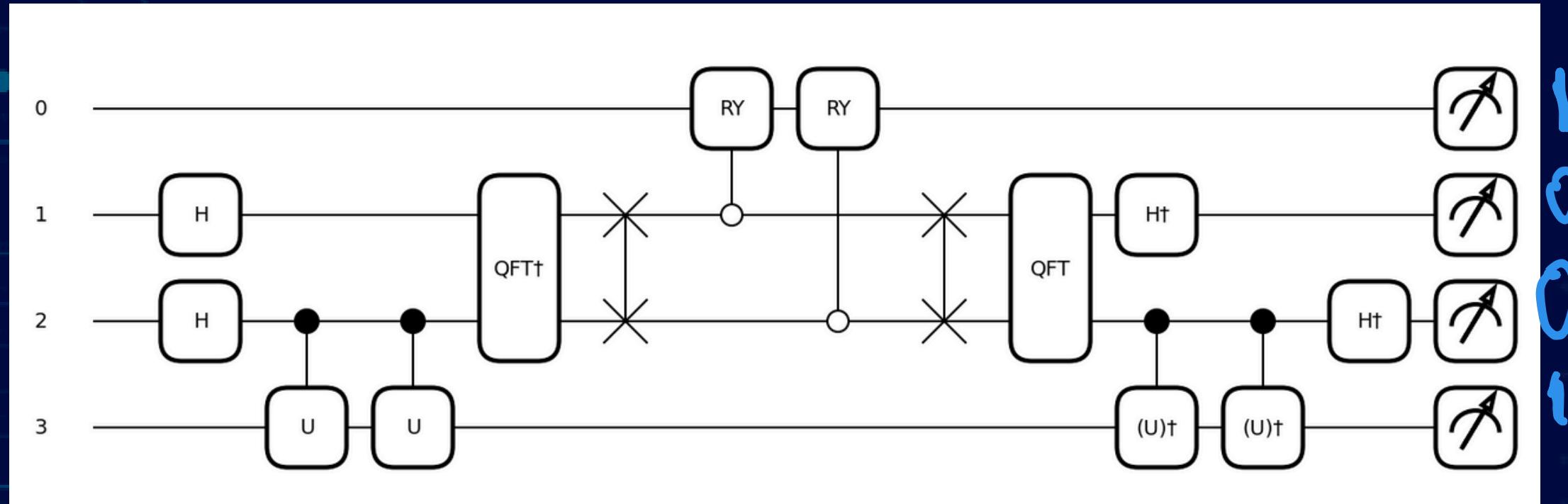
errors are
induced here



$$e^{iA}$$

Finding this oracle can be a headache

Implementation



$$\begin{pmatrix} \frac{\sqrt{3}}{2} & \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{3}}{2} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0.75 \\ -0.25 \end{pmatrix}$$

$P(|1000\rangle) = 0.858$

$P(|1001\rangle) = 0.053$

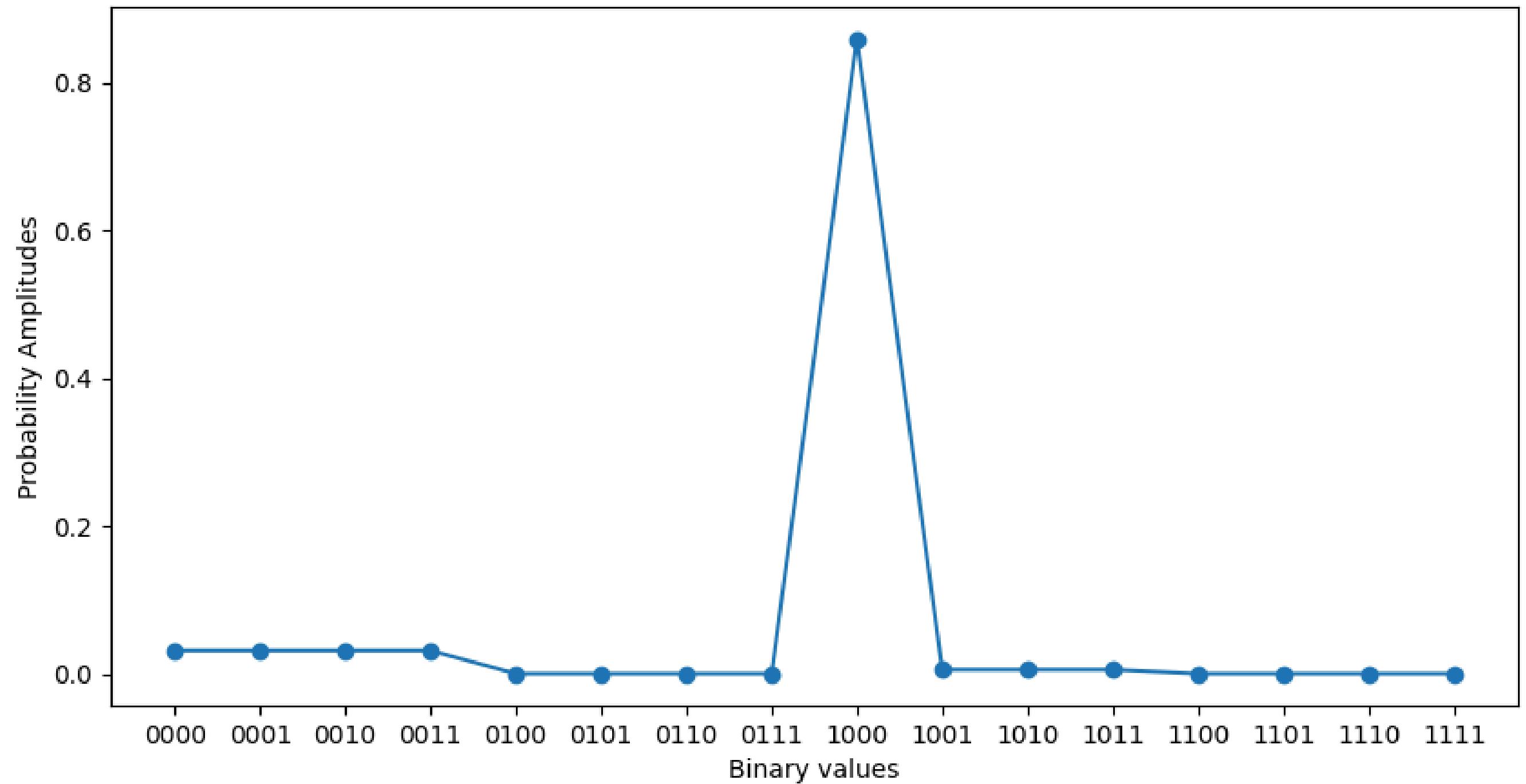
```

1 import numpy as np
2 import pennylane as qml
3 import matplotlib.pyplot as plt
4
5
6
7 dev=qml.device("default.qubit",wires=4)
8
9 U1=np.array([[-1+1j,-1-1j],[-1-1j,-1+1j]])*(0.5)
10 U2=np.array([[0,1],[1,0]])
11 @qml.qnode(dev)
12 def Part1():
13     #print(qml.state(wires=[0,1,2,3]))
14
