

Quantum Computing and Quantum Information Theory

Pradipta Parag Bora

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1 Introduction

Through this Summer of Science report, we attempt to summarise everything we covered related to Quantum Computing and Quantum Information Theory. We will mainly follow the book **Quantum Computation and Quantum Information** by **Nielsen and Chuang**. We will begin by going through the mathematical preliminaries, including an introduction to the linear algebraic treatment of Quantum Mechanics through which we will explore Quantum Computation. Then we will move onto qubits, their representation and on the way we will study and go through various fascinating quantum algorithms and their uses. Hopefully through this report one will be able to gain a decent understanding of Quantum Computation.

The first two sections do not deal with quantum computation and quantum mechanics and instead provide a brief overview of the mathematics that we will require along with the basic ideas of computer science. We then move onto the postulates of quantum mechanics and then we start with basic quantum circuits.

2 Mathematical Preliminaries

We assume that the reader is familiar with basic linear algebra that is covered during the first year. As our treatment of Quantum Computation heavily uses this, we will summarise some key theorems and notations that we will be using. The reader is free to skip this section as it merely summarises the mathematical tools that we will use. Also note that the following section is very direct and rarely provides any motivation behind the definitions and results to keep it concise and simply provides a list of the preliminaries. Readers are advised to refer to various references of linear algebra to get a better understanding.

2.1 Vectors and Vector Spaces

The most important structure in linear algebra is a vector which we will represent as $|\psi\rangle$. This vector resides in a vector space that satisfies some axioms that give this structure its core defining properties. This representation of a vector is famously known as the *bra-ket* notation and the $|\cdot\rangle$ is used to represent a vector. The same vector in the vector space \mathbb{C}^n is conveniently represented as a column vector.

We will be generally working in the vector space \mathbb{C}^n over the field \mathbb{C} . A set of vectors $|\psi_1\rangle, |\psi_2\rangle \dots |\psi_n\rangle$ is said to be *linearly-independent* if for any set of coefficients a_i ,

$$\sum_{i=1}^n a_i |\psi_i\rangle = 0 \implies a_i = 0 \forall 1 \leq i \leq n$$

A set of vectors $|\psi_i\rangle$ is said to be a spanning set of a vector space \mathbf{W} if any vector in \mathbf{W} can be represented as a linear combination of that set. Note that such a set may be non-finite however we will be dealing with finite dimensional vector spaces only. If this set is linearly independent then this set is called a *basis* of that vector space. It is left as an exercise to show that the cardinality of all bases are same and this value is termed as the *dimension* of that vector space.

2.2 Linear Operators and Matrices

A linear operator A from vector space \mathbf{V} to \mathbf{W} is a function mapping vectors from \mathbf{V} to \mathbf{W} satisfying linearity, that is:

$$A\left(\sum_i a_i |\psi_i\rangle\right) = \sum_i a_i A|\psi_i\rangle$$

We can then define compositions of operators, BA as function that first operates A and then operates B . It is easy to see that matrix multiplication is a linear operator. Conversely any linear operator has a matrix representation. To see this, let $A : V \rightarrow W$ be a linear operator. Let $|v_i\rangle$ be an ordered basis (basis formed by fixing order of its elements, in this case by enforcing an order restriction on $|v_i\rangle$.) of V and $|w_j\rangle$ be an ordered basis of W .

Then there exists complex numbers $A_{i,j}$ such that:

$$A|v_i\rangle = \sum_j A_{i,j} |w_j\rangle$$

If we represent any vector in V by its coordinate vector representation, that is form a column vector whose elements with respect to an ordered basis are the scalar coefficients along the basis elements and multiply it by the matrix formed by $A_{i,j}$ we get the coordinate vector representation of the output vector in the ordered basis $|w_j\rangle$.

2.3 Inner Products

An inner product of a vector space \mathbf{V} is defined as a function from $\mathbf{V} \times \mathbf{V}$ to \mathbf{C} . We will denote the inner product of $|v\rangle$ and $|w\rangle$ as $(|v\rangle, |w\rangle)$ or as $\langle v|w\rangle$. The inner product satisfies the following properties:

1. $\langle v|w\rangle = (\langle w|v\rangle)^*$
2. $\left\langle v \left| \sum_j a_j |w_j\rangle \right. \right\rangle = \sum_j a_j \langle v|w_j\rangle$
3. $\langle v|v\rangle \geq 0$ with equality *iff* $v = 0$

We can now define the *length* or *norm* of a vector $\| |v\rangle \| = \sqrt{\langle v|v\rangle}$. Two vectors are said to be orthogonal if their inner product is 0. We now define an *orthonormal* basis as a basis $|v_i\rangle$ whose elements have unit norm and they are pairwise orthogonal, that is for $i \neq j$ $\langle v_i|v_j\rangle = 0$. It can be proven that every finite dimensional vector space has an orthonormal basis. This is left as an exercise. The procedure for determining this orthonormal set from any basis is termed as the *Gram-Schmidt* orthogonalisation process.

This leads to a simple representation of the inner products of two vectors in matrix form. Let $|w\rangle = \sum_i w_i |i\rangle$ and $|v\rangle = \sum_j v_j |j\rangle$ be two vectors written with the same orthogonal basis. Then $\langle w|v\rangle = (\sum_i w_i |i\rangle, \sum_j v_j |j\rangle) = \sum_i w_i^* v_i$ which we can nicely write in matrix form as

$$\begin{bmatrix} v_1^* & v_2^* & \cdots & v_n^* \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{bmatrix}$$

We now introduce the outer product notation. Take two vectors $|w\rangle$ and $|v\rangle$ belonging to inner product spaces. Define the outer product between them to be the linear operator $|w\rangle \langle v|$ which operates on $|v'\rangle$ belonging to the same vector space as $|v\rangle$ yielding:

$$(|w\rangle \langle v|) |v'\rangle = \langle v|v'\rangle |w\rangle$$

One use of this can be seen as follows:

Let $|i\rangle$ be an orthonormal basis of a vector space \mathbf{V} . Then for any vector $|v\rangle$ in this vector space, it is easy to see that

$$\left(\sum_i |i\rangle \langle i| \right) |v\rangle = |v\rangle$$

which in turn implies that

$$\sum_i |i\rangle \langle i| = I_v$$

This is termed as the completeness relation.

Using this we can easily get the matrix representation of an operator $A : \mathbf{V} \rightarrow \mathbf{W}$. Let $|v_i\rangle$ be an orthonormal basis for \mathbf{V} and $|w_j\rangle$ be an orthonormal basis for \mathbf{W}

$$A = I_W A I_V = \sum_{i,j} |w_j\rangle \langle w_j| A |v_i\rangle \langle v_i|$$

The terms $\langle w_j| A |v_i\rangle$ are nothing but the entries of the matrix corresponding to this transformation written in coordinate vector form.

2.4 Eigenvalues and Eigenvectors

An eigenvector for a linear operator A is a non-zero vector $|v\rangle$ such that $A|v\rangle = v|v\rangle$ for some complex number v . This number is called the eigenvalue corresponding to this eigenvector and we will refer to both the eigenvector and the eigenvalue by the same letter.

A sufficient and necessary condition for a number λ to be an eigenvalue is

$$\det |A - \lambda I| = 0$$

We will term the subspace spanned by eigenvectors corresponding to the same eigenvalue as the eigenspace corresponding to that eigenvalue. Now we introduce diagonalizable operators. An operator is said to be diagonalizable on a vector space if there exists an orthonormal basis composed of eigenvectors of that operator. It is left as an exercise to show that such an operator A can be represented as:

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

where $|i\rangle$ corresponds to the orthonormal eigenvectors and λ_i are the corresponding eigenvalues. We will now look at some conditions under which operators are diagonalizable. This is of importance to us as we will be using the diagonal representation of the operator many times later on.

2.5 Adjoints and Hilbert Spaces

For any linear operator A it can be proven that there is a unique linear operator A^\dagger satisfying

$$(|w\rangle, A|v\rangle) = (A^\dagger|w\rangle, |v\rangle)$$

This operator is termed as the adjoint of A . It can be shown that the matrix representation of A^\dagger is obtained by taking the complex conjugate and then the transpose of A or $A^\dagger = (A^*)^T$.

Definition 2.1. An operator is said to be **Hermitian** if $A^\dagger = A$.

We proceed to show an example of an useful hermitian operator, the Projector. Let \mathbf{W} be a d dimensional subspace of \mathbf{V} which is k dimensional. Let $|v_1\rangle \cdots |v_d\rangle$ be an orthonormal basis of \mathbf{W} . By Gram Schmidt process we can extend this to an orthonormal basis of \mathbf{V} . The Projector $P : \mathbf{V} \rightarrow \mathbf{W}$ is defined as:

$$P = \sum_{i=1}^d |v_i\rangle \langle v_i|$$

Informally this projector projects vectors from \mathbf{V} to \mathbf{W} as it only leaves those components of the vector that are along the basis vectors which are in \mathbf{W} . We leave some trivial exercises corresponding to the Projector below that can be done by following the definitions.

Exercise 2.1. Prove that P is Hermitian

Exercise 2.2. Prove that $P^2 = P$ and thus $P^\dagger P = P$

Definition 2.2. An operator is A said to be **normal** if $A^\dagger A = AA^\dagger$.

The following theorem is incredibly useful and we omit its proof as it is rather involved:

Theorem 2.1 (Spectral Theorem). An operator A is diagonalizable if and only if it is normal.

Now we talk about a special type of normal operators, Unitary operators. These operators are of particular interest to us as all single quantum gates can be represented as unitary operators.

Definition 2.3. An operator is A said to be **unitary** if $A^\dagger A = I$.

A special class of hermitian operators is of particular interest to us. These operators called Positive operators are useful as they come up in the polar and singular decompositions which we will discuss shortly.

Definition 2.4. An operator is A said to be **positive** if $(|v\rangle, A|v\rangle) \in \mathbb{R}^+$ for all $|v\rangle$

Please go through the following exercises related to positive operators.

Exercise 2.3. Prove that for any operator A , AA^\dagger is positive.

Exercise 2.4. Prove that any positive operator is necessarily hermitian.

Solution. Let A be a positive operator. Observe that we can write A as:

$$A = \frac{A + A^\dagger}{2} + i \frac{A - A^\dagger}{2i}$$

Observe that $\frac{A+A^\dagger}{2}$ and $\frac{A-A^\dagger}{2i}$ are both hermitian operators. We will call them B and C from now on.

$$(|v\rangle, A|v\rangle) = (|v\rangle, B|v\rangle) + i(|v\rangle, C|v\rangle)$$

Observe that for any hermitian operator B ,

$$(|v\rangle, B|v\rangle) \in \mathbb{R}$$

Thus it follows that $(|v\rangle, C|v\rangle) = 0$ for all $|v\rangle$ which implies $C = 0$. Thus $A = B$ and so A is hermitian.

2.6 Operator Functions

It becomes useful for us to define functions on operators that are normal. Formally if A is a normal operator and its diagonal representation is as follows:

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

then we define

$$f(A) = \sum_i f(v_i) |v_i\rangle \langle v_i|$$

It is easy to see that $f(A)$ is uniquely determined which allows us to define things like square roots of positive operators, exponentials of normal operators etc.

Exercise 2.5. If A is a normal operator satisfying $A^2 = I$ then prove that

$$\exp(i\theta A) = \cos \theta I + i \sin \theta A$$

Solution. Observe that the possible eigenvalues are 1 and -1 . Let $|v_j\rangle$ be an orthonormal basis of the eigenspace corresponding to the eigenvalue 1 and $|w_k\rangle$ be the orthonormal basis for the eigenspace corresponding to -1 . Then,

$$A = \sum_j |v_j\rangle \langle v_j| - \sum_k |w_k\rangle \langle w_k|$$

From the definition of the exponential for normal matrices,

$$\begin{aligned} \exp\{i\theta A\} &= \sum_j \exp\{i\theta\} |v_j\rangle \langle v_j| + \sum_k \exp\{-i\theta\} |w_k\rangle \langle w_k| \\ &= \cos \theta \left(\sum_j |v_j\rangle \langle v_j| + \sum_k |w_k\rangle \langle w_k| \right) + i \sin \theta \left(\sum_j |v_j\rangle \langle v_j| - \sum_k |w_k\rangle \langle w_k| \right) = \cos \theta I + i \sin \theta A \end{aligned}$$

where we have used the completeness relation and the diagonal decomposition of A .

We briefly mention that the notion of trace for a matrix can be extended for operators. This is because the trace satisfies

$$\text{tr}(AB) = \text{tr}(BA)$$

for matrices A, B . Suppose A' is one matrix representation of A . Then it is related to another matrix representation by A'' by

$$A' = P A'' P^{-1}$$

for some change of basis matrix P . Clearly by the previous result $\text{tr}(A') = \text{tr}(A'')$ and so trace of an operator is independent of its matrix representation.

2.7 Polar and Singular Value Decompositions

We omit the proof of the Polar Decomposition as it is rather involved. The singular value decomposition follows from the polar decomposition.

Theorem 2.2. Polar Decomposition For any linear operator A on a vector space \mathbf{V} there exists unitary U and positive J, K such that

$$A = UJ = KU$$

where

$$J = \sqrt{A^\dagger A}, K = \sqrt{AA^\dagger}$$

Theorem 2.3. Singular Decomposition For any square matrix A there exists unitary matrices M, N and diagonal matrix D with positive entries such that

$$A = MDN$$

Proof. By the polar decomposition, there exists unitary U and positive J such that

$$A = UJ$$

As J is positive, it is diagonalisable, so

$$J = T^\dagger DT$$

where the entries of D are the eigenvalues of J which are positive and T is unitary. Setting $M = UT^\dagger$ and $N = T$ completes the proof. \square

We end the mathematical preliminaries at this stage and move onto the postulates of Quantum mechanics which will provide an axiomatic approach to quantum mechanics.

3 Introduction to Computer Science

Our main motivation behind using quantum computation is to provide an improvement over the classical model of computation. In order to do so we must have a proper notion of how we compare two different models of computation. Many classically difficult to solve problems have in principle quantum algorithms that solve these problems with ease. For instance, the factorisation problem of finding the prime factors of a number N , takes exponentially more time classically than via the quantum Shor's algorithm. But how do we quantify which algorithm is better or efficient? We will deal with this at the end of this section. We begin by first briefly discussing a model of computation that is deemed to be universal, the Turing model. We then provide a circuit equivalent of the same which is more feasible to us. We then look into how we can compare algorithms and finally end this section by describing different classes of algorithms on the basis of their complexities.

3.1 Turing Machines

The field of computer science arose back in the 1930s when the pioneers of the field Alan Turing and Alonzo Church formalized the notion of an algorithm used to solve mathematical problems by providing a model of computation. Turing's model, called the Turing Machine provides an abstract notion of a machine, that in principle can solve all problems associated with an algorithmic process. We will discuss its construction now.

A Turing machine contains four main elements:

1. a program, rather like an ordinary computer which contains the instructions corresponding to the algorithm
2. a finite state control, which acts like a stripped-down microprocessor working with the other components and containing the states of the machine
3. a tape which is the memory/output of the machine
4. a read write tape head, which points to the current position on the tape which is being accessed.

The main component of the turing machine is the finite state control, which contains a collection of states of the machine and the machine works by trying to look for the instruction in this collection that corresponds to the current state the Turing machine is in. Let the states of the machine be $q_1, q_2 \dots q_m$ and two other special states q_s and q_h which correspond to the starting and halt state respectively. The machine stops operating whenever q_h is obtained.

The tape can be thought of as a linear sequence of squares where each element is either 0, 1, b or s where b corresponds to a blank square and s refers to the left most starting square on the tape. The tape will start with s and then contain some zeros and ones and finally have blanks on it. The read write tape head is responsible for pointing at the current square and overwriting it with the new value on the square as the machine moves onto the next square. This can be regarded as both the input and the output of the device. Let us see how the turing machine works. We will illustrate this via an example. We'll show an explicit construction of a turing machine that outputs the function $f(x) = 1$. Firstly let us see how the program of the machine is defined. Formally a *program* in a turing machine is a list of tuples (which correspond to commands) of the form

$$(q_i, x, q_j, x', d)$$

where q_i is the current state in which the machine is supposed to be, x is the value on the current square in the tape head, q_j is the state the machine goes into after executing the current command, x' is the value that the machine will write on the current square after this command is executed and d is either 0, +1, -1 where a value of 0 instructs the machine to not move the tape head, +1 means move the pointer to the right and -1 means move the tape to the left. The only exception to this rule is when the current value on the tape is s which would ignore the -1 instruction as we are at the beginning of the tape.

The machine will obtain its current working condition by reading the value on the tape square and scan for the program line that has its current state and this value as the first two elements of the tuple. If such a

line does not exist it automatically goes to q_h and halts execution.