15     qml.Hadamard(wires=1)
16     qml.Hadamard(wires=2)
17     qml.ctrl(qml.QubitUnitary(U1,wires=3),2,1)
18     qml.ctrl(qml.QubitUnitary(U2,wires=3),2,1)
19     qml.adjoint(qml.QFT)(wires=[1,2])
20
21
22 qml.SWAP(wires=[1,2])
23 qml.ctrl(qml.RY,1,0)((np.pi),wires=0)
24 qml.ctrl(qml.RY,2,0)((np.pi/2),wires=0)
25 qml.SWAP(wires=[1,2])
26
27 qml.QFT(wires=[1,2])
28 qml.adjoint(qml.ctrl(qml.QubitUnitary(U2,wires=3),2,1))
29 qml.adjoint(qml.ctrl(qml.QubitUnitary(U1,wires=3),2,1))
30 qml.adjoint(qml.Hadamard)(wires=2)
31 qml.adjoint(qml.Hadamard)(wires=1)
32 #m_0 = qml.measure(wires=0)
33 return qml.probs(wires=range(4))
34
35
36 c=Part1()
37 print(c)
38 fig, ax = qml.draw_mpl(Part1)()
39 fig.show()

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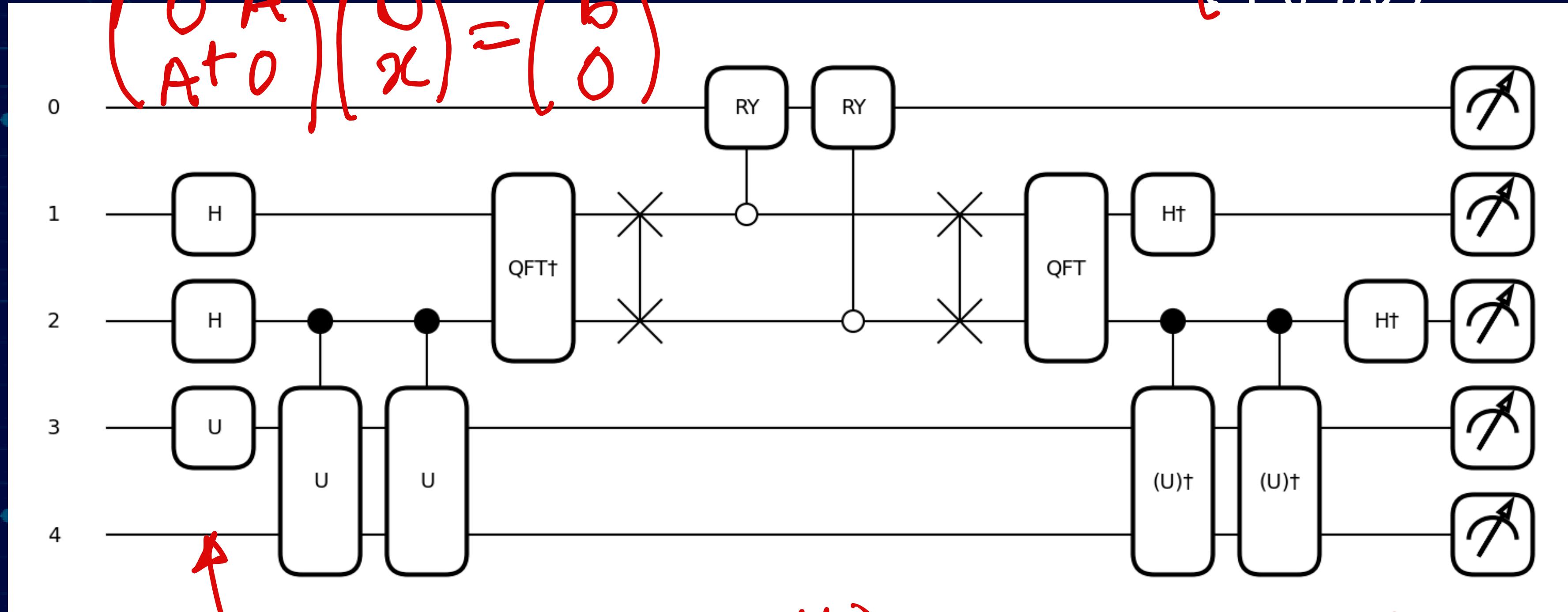
Results

Graph of the given values



Our Modifications $A' = \begin{pmatrix} 0 & A \\ A^+ & 0 \end{pmatrix} = \begin{pmatrix} 00 & 01 \\ 00 & 20 \\ 02 & 00 \\ 10 & 00 \end{pmatrix}$

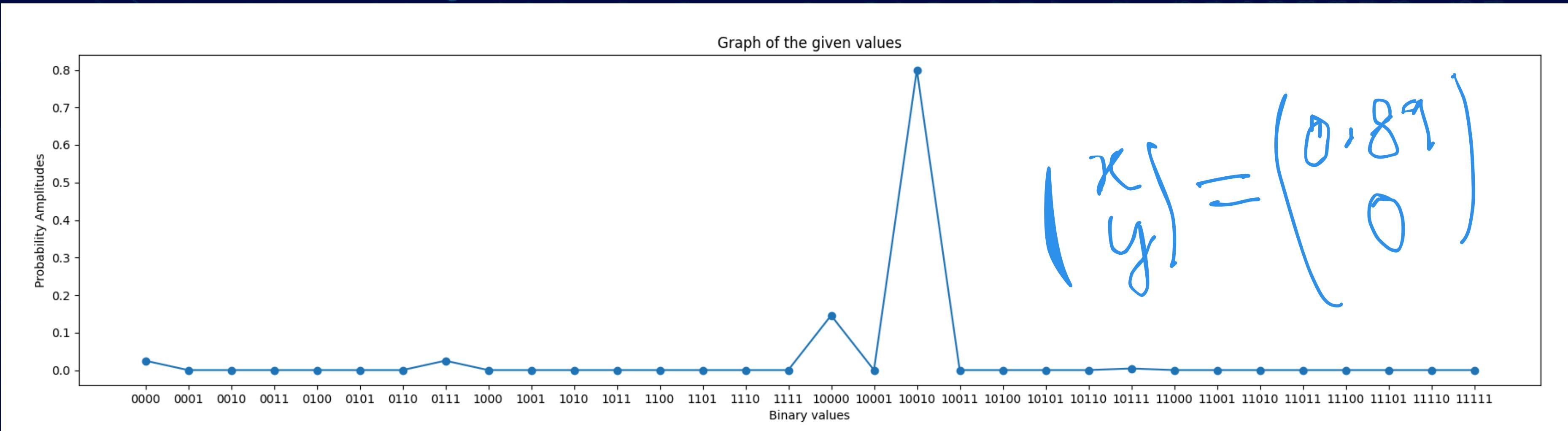
$$\begin{pmatrix} 0 & A \\ A^+ & 0 \end{pmatrix} \begin{pmatrix} 0 \\ x \end{pmatrix} = \begin{pmatrix} b \\ 0 \end{pmatrix}$$



Put the state in $\begin{pmatrix} b \\ 0 \end{pmatrix}$

$$b = \frac{1}{\sqrt{5}}|0\rangle + \frac{2}{\sqrt{5}}|1\rangle$$

Results



$$P(|1000(0)\rangle) = 0.8$$

$$|\psi\rangle = \begin{pmatrix} 0 \\ 0.8 \\ 0 \end{pmatrix}$$

APPLICATIONS OF USING QC FOR SOLVING MULTI-QUBIT DIFFERENTIAL EQUATIONS

Data Search in