Now we will list an explicit program for computing $f(x) = 1$. Consider the following program:

1. $(q_s, s, q_1, s, 1)$
2. $(q_1, 0, q_1, b, 1)$
3. $(q_1, 1, q_1, b, 1)$
4. $(q_1, b, q_2, b, -1)$
5. $(q_2, b, q_2, b, -1)$
6. $(q_2, s, q_3, s, 1)$
7. $(q_3, b, q_h, 1, 0)$

We can see that this program outputs $f(x) = 1$ followed by a series of blanks. It is upto you to verify this by tracing the working of the machine line by line. Although it seems that we certainly did not require such a complex setup to simply compute $f(x) = 1$ the beauty of turing machines lie in the fact that this process can be used to compute various functions ranging from addition, polynomial evaluation to complex arithmetic. Infact all operations that you can do on your current computer can be done on a turing machine. The efficiencies may be different but we are more interested in knowing which problems can indeed be solved on a turing machine. Church and Turing came up with their thesis, the Church Turing thesis that answers exactly this question.

Thesis. Church Turing Thesis: The class of functions computable by a Turing machine corresponds exactly to the class of functions which we would naturally regard as being computable by an algorithm.

It is to note that this is a hypothesis that has not been proven, however over the last 60 years no counter evidence for the same has been found. It is interesting to note that quantum computers, and in general, the quantum model of computation do not change the class of functions computable by the model, it merely provides a more efficient way to do so.

Showing that the turing model of computation universally reflects what it means to have an algorithm compute a function is beyond the scope of this report. We will now go through some variations of turing machines. For instance, instead of having just one single tape, the machine could contain two tapes wherein you can use one for the input and other for the output. This variation is identical to the simple turing machine in the sense that both can compute the same set of functions. However for explicitly writing programs for a turing machine it is indeed more convenient to work with the two tape version. A program line of this turing machine will be of the form $(q, x_1, x_2, q', x'_1, x'_2, d_1, d_2)$ where we will use subscript 1 for the first tape and 2 for the other tape.

Exercise 3.1. Write the program lines for a Turing Machine to compute the binary NOT of a binary number provided to you as input in a tape. (Hint: for such problems it makes more sense to use multiple taped turing machines and more symbols than just $0, 1, b, s$)

Exercise 3.2. Describe a Turing program to reverse a binary number provided to you as input in the tape.

Another version of a turing machine is a probabilistic turing machine. This machine will transition from one state to another by randomly choosing a state from various possibilities. Note that a deterministic turing machine can in essence simulate a probabilistic machine by going through all possibilities (searching through the search space). Thus they both can compute the exact same class of functions. The probabilistic machine may be more efficient but right now we are only interested in knowing the differences in the set of possible functions computable by different computation models.

Now we describe the circuit model of computation in brief.

3.2 Circuit Model of Computation

The circuit model of computation is essentially the model we are most familiar with, wires and gates that compute binary functions. Taking multiple inputs we can actually compute all kinds of functions which are equivalent to those that can be simulated on a Turing machine. In general our circuits will have gates, which are blocks that compute functions of the form $f : \{0, 1\}^k \rightarrow \{0, 1\}^l$. However we would like to reduce all our circuits to those made up of some universal blocks. There are a few gates called the universal gates in classical computation that can build up any gate. One such gate is the NAND gate which takes two binary inputs and computes their binary AND and then negates the result.

There are other universal gates as well and some common gates like OR, AND, XOR which are self explanatory. Two other gates are the FANOUT and crossover gates. The FANOUT gate takes an input and copies it out. We will later see that the FANOUT gate has no quantum equivalent because of the *no-cloning* theorem. The crossover gate takes two inputs and simply swaps them.

Later on when we get into quantum circuits, we will show quantum equivalents of these gates and also look at universal quantum gates. It is important to note that quantum mechanics is reversible, in fact all of quantum mechanics can be represented as unitary transformations which we will look into in the next section. As a result it is not possible to have irreversible gates in our quantum computational model. So a quantum variant of an AND gate is not possible as AND is irreversible which means that given the output of the AND gate we cannot determine what the inputs were. We reserve this discussion for later when we start dealing with quantum circuits.

3.3 Analysis of Computational Problems

So far we have said that quantum computers provide a more efficient solution of various classical problems. How do we deem which algorithm is better? This section deals with that.

First we talk about orders of magnitude of functions. Let us take an algorithm which scans a list of n numbers and outputs the maximum among them. Clearly such an algorithm takes n operations to finish. In general lesser the number of steps the better the algorithm. But we cannot really talk about the exact number of steps and instead we need an estimate on the number of steps as this value can be variable. Suppose we have an algorithm that scans a list of n numbers to check if 1 is present or not and if it finds 1 it terminates. Clearly this can take any number of steps ranging from 1 to n . What we know is that at worst it takes n steps (the upper bound) and in the best case it takes 1 step (lower bound). So the ideas of upper bound and lower bound on the number of steps provide a rough estimate of how good the algorithm is while average case analysis provide a better picture by accounting for all cases. To talk about such cases in a more formalised manner we use the big-O, big-Omega and big-theta notation.

Definition 3.1. Big O A function $f(n)$ is said to be $O(g(n))$ if there exists some constant c and some n_0 such that for all $n \geq n_0$,

$$f(n) \leq cg(n)$$

An algorithm A is said to have (upper) time complexity $O(g(n))$ if the function $\mathbf{TIME}(A(n)) = f(n)$ is $O(g(n))$ where $f(n)$ is the number of steps the algorithm takes on that input. For example the previous algorithm is $O(n)$ as for $c = 1$ and $n_0 = 1$ $f(n) \leq n$ is satisfied where n is the size of our input.

Similarly we have big theta and big O notations that deal with different bounds.

Definition 3.2. Big Theta A function $f(n)$ is said to be $\Theta(g(n))$ if there exists some constant c_1, c_2 and some n_0 such that for all $n \geq n_0$,

$$c_1g(n) \leq f(n) \leq c_2g(n)$$

Definition 3.3. Big Omega A function $f(n)$ is said to be $\Omega(g(n))$ if there exists some constant c and some n_0 such that for all $n \geq n_0$,

$$cg(n) \leq f(n)$$

So for instance, the previous algorithm is $\Omega(1)$ and $O(n)$. It is upto you to prove that we cannot have any Θ bound for this algorithm. We can also have space complexity where instead of talking about the number of steps, we refer to the number of memory units the program requires. Therefore the previous algorithm

has $\Theta(n)$ space complexity as it is using roughly n memory units to store all values (measured as multiples of memory units used to store one number).

Now let us compare the algorithms for prime factorisation. If a number N is given the size of its input is $\log_2 N$. Remember that computer scientists always deal with logarithms to the base 2 as values are stored in binary. So the implicit base is 2. Now Shor's algorithm is $O(\log^3 N)$ which is a massive improvement over classical algorithms. A simple trial algorithm is $O(N)$ which becomes exponential in the input size $\log_2 N$. Notice that we are only concerned with the behaviour of the function at large values of the input as computation becomes infeasible only at those limits.

Exercise 3.3. Consider the Merge Sort algorithm. This algorithm can be defined in pseudocode as follows in recursive form (self calling):

```

MERGESORT(List A, SIZE N):
    List B = List A[0, N/2]
    List C = List A[N/2 + 1, N]
    B = MERGESORT(B, N/2)
    C = MERGESORT(C, N/2)
    A = MERGE(B, C, N/2, N/2)
    return A

```

The merge function combines the two lists B and C which have half the number of elements that are already sorted and outputs the sorted list formed by combining B and C.

1. Give an $O(N_1 + N_2)$ complexity algorithm for MERGE(List A, List B, Size N_1 , Size N_2).
2. Using the above algorithm write the recursive definition for **TIME**(List A, Size N) in terms of N. Use this to compute the time complexity of Merge Sort.

3.4 Complexity Classes

When we covered big-o notation to compare running times of different algorithms we found a way of quantifying the upper bound of the running time. Naturally it makes sense to now categorize different algorithms into different classes on the basis of their running time.

We will start by classifying decision problems, that is problems that have a yes/no answer. These problems provide a lot of insight into the generic case of computable problems and so we stick to them for this report. For instance we may ask the question whether a number n is a prime or not. This is a decision problem as it has a pure yes/no answer.

To formalize this notion of a decision problem, we rephrase it in terms of a *language* L. A language is a subset of all possible strings formed from a subset of characters which we term as alphabets. For instance we may define the *prime* language L_P as the subset of all possible strings formed from $\{0, 1\}$ such that the corresponding binary number is a prime.