Unstructured Data

Cryptography and
Security

Machine Learning and AI

Traffic and Transportation

Optimization Problems



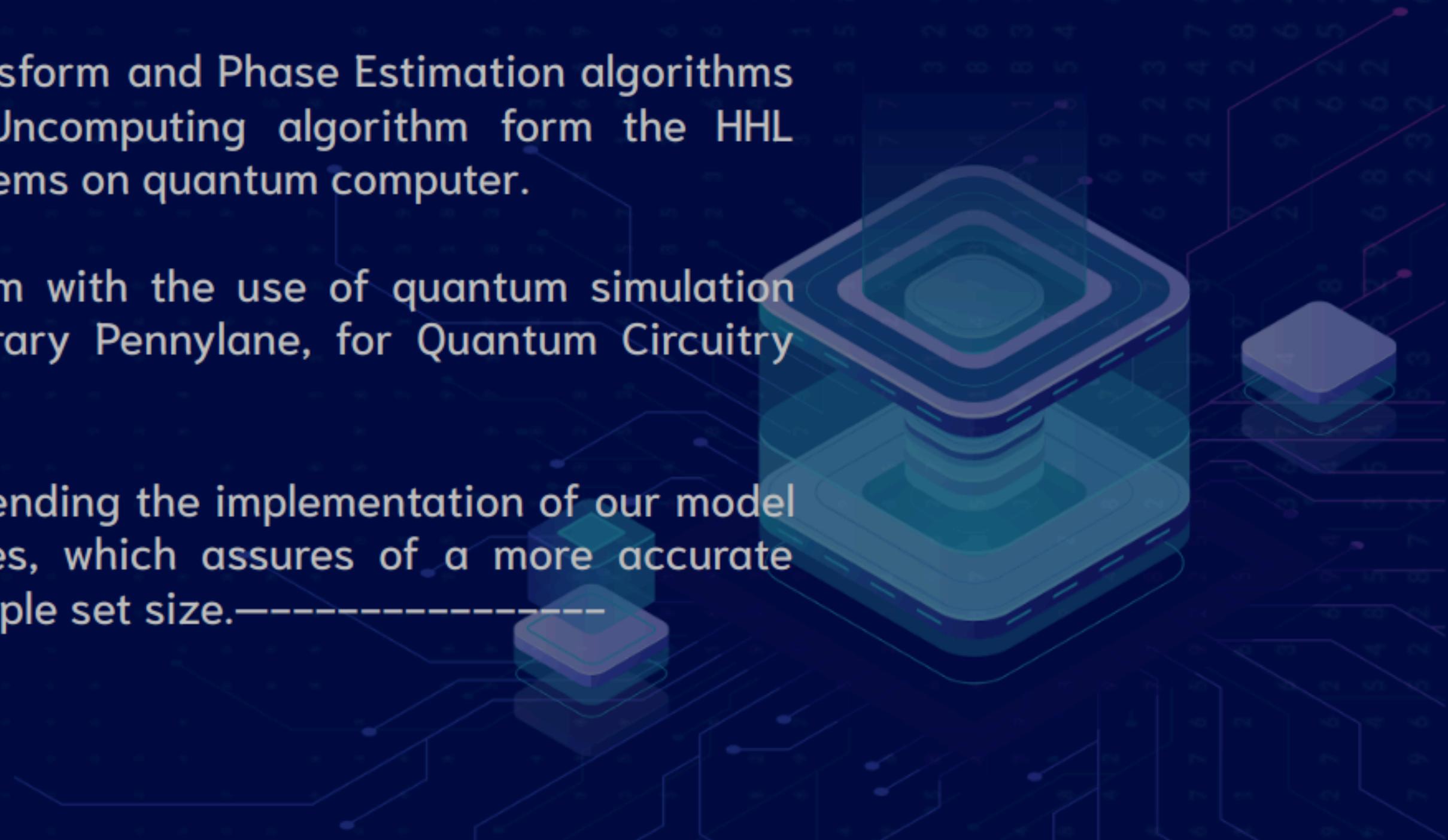
CONCLUSION

In this project presentation, we explored quantum computation and how it is used to solve systems of linear equations.

We used the Quantum Fourier Transform and Phase Estimation algorithms which in combination with the Uncomputing algorithm form the HHL algorithm for solving the linear systems on quantum computer.

We implemented the HHL algorithm with the use of quantum simulation softwares Quirk and Python Library Pennylane, for Quantum Circuitry Computation.

We then took a step further by extending the implementation of our model on higher frequency of qubit lines, which assures of a more accurate probability due to an increased sample set size.



Challenges or Opportunities

- Demonstrated superior Efficiency but only for sparse matrices
- Limited to certain kind of matrices
- Answers are constant multiples
- Rotation is more difficult to be implemented
- Mostly Eigenvalues and eigenvectors should be known to have superior efficiency

However this is a demonstration that there might be better ways to Solve Linear Equation

References

- HHL Analysis and Simulation Verification Based on Origin Quantum Platform
by Xiaonan Liu , Lina Jing , Lin Han , and Jie Gao
- USING QUANTUM ALGORITHMS TO SOLVE LINEAR SYSTEMS
by Michael C.R. Byrd Jr.

THANK YOU