Then we can rephrase decision problems into creating a Turing Machine (or an algorithm) that will output yes if the corresponding input is part of that language and else no. These outputs can be two end states of the turing machine q_y, q_n .

A complexity class is a set of problems that satisfy a particular property with respect to its time or space or energy complexity. For an algorithm (or a machine implementing it) A solving a decision problem D we say that $D \in \mathbf{P}$ where \mathbf{P} is a complexity class if for an input of size n , $\mathbf{TIME}(A(n)) = O(n^k)$ for some $k > 0$. Roughly \mathbf{P} is the collection of those decision problems where the decision can be found out in polynomial size of the input.

\mathbf{P} is an important complexity class as it contains those problems that we can solve in a reasonable amount of time compared to say exponential or factorial growth algorithms.

Another important complexity class is \mathbf{NP} . Roughly speaking it is that collection of decision problems whose truth value can be verified with the help of another input called the witness W in polynomial size of the main input. Formally a decision problem is in \mathbf{NP} if

1. given $x \in L$, there exists a witness W such that using the W a turing machine can deterministically state that $x \in L$ by going to the positive state q_y in time proportional to a polynomial in size of x.

2. given $x \notin L$, for all possible witnesses W the turing machine will deterministically state that $x \notin L$ by going to the state q_n in time proportional to a polynomial in size of x .

For example consider the decision problem D asking whether a number x is **NOT** in the prime language, that is $x \in L'_P$ that we previously described. Then a possible witness could be a factor of the number x where we check if the witness divides the number (where the witness is not x or 1). If x was a prime then for any possible witness we cannot use it to say that witness divides the number and so the second condition is met. Thus this problem is in **NP**.

In general it is not easy to provide a witness for the complement of a language. For example in the previous case the complement would be the language of primes. There, coming up with a witness that gives a guarantee that $x \in L_P$ is not easy.

Exercise 3.4. Prove that $\mathbf{P} \subseteq \mathbf{NP}$ (Hint: it is not necessary to use the witness always)

It is not yet known whether $\mathbf{P} \neq \mathbf{NP}$ and this is infact a millennium problem. There are other complexity classes related in terms of space. **PSPACE** refers to those problems which take polynomial space while **EXSPACE** take exponential space. Similarly we have **L** for logarithmic time complexity and **LSPACE** for logarithmic space complexity.

Exercise 3.5. Prove that $\mathbf{P} \subseteq \mathbf{PSPACE}$

With respect to probabilistic turing machines, we have **BPP** which contains all those decision problems which we can be probabilistically solved in polynomial time with probability atleast k where $k > \frac{1}{2}$. However this k can be made as close to 1 as we desire by running the algorithm many times for the same input and choosing the maximum occurring result. Thus **BPP** is more indicative of "fast" solvable problems than **P** in practice. We end this section by allauding to another complexity class of particular interest to this, **BQP**. **BQP** is the quantum equivalent of **BPP**, that is those decision problems that can be solved with a high degree of probability $> \frac{1}{2}$ using a quantum circuit taking time which is polynomial in the size of the input.

4 Postulates of Quantum Mechanics

We aim to have a self sufficient understanding of Quantum Mechanics and in order to do so we will list the postulates of quantum mechanics in this section and see how they can be useful in our study of quantum computation.

4.1 The State Space

Postulate 1. Associated to any isolated physical system is a complex vector space with inner product (that is, a Hilbert space) known as the state space of the system. The system is completely described by its state vector, which is a unit vector in the system's state space.

This postulate states that any system can be described as a vector spanned by some basis elements. Knowing a basis for our vector space allows us to therefore express the entire system at any point of time in terms of these elements and the transformation of this representation with time is sufficient for us to describe how the system evolves with time.

It is imperative to note that this postulate merely states the existence and does not provide any knowledge about what this state space is. For this we need to develop additional quantum field theories that model different scenarios.

Let us take a simple example. We introduce the qubit at this point of time. Classical computers always process data in forms of the bits 0 and 1. But any bit at any point of time can either be 0 or 1. Quantum Computers however allow superposition of these bits.

A Qubit can therefore be one of the possible states, that is an element of the state space. The basis elements of this state space are taken as 0 and 1, which we write in vector form as $|0\rangle$ and $|1\rangle$. We can therefore write any qubit $|\psi\rangle$ as

$$|\psi\rangle = a|0\rangle + b|1\rangle$$

The postulate dictates that $|\psi\rangle$ is a unit vector. This ensures that $|a|^2 + |b|^2 = 1$. This must hold even after any transformation is done on the current state vector.

Note that our choice of using $|0\rangle$ and $|1\rangle$ as the basis is arbitrary. We can use other bases as well. The basis we use is referred to as the computational basis.

4.2 Evolution of the System

Postulate 2. The evolution of a closed quantum system is described by a unitary transformation. That is, the state $|\psi_1\rangle$ of the system at time t_1 is related to the state $|\psi_2\rangle$ of the system at time t_2 by a unitary operator U which depends only on the initial and final time.

$$|\psi_2\rangle = U|\psi_1\rangle$$

In our study of quantum computation we will be using various gates. We can consider a gate as a machine that takes in a quantum state and churns out another at a later time. The above postulate thus allows us to model all gates in the form of unitary transformations. It however does not tell us which unitary operators can be used to describe evolution of a system.

It turns out that for quantum gates on a single qubit any unitary operator can be used to describe evolution via a suitable quantum gate. We describe some gates here in brief. Consider a quantum NOT gate. This must take $|0\rangle$ to $|1\rangle$ and viceversa. This operator is simply:

$$X = |0\rangle\langle 1| + |1\rangle\langle 0|$$

which in matrix form is simply $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

Similarly we have the phase flip gate Z which takes keeps $|0\rangle$ the same and takes $|1\rangle$ to $-|1\rangle$. Another useful quantum gate is the Hadamard Gate,

$$H = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

Exercise 4.1. Verify that the Hadamard Gate is indeed unitary and find it's eigenvectors and thus diagonalise it.

4.3 Measurement

Postulate 3. Quantum measurements are described by a collection $\{M_m\}$ of measurement operators. These are operators acting on the state space of the system being measured. The index m refers to the measurement outcomes that may occur in the experiment. If the state of the quantum system is $|\psi\rangle$ immediately before the measurement then the probability that result m occurs is

$$p(m) = \langle\psi| M_m^\dagger M_m |\psi\rangle$$

After this measurement the new state vector of the system (given that we obtained m) is

$$\frac{M_m |\psi\rangle}{\sqrt{\langle\psi| M_m^\dagger M_m |\psi\rangle}}$$

These measurement operators must also satisfy the completeness relation as

$$\sum_m p(m) = 1$$

which implies

$$\sum_m \langle\psi| M_m^\dagger M_m |\psi\rangle = 1 \implies \sum_m M_m^\dagger M_m = I$$

Exercise 4.2. Define a set of Measurement operators that extract the coefficients along the computational basis.

This postulate fundamentally describes one of the major features of quantum mechanics, that measurement changes the system. After measuring the state vector collapses into another state.

We now talk about a different way of talking about measurements, the projective measurement operators. Suppose that the act of measurement is denoted by an operator M acting on our state space. Note that this can be done only when your system is closed, or when unitary dynamics can be applied to it (energy is conserved). The act of measurement may change this as external influence is applied.

However assuming energy is conserved then correspondingly M will be unitary and hence will have a spectral decomposition. Moreover as we will be measuring "real" values the eigenvalues will be real and thus M will be hermitian. Let m refer to the eigenvalues of this observable. Then

$$M = \sum_m m P_m$$

where P_m is the projector onto the eigenspace corresponding to the eigenvalue m . The act of observing the observable M will collapse the state vector into one of the eigenstates. Notice that this is similar to the previous postulates in the special case when our measurement operators are the projectors onto the eigenstates, as if $M_m = P_m$ then

$$P_m = M_m^\dagger M_m$$

Hence by the previous postulate the probability of getting a value m is

$$p(m) = \langle\psi| P_m |\psi\rangle$$

and the state after the measurement is

$$\frac{P_m |\psi\rangle}{\sqrt{p(m)}}$$

The average value of the measurement is then simply

$$E(m) = \sum_m m p(m) = \sum_m m \langle\psi| P_m |\psi\rangle = \langle\psi| M |\psi\rangle$$

However projective measurements are not all type of measurements. It can be shown that projective measurements along with unitary transformations can account for all measurement operators. We can however make the math corresponding to postulate 3 simpler by invoking a mathematical tool called POVM measurements.

4.4 POVM Measurement

Notice that in Projective measurement the measurement operators were the projectors themselves. As projectors satisfied $P^\dagger P = P$ the math corresponding to the probabilities becomes simple. Notice that the only place where we need M_m is to get the state of the system after the measurement. If we do not need or care about this, we can define a new set of operators

$$E_m = M_m^\dagger M_m$$

Then corresponding to E_m there is an eigenvalue m which could be obtained after the measurement. The probability simply becomes

$$p(m) = \langle \psi | E_m | \psi \rangle$$

and the completeness relation implies

$$\sum_m M_m^\dagger M_m = \sum_m E_m = I$$

The set of operators $\{E_m\}$ is termed as *POVM* and each E_m is called a *POVM* element. A POVM is sufficient to obtain the probabilities of each measurable value and makes it more elegant reducing the need for adjoints. We have shown that for every set of measurement operators there is a corresponding set of POVMs.

Exercise 4.3. Show that for every POVM there is a corresponding set of Measurement operators.

Exercise 4.4. Show that projective measurement operators are a special case of POVMs

Note that POVMs and general measurement operators are more general than projective measurements as we do not comment on the nature of the measurement operators. However it can be proven using composite quantum systems that projective measurements along with unitary operators are sufficient to describe any general measurement. We reserve this proof and discussion for a later time.

We briefly mention the final postulate which deals with composite quantum systems. We will revisit this later when we deal with tensor products and partial traces.

4.5 Composite Quantum Systems

Postulate 4. The state space of a composite physical system is the tensor product of the state spaces of the component physical systems. If the state of the i th component is $|\psi_i\rangle$ and total we have n components then the net state is given by

$$|\psi_1\rangle \otimes |\psi_2\rangle \otimes \cdots \otimes |\psi_n\rangle$$

The symbol \otimes is used for the tensor product. We will go through this later when we deal with composite quantum systems in the next section. Essentially tensor products are used to build composite systems out of individual hilbert spaces. Thus we can deal with quantum systems having multiple components like say 3 qubits simultaneously.

We now proceed onto simple quantum circuits. We will start with the most simple case, a quantum gate operating on a single qubit. After we analyse this we will look into tensor products and partial traces in detail which would allow us to analyse multiple qubit gates properly.

5 Single Qubit Gates

In this section we will deal with single qubit gates and different ways we can represent them.

A single qubit gate can be represented as an unitary operator acting on the state vector of the qubit. That is for every gate G there exists a unitary transformation U such that the final state of a state vector $|\psi\rangle$ is given by

$$|\psi_2\rangle = U|\psi\rangle$$

This is because the gate can be thought of as an abstract machine that takes in a state vector and operates on it with time. As all time transformations of quantum state vectors are unitary transformations the above result holds. Conversely we assume that for every unitary operator on a single qubit there exists a gate for the same. Achieving the construction of such a gate would venture into physical realisation of quantum computation which we reserve for later.

Exercise 5.1. Mathematically prove that all single gate qubit operators have to be unitary by using the normalisation condition of qubit state vectors.

We now introduce the bloch sphere notation for qubits. Observe that we can write any qubit as follows:

$$|\psi\rangle = e^{ix} \left(\cos \frac{\theta}{2} + ie^{i\phi} \sin \frac{\theta}{2} \right)$$

where

$$0 \leq x < 2\pi, 0 \leq \theta \leq \pi, 0 \leq \phi < 2\pi$$

The global phase e^{ix} is irrelevant as it is common to both. Observe that by neglecting this our state vector is parameterised by θ and ϕ and these values are precisely in the range of the spherical angular coordinates of a sphere of radius 1. Thus corresponding to every $|\psi\rangle$ (θ, ϕ) there exists a point on the unit sphere given by $(1, \theta, \phi)$ in spherical polar coordinates. This sphere is termed as the bloch sphere and the corresponding position vector the bloch vector.

It is interesting to note that any unitary transformation in the state space can be visualised to be a rotation of the bloch sphere about an axis. That is this mapping corresponds to a rotation of the bloch sphere (upto a global phase). You will prove this result later on. We first define some standard gates in their matrix forms. These gates are incredibly useful and are known as the pauli matrices.

$$X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad Y = \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} \quad Z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

First we would like to describe some standard transformations (or gates) that rotate the sphere about x, y, z axes.

$$\begin{aligned} R_x(\theta) &= e^{-i\theta \frac{X}{2}} = \begin{bmatrix} \cos \frac{\theta}{2} & -i \sin \frac{\theta}{2} \\ -i \sin \frac{\theta}{2} & \cos \frac{\theta}{2} \end{bmatrix} \\ R_y(\theta) &= e^{-i\theta \frac{Y}{2}} = \begin{bmatrix} \cos \frac{\theta}{2} & -\sin \frac{\theta}{2} \\ \sin \frac{\theta}{2} & \cos \frac{\theta}{2} \end{bmatrix} \\ R_z(\theta) &= e^{-i\theta \frac{Z}{2}} = \begin{bmatrix} e^{-i\frac{\theta}{2}} & 0 \\ 0 & e^{i\frac{\theta}{2}} \end{bmatrix} \end{aligned}$$

Exercise 5.2. Show that $X^2 = Y^2 = Z^2 = I$ and hence deduce the expansion of their exponentiation as done above using the result of exercise 2.5

It is a bit lengthy to show that the above matrices indeed do what we are claiming them to do. You can refer to this resource for more detail (you will need to go through density matrices that we introduce in the next section).

It is important to note that in the bloch sphere representation the coefficient along $|0\rangle$ is always a real positive number. After we apply any gate on the state vector this may not be true and so we must first divide by a common phase factor to ensure the above condition is met before we figure out the bloch vector. We prove this important theorem that relates unitary transformations to rotation transformations using the rotation matrices.

Theorem 5.1. Suppose U is a unitary transformation. Then there exists reals α, β, γ and δ such that

$$U = e^{i\alpha} R_z(\beta) R_y(\gamma) R_z(\delta)$$

Proof. Since U is unitary, the rows and columns of U are orthonormal. Thus there exist reals α, β, γ and δ such that

$$U = \begin{bmatrix} e^{i(\alpha - \frac{\beta}{2} - \frac{\delta}{2})} \cos \frac{\delta}{2} & -e^{i(\alpha - \frac{\beta}{2} + \frac{\delta}{2})} \sin \frac{\delta}{2} \\ e^{i(\alpha + \frac{\beta}{2} - \frac{\delta}{2})} \sin \frac{\delta}{2} & e^{i(\alpha + \frac{\beta}{2} + \frac{\delta}{2})} \cos \frac{\delta}{2} \end{bmatrix}$$

By expanding the right hand side of the statement to prove we immediately get the above and thus the given statement is proved. \square

The above theorem allows us to construct any gate using just three parameterised gates. There exists an important corollary of the above that we leave as an exercise. This corollary is useful when we want to construct multi qubit gates.

Exercise 5.3. Suppose U is a unitary gate. Show that there exists unitary operators A, B, C on a single qubit such that $ABC = I$ and U is identical to $AXBXC$ upto a global phase.

We end this section by mentioning two other important gates that we will later use. The Hadamard gate was previously introduced and is given by

$$H = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

. The T gate (also known as the $\frac{\pi}{8}$ gate) is given by

$$T = \exp(i\pi/8) \begin{bmatrix} \exp(-i\pi/8) & 0 \\ 0 & \exp(i\pi/8) \end{bmatrix}$$

We will revisit quantum gates in detail later. First let us look into composition of quantum systems and tensor products as we will require this when dealing with multi qubit gates.

6 Composite Quantum Systems

In this section we will deal with making composite quantum state spaces out of elementary state spaces. This will be important for instance, when we deal with multiple qubits. We begin with tensor products which serve as the mathematical backbone behind composite systems.

6.1 Tensor Products

Suppose \mathbf{A} and \mathbf{B} are two vector spaces that are n and m dimensional. The tensor product of \mathbf{A} and \mathbf{B} is denoted as

$$\mathbf{A} \otimes \mathbf{B}$$

This is a vector space that is nm dimensional whose basis elements are the tensor products of the basis elements of the individual vector spaces. An informal way to think of a tensor product is that it is just a placeholder where the individual vector spaces are kept along different dimensions. Suppose $|v\rangle \in \mathbf{A}$ and $|w\rangle \in \mathbf{B}$, then

$$|v\rangle \otimes |w\rangle \in \mathbf{A} \otimes \mathbf{B}$$

The tensor product is a vector space and along with that it must satisfy the following properties:

1. For scalar $c \in \mathbb{C}$ and $|v\rangle \in \mathbf{A}, |w\rangle \in \mathbf{B}$

$$c(|v\rangle \otimes |w\rangle) = c|v\rangle \otimes |w\rangle = |v\rangle \otimes c|w\rangle$$

2. For $|v_1\rangle, |v_2\rangle \in \mathbf{A}, |w\rangle \in \mathbf{B}$

$$(|v_1\rangle + |v_2\rangle) \otimes |w\rangle = |v_1\rangle \otimes |w\rangle + |v_2\rangle \otimes |w\rangle$$

3. For $|v\rangle \in \mathbf{A}, |w_1\rangle, |w_2\rangle \in \mathbf{B}$

$$|v\rangle \otimes (|w_1\rangle + |w_2\rangle) = |v\rangle \otimes |w_1\rangle + |v\rangle \otimes |w_2\rangle$$

Exercise 6.1. Using the above axioms prove that if $|i\rangle$ is a basis of \mathbf{A} and $|j\rangle$ is a basis of \mathbf{B} then $|i\rangle \otimes |j\rangle$ is a basis of $\mathbf{A} \otimes \mathbf{B}$

Similarly we can talk about operators on this tensor space. If M is a linear operator from $\mathbf{A} \rightarrow \mathbf{A}'$ and N is a linear operator from $\mathbf{B} \rightarrow \mathbf{B}'$ then $M \otimes N$ is a linear operator from $\mathbf{A} \otimes \mathbf{B} \rightarrow \mathbf{A}' \otimes \mathbf{B}'$ defined by

$$(M \otimes N)|i\rangle \otimes |j\rangle = M|i\rangle \otimes N|j\rangle$$

for $|i\rangle \otimes |j\rangle \in \mathbf{A} \otimes \mathbf{B}$. It is trivial to show using the above properties that the above operator is indeed linear. This is left as an exercise.

We can also define an inner product on this vector space. Let \mathbf{A} be an inner product space and \mathbf{B} be an inner product space. Then we define an inner product on $\mathbf{A} \otimes \mathbf{B}$ by

$$(|v_1\rangle \otimes |w_1\rangle, |v_2\rangle \otimes |w_2\rangle) = (|v_1\rangle, |v_2\rangle)(|w_1\rangle, |w_2\rangle)$$

for $|v_1\rangle, |v_2\rangle \in \mathbf{A}, |w_1\rangle, |w_2\rangle \in \mathbf{B}$. This way two sets of orthonormal bases created an orthonormal base of their tensor product space.

As an example suppose we have two qubits, then this composite system will have the bases $|0\rangle \otimes |0\rangle, |0\rangle \otimes |1\rangle, |1\rangle \otimes |0\rangle$ and $|1\rangle \otimes |1\rangle$. For convenience we can omit the middle symbol and write $|0\rangle \otimes |0\rangle$ as $|00\rangle$.

There is a useful way of writing the matrix representations of the operator formed by the tensor product of two operators using Kronecker Products. Suppose A is a linear operator whose matrix representation is

given by A_{ij} where $1 \leq i \leq n$ and $1 \leq j \leq m$. Similarly consider the matrix representation of another operator B that has p rows and q columns. It can be shown that the matrix form of $A \otimes B$ is given by:

$$A \otimes B = \begin{bmatrix} A_{11}B & A_{12}B & \cdots & A_{1m}B \\ A_{21}B & A_{22}B & \cdots & A_{2m}B \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1}B & A_{n2}B & \cdots & A_{nm}B \end{bmatrix}$$

where $A_{ij}B$ is a submatrix of the dimension of B formed by scaling up B with A_{ij} .

We end by mentioning that the notation $|\psi\rangle^{\otimes k}$ is written in shorthand for $|\psi\rangle \otimes |\psi\rangle \cdots \otimes |\psi\rangle$ with the tensor product being done k times.

Exercise 6.2. Given that $|\psi\rangle = (|0\rangle + |1\rangle)/\sqrt{2}$ find $|\psi\rangle^{\otimes 3}$ using the Kronecker product (Hint: consider the Hadamard gate)

6.2 Mixed Quantum States

We will now study quantum states that are mixed. Formally suppose you have a quantum system that can be in one of the states from the set $\{|\psi_i\rangle\}$ with probability $\{p_i\}$. Such a system is said to be in a mixed quantum state. It is important to understand that this is not the same as a superposition. Our system is in one of the states, we simply do not know which and it has a probability corresponding to that state. We then define the density operator corresponding to that state as

$$\rho = \sum_i p_i |\psi_i\rangle \langle \psi_i|$$

The above is useful because we can rewrite all of quantum mechanics in the form of density matrices only. For instance suppose we allow a unitary transformation U to take place on our system. We can show that the final density matrix will be given by

$$\rho' = U\rho U^\dagger$$

Suppose an ensemble of measurement operators $\{M_m\}$ are to be used to measure an observable on the mixed state ρ . We can show that the probability of obtaining a result m is given by

$$p(m) = \text{tr}(M_m^\dagger M_m \rho)$$

Exercise 6.3. Prove the above result. (Hint: use the measurement postulates, conditional probability, and the fact that $\text{tr}(A|\psi\rangle\langle\psi|) = \langle\psi|A|\psi\rangle$)

After we obtain a measurement result m it can be shown using conditional probability that the final mixed state of the system is,

$$\rho' = \frac{M_m \rho M_m^\dagger}{\text{tr}(M_m^\dagger M_m \rho)}$$

We must note that if a system has only one state in this ensemble (that is we are sure of which state it is in) then it is stated to be in a pure state. The density matrix corresponding to this state $|\psi\rangle$ is then simply

$$\rho = |\psi\rangle \langle \psi|$$

Density matrices for composite systems can be found in a similar manner by taking the tensor products of the individual density matrices. We also must note that given an ensemble of states there exists a unique density matrix but the reverse is not true. Given a density matrix there may be many ensembles that correspond to this density matrix. You can try to generate such a pair of states with the same density matrix for a simple density matrix. Below we state without proof the relation between two sets of ensembles such that they have the same density matrix.

Theorem 6.1. Two set of ensembles $\{|\psi\rangle\}$ and $\{|\phi\rangle\}$ have the same density matrix if there exists a unitary matrix of complex numbers U such that

$$|\psi_i\rangle = \sum_j u_{ij} |\phi_j\rangle$$

where add extra 0s to the smaller ensemble so that both the ensembles have the same number of elements.

Exercise 6.4. Prove that the trace of the density matrix is one and that the density matrix is a positive operator

Exercise 6.5. Prove that for any density matrix ρ satisfies $\text{tr}(\rho^2) \leq 1$ with equality if and only if ρ corresponds to a pure state.

Solution. Let the ensemble corresponding to ρ be $|\psi_i\rangle$ with probabilities p_i . Then

$$\rho^2 = \sum_{ij} p_i p_j |\psi_i\rangle \langle \psi_i | \psi_j\rangle \langle \psi_j| = \sum_i p_i^2 |\psi_i\rangle \langle \psi_i|$$

Clearly the trace of this is $\sum_i p_i^2$ which is ≤ 1 by the cauchy schwartz inequality with equality if and only if ρ corresponds to a pure state.

It is of interest to us to find the density matrix corresponding to a subsystem of a quantum system. This is done using partial traces and we will go over it in brief.

6.3 Reduced Density Matrices

Suppose the composite system composed of two components A, B has a density matrix given by ρ_{AB} . The density matrix of a component A is given by

$$\rho_A = \text{tr}_B(\rho_{AB})$$

where tr_B is the partial trace, where the trace operator is applied only on the second component. For instance when we expand the sum in ρ_{AB} we get terms of the form

$$|\psi_a\rangle \langle \psi_a| \otimes |\psi_b\rangle \langle \psi_b|$$

The partial trace of this will be

$$\text{tr}_B(|\psi_a\rangle \langle \psi_a| \otimes |\psi_b\rangle \langle \psi_b|) = |\psi_a\rangle \langle \psi_a| \text{tr}(|\psi_b\rangle \langle \psi_b|) = |\psi_a\rangle \langle \psi_a|$$

We can see why this works as suppose the individual systems had density matrices ρ_A and ρ_B then

$$\rho_{AB} = \rho_A \otimes \rho_B$$

which means that

$$\text{tr}_B(\rho_{AB}) = \rho_A \text{tr}(\rho_B) = \rho_A$$

which has thus extracted the individual density matrix out.

It is important to note that the density matrix alone is sufficient to determine not only the measurement statistics but also the state of the system with time. Density matrices thus provide an alternative to dealing with state vectors in quantum mechanics.

At this stage we have covered most of the basic tools needed to study composite systems. Partial traces and the results of measurement operators on density matrices are useful when we deal with measurements on composite systems. We will now look at applications of the last few sections by studying how arbitrary quantum gates operating on multiple qubits can be formed from a few universal gates.

With this we have mostly ended with the prerequisites for studying Quantum Computation and we will be using the toolset that we have developed so far for various applications of quantum mechanics in computation. We will begin with a relatively simpler application, super dense coding.

7 Superdense Coding

The main question that we seek to answer in this section is that can we transmit more than a single classical bit on information using a single qubit? Intuitively it may seem so and this is infact true as we will see. However simply transferring a single qubit does not work because measurement collapses the qubit. Instead we will be using the properties of the four possible EPR pairs which we describe below.

The EPR pairs or the bell states refer to the four following states comprising of two qubits that are entangled with each other.

$$\frac{|00\rangle + |11\rangle}{\sqrt{2}}, \frac{|00\rangle - |11\rangle}{\sqrt{2}}, \frac{|10\rangle + |01\rangle}{\sqrt{2}}, \frac{|01\rangle - |10\rangle}{\sqrt{2}}$$

The main idea is that we will use the entanglement of the two qubits in one of these states to transfer two bits of information by simply transferring one bit. Suppose Alice wants to send Bob two classical bits of information. To do so using superdense coding, an external party will first prepare the state

$$|\psi\rangle = \frac{|00\rangle + |11\rangle}{\sqrt{2}}$$

and he will then give the first qubit to Alice and the second to Bob. If Alice wants to send the information 00 she will do nothing and pass her qubit to Bob. If she wants to send 01 then she applies the phase flip gate Z , if she wants to send 10 she applies the quantum NOT gate X to her qubit, if she wants to send 11 she applies the iY gate. The final state of the entangled qubits corresponding to the information she wants to send is:

$$00 : |\psi\rangle = \frac{|00\rangle + |11\rangle}{\sqrt{2}}$$

$$01 : |\psi\rangle = \frac{|00\rangle - |11\rangle}{\sqrt{2}}$$

$$10 : |\psi\rangle = \frac{|10\rangle + |01\rangle}{\sqrt{2}}$$

$$11 : \frac{|01\rangle - |10\rangle}{\sqrt{2}}$$

Notice that all four of these states form a basis and thus when Alice passes her qubit to Bob then Bob can find out which state he got by simply measuring with respect to this basis to figure out which basis vector he had received.

Exercise 7.1. Specify which measurement operators Bob should use to figure out the two classical bits that Alice wanted to send him

Notice that this process did involve two qubits but Alice had to send only one qubit. The entanglement of the two qubits allowed Bob to figure out two classical bits of information on transfer of a single qubit.

Let's now look at the reverse, can we transmit a qubit by sending only classical information?

8 Quantum Teleportation

This time we want Alice to send two classical bits of information to "send" a qubit. This seems a bit complicated as a qubit is determined by one of it's coefficients along a computational basis. This after taking out a global phase factor can be a real number and so may have a lot of digits in it's expansion and it seems counter intuitive that two classical bits can transmit this information.

However we again leverage the power of entanglement and the EPR pairs. We first let an external agent prepare the first bell state given by

$$|\beta_{00}\rangle = \frac{|00\rangle + |11\rangle}{\sqrt{2}}$$

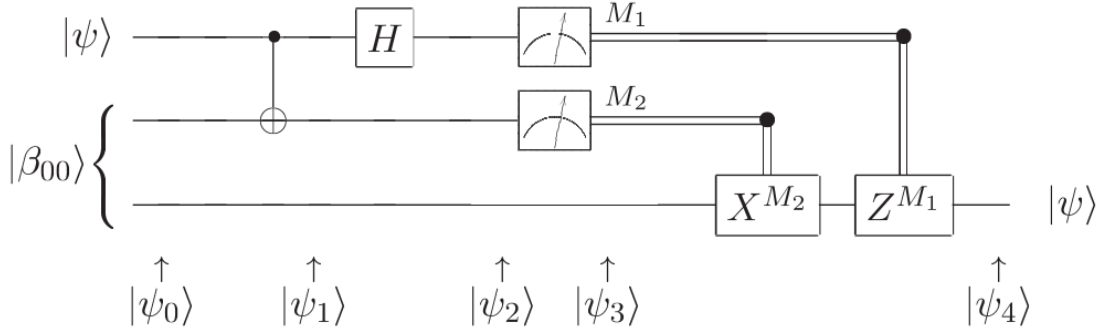
We again give the first qubit to Alice and the second to Bob. Now suppose Alice wants to send or teleport the following qubit to Bob.

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

The composite quantum state which we have considering $|\psi\rangle$ will be

$$|\psi_0\rangle = |\psi\rangle |\beta_{00}\rangle$$

Alice can use the following circuit to convert the composite system of the bell state and this qubit to convert the state into a form that can be used for teleportation.



Let's go through this circuit one by one. Observe that Alice only has the first two qubits. First a controlled CNOT operation with the qubit corresponding to the state we want to transfer as the control bit is applied. This results in the state

$$|\psi_1\rangle = \frac{\alpha|0\rangle(|00\rangle + |11\rangle) + \beta|1\rangle(|10\rangle + |01\rangle)}{\sqrt{2}}$$

Then a Hadamard gate on the first qubit is applied. This results in the net state being (verify that this is indeed the third state)

$$|\psi_3\rangle = \frac{|00\rangle(\alpha|0\rangle + \beta|1\rangle) + |01\rangle(\alpha|1\rangle + \beta|0\rangle) + |10\rangle(\alpha|0\rangle - \beta|1\rangle) + |11\rangle(\alpha|1\rangle - \beta|0\rangle)}{2}$$

Now this state is rather useful for us. Observe that if we measure the first two qubits, depending upon which state the first two qubits collapse into the third qubit will be one of the following states,

$$|00\rangle : \alpha|0\rangle + \beta|1\rangle$$

$$|01\rangle : \alpha|1\rangle + \beta|0\rangle$$

$$|10\rangle : \alpha|0\rangle - \beta|1\rangle$$

$$|11\rangle : \alpha|1\rangle - \beta|0\rangle$$

Thus if Alice measures the first two qubits that she possesses, depending on her result, the third qubit with Bob will change into one of the four states. If Alice passes her measurement result information (which is given in the form of two classical bits) to Bob, Bob can fix his state to obtain the state $|\psi\rangle$ back. Verify from the diagram that applying that the operations X and Z do indeed result in the original state reverting. For instance if the result was 11 then Bob will apply both the X and Z gates while if it was 01 then only the Z gate is applied. Thus using the EPR pair we can achieve quantum teleportation.

An important point to note here is that Quantum Teleportation does not violate the quantum no cloning theorem. Observe that we did not copy the original state to be transferred as the original state transformed into a different state losing its initial properties. In a sense we have transferred the information, not copied it.

9 Deutsch-Jozsa Algorithm

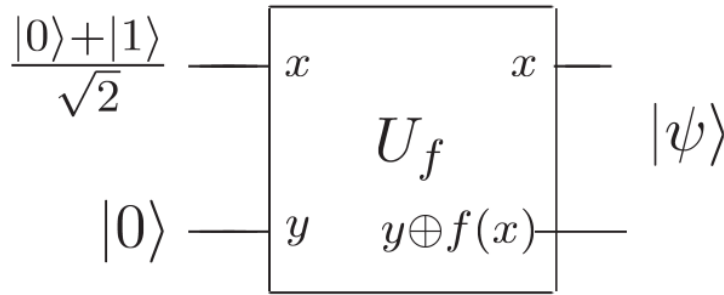
9.1 Motivation

One interesting property of quantum circuits that we want to leverage is parallelism. This follows directly from the principle of superposition. To make it a bit more clear let us consider a black box that allows us to compute a function $f : \{0, 1\}^n \rightarrow \{0, 1\}$. We however need a place to store the result. We thus want to have a function that achieves the following operation for us,

$$|x, y\rangle \rightarrow |x, f(x) \oplus y\rangle$$

This can be achieved with a unitary operator U_f . The process for constructing such an oracle would be to take the classical circuit for computing $f(x)$ and use quantum equivalents of the gates used in this circuit. Leaving aside this implementational detail we observe that if we pass in a superposition of states in x in principle we can obtain multiple values of the function in a single pass. For instance consider the simple case of $n = 1$ and suppose that

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



Then the final state $|\psi\rangle$ would then be simply

$$|\psi\rangle = \frac{|0, f(0)\rangle + |1, f(1)\rangle}{\sqrt{2}}$$

This means that we have obtained two values of the function $f(x)$ by evaluating our quantum circuit just once. But unfortunately this state is a superposition. We cannot obtain both the values as measurement once collapses this state. What we therefore desire is to use this power of superposition to obtain non trivial values that would normally require us to compute the classical circuit multiple times. The Deutsch Algorithm allows to find $f(0) \oplus f(1)$ by just evaluating the circuit once.

9.2 Deutsch Algorithm

In order to obtain $f(0) \oplus f(1)$ we will use the oracle after applying the hadamard gate on the input qubits $|0\rangle$ and $|1\rangle$. This gives us a form that becomes useful in determining $f(0) \oplus f(1)$ as we will see.

Refer to the figure on the next page to see the circuit we will be implementing. The state $|\psi_1\rangle$ is given by

$$|\psi_1\rangle = \frac{|00\rangle - |01\rangle + |10\rangle - |11\rangle}{2} = \left[\frac{|0\rangle + |1\rangle}{\sqrt{2}} \right] \left[\frac{0 - |1\rangle}{\sqrt{2}} \right]$$

The state $|\psi_2\rangle$ after applying the oracle can be written in the following form

$$|\psi_2\rangle = \pm \left[\frac{|0\rangle + |1\rangle}{\sqrt{2}} \right] \left[\frac{|0\rangle - |1\rangle}{\sqrt{2}} \right]$$

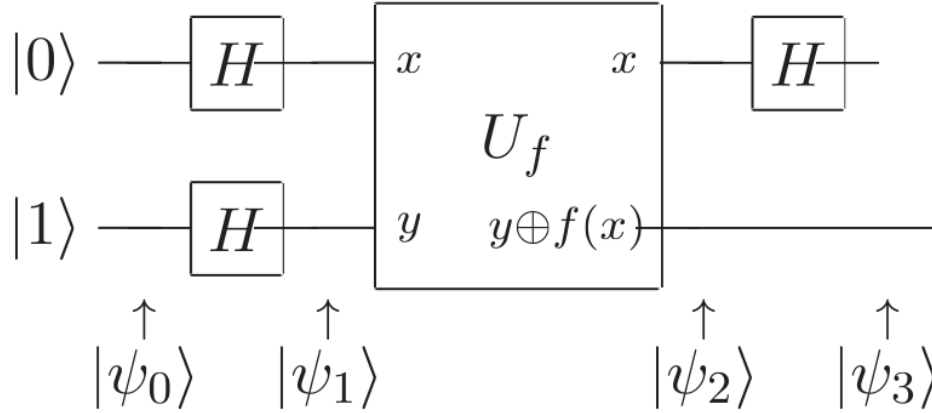
if $f(0) = f(1)$ otherwise

$$|\psi_2\rangle = \pm \left[\frac{|0\rangle - |1\rangle}{\sqrt{2}} \right] \left[\frac{|0\rangle - |1\rangle}{\sqrt{2}} \right]$$

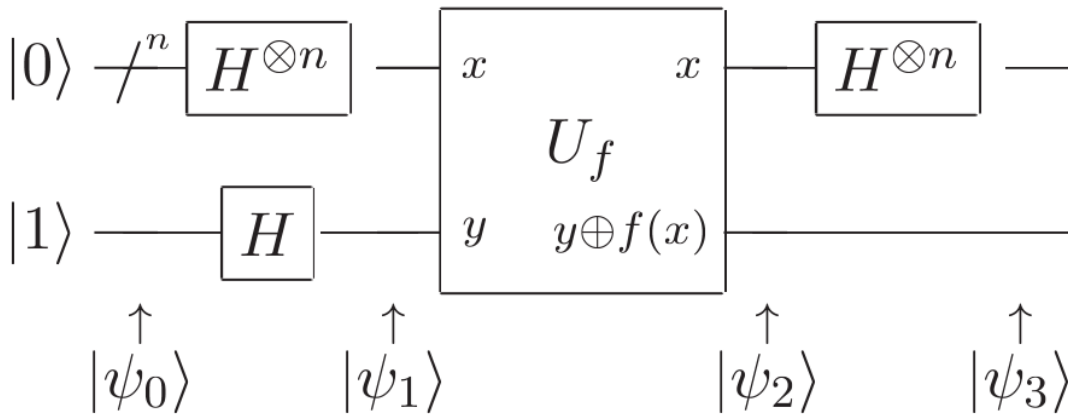
Observe that by applying the hadamard on the first qubit and observing that $|f(0) \oplus f(1)\rangle = |0\rangle$ if $f(0) = f(1)$ and $|1\rangle$ otherwise gives the state $|\psi_3\rangle$ as

$$|\psi_3\rangle = |f(0) \oplus f(1)\rangle \left[\frac{|0\rangle - |1\rangle}{\sqrt{2}} \right]$$

Now simply measuring the first qubit allows us to find $f(0) \oplus f(1)$ in one single evaluation of the quantum circuit. This is Deutsch's algorithm.



9.3 The Deutsch-Jozsa Algorithm



Deutsch Algorithm is a special case of the more specific Deutsch-Jozsa Algorithm. Consider the following problem which can be formulated as a game. Alice, in Amsterdam, selects a number x from 0 to $2^n - 1$, and mails it in a letter to Bob, in Boston. Bob calculates some function $f(x)$ and replies with the result, which is either 0 or 1 . Now, Bob has promised to use a function f which is of one of two kinds; either $f(x)$ is constant for all values of x , or else $f(x)$ is balanced, that is, equal to 1 for exactly half of all the possible x , and 0 for

the other half. Alice's goal is to determine with certainty whether Bob has chosen a constant or a balanced function, corresponding with him as little as possible. Classically clearly we need atleast $2^{n-1} + 1$ classical bits to resolve this. Can we do better using quantum algorithms?

It turns out we need to evaluate the oracle only once to solve this. Notice that the previous algorithm solved exactly this for the case $n = 1$. Our modified oracle takes in n qubits and outputs the value of $f(x)$ in the target register which we set to $|1\rangle$ initially. The circuit is shown in the previous page.

Observe that we can write $|\psi_1\rangle$ as

$$|\psi_1\rangle = \frac{\sum_{x \in \{0,1\}^n} |x\rangle}{\sqrt{2^n}} \left[\frac{|0\rangle - |1\rangle}{\sqrt{2}} \right]$$

$|\psi_2\rangle$ then becomes

$$|\psi_2\rangle = \frac{\sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle}{\sqrt{2^n}} \left[\frac{|0\rangle - |1\rangle}{\sqrt{2}} \right]$$

Finally we apply the n gate hadamard on the first register of n qubits. These qubits are already in superposition so it seems a bit complex to apply it on all of them. We can however write it as a double sum.

Exercise 9.1. Prove that

$$H^{\otimes n} |x_1, x_2 \dots x_n\rangle = \frac{\sum_{z_1, z_2, \dots, z_n} (-1)^{x_1 z_1 + x_2 z_2 + \dots + x_n z_n} |z_1, z_2 \dots z_n\rangle}{\sqrt{2^n}}$$

The above can be written simply in the form

$$H^{\otimes n} |x\rangle = \frac{\sum_z (-1)^{x \cdot z} |z\rangle}{\sqrt{2^n}}$$

where $x \cdot z$ is the bitwise product modulo 2.

Thus we can write $|\psi_3\rangle$ as

$$|\psi_3\rangle = \sum_z \sum_x \frac{\sum_z (-1)^{x \cdot z + f(x)} |z\rangle}{\sqrt{2^n}} \left[\frac{|0\rangle - |1\rangle}{\sqrt{2}} \right]$$

Now suppose $f(x)$ is constant. Then we can see that the coefficient of $|0\rangle^{\otimes n}$ is either $+1$ or -1 depending on $f(x)$. But this means that if we measure the first n qubits in the register we will obtain $|0\rangle^{\otimes n}$ with certainty as our state vector is of unit length and so the other terms have amplitude 0. Conversely if $f(x)$ is balanced the coefficient of $|0\rangle^{\otimes n}$ is 0. Thus if we measure the qubits in the first register and if we obtain $|0\rangle^{\otimes n}$ we see that $f(x)$ is constant and otherwise it is balanced.

Thus using only one pass through the oracle we can solve this problem. This is far far better than the classical case where we need to evaluate the function at $2^{n-1} + 1$ values. This sums up the Deutsch-Jozsa Algorithm.