

Introduction to Quantum Information Theory

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Abstract

Will finish this when i actually know what the jeebuz is going on in the class.

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1 Vectors and Vector Spaces

1.1 Vector Spaces

In modern mathematics, an area of importance is first defining what field the problems are being asked in. When speaking on vectors and vector spaces, it is easiest to work in two main fields of numbers.

1.1.1 Relevant Vector Spaces

1. \mathbb{R} : All Real Numbers $(-\infty, \infty)$

This includes all real numbers from zero to infinity, something of note is that this number field is **not algebraically closed**, meaning not all numbers can be produced from combinations of other numbers in the field. An example of this is $x^2 + 1 = 0$, meaning $x = 1 * i$

2. \mathbb{C} : All Complex Numbers

This number field is **algebraically closed**. This means that every possible number in the field can be produced from combinations of other numbers in the field. This makes sense as two complex numbers multiplied together can produce 1. A completely real number 2. A complex number with a real and imaginary component 3. A complex number with only an imaginary component.

1.2 Complex Number Operations

In the following sections, we will make the assumption $a, b, c, d, e \in \mathbb{R}$

1.2.1 Addition

$$(a + bi) + (c + di) = (a + c) + (b + d)i \quad (1.1)$$

1.2.2 Multiplication

$$(a + bi) * (c + di) = (ac + adi) + (bci - bd) \quad (1.2)$$

1.2.3 Conjugation

$$\overline{a + bi} = (a + bi)^* = a - bi \quad (1.3)$$

1.2.4 Modulus

$$|z| = |a + bi| = \sqrt{a^2 + b^2} \quad (1.4)$$

1.3 Common Vector Space Notation and Definitions

This section is to lay out some of the most important pieces of notation and definitions for the complex mathematics required for this course.

1.3.1 Linear Algebra Terms

Definition 1.1 An element $|v\rangle \in \mathbb{C}^n$ is called a **(ket) vector** and is expressed as a **column** of n complex numbers. The integer n is called the **dimension** of the vector space \mathbb{C}^n . ♦

Definition 1.2 A **linear combination** of $|v_1\rangle, \dots, |v_k\rangle \subset \mathbb{C}^n$ is just a single vector in the form $\lambda_1 |v_1\rangle + \lambda_2 |v_2\rangle + \dots + \lambda_k |v_k\rangle$ for some $\lambda_1, \lambda_2, \dots, \lambda_k \in \mathbb{C}$. ♦

Example 1.3 Create a **linear combination** of $|v_1\rangle = \begin{bmatrix} i \\ 1 \end{bmatrix}$, $|v_2\rangle = \begin{bmatrix} -1 \\ i+1 \end{bmatrix}$

To create a linear combination, we simply must add two complex number constants λ_1, λ_2 in the form shown in **Definition 1.2**, $\lambda_1 |v_1\rangle + \lambda_2 |v_2\rangle$.

$$\begin{aligned} \lambda_{final} |v_{final}\rangle &= \lambda_1 |v_1\rangle + \lambda_2 |v_2\rangle & \lambda_1 &= 1 \\ & & \lambda_2 &= 3 + 2i \end{aligned}$$

$$\begin{aligned} \lambda_{final} |v_{final}\rangle &= 1 \begin{bmatrix} i \\ 1 \end{bmatrix} + (3 + 2i) \begin{bmatrix} -1 \\ i+1 \end{bmatrix} \\ &= \begin{bmatrix} i \\ 1 \end{bmatrix} + \begin{bmatrix} -3 - 2i \\ 1 + 5i \end{bmatrix} & (1.5) \\ &= \begin{bmatrix} -3 - i \\ 3 + 5i \end{bmatrix} \end{aligned} \quad \blacklozenge$$

Definition 1.4 Two vectors $|v_1\rangle, |v_2\rangle$ are **linearly independent** if the only way to create the 0 vector from a linear combination is through setting all complex constants $\lambda_1, \lambda_2, \dots, \lambda_k$ to zero in the linear combination formula described in **Definition 1.2**. ♦

Definition 1.5 A **subspace** of complex plane \mathbb{C} is a set $\mathbf{M} \in \mathbb{C}^n$ that satisfies three properties:

- $|0\rangle \in \mathbf{M}$
- $|v\rangle + |u\rangle \in \mathbf{M}$ for all $|v\rangle, |u\rangle$
- $c|v\rangle \in \mathbf{M} \quad \forall \quad |v\rangle \in \mathbf{M}, c \in \mathbb{C}$

♦

Definition 1.6 The **span** of a set \mathbf{S} is the smallest linear subspace $\in \mathbf{S}$. ♦

Theorem 1.7 (*Span of a set is always a subspace theorem*)

Let $\mathbf{s} = \{|v_1\rangle, |v_2\rangle, \dots, |v_k\rangle\} \subset \mathbb{C}^n$, then $\text{span}(\mathbf{s})$ is a subspace of \mathbb{C}^n

Proof. 1. Note that $|0\rangle = 0|v_1\rangle + 0|v_2\rangle + \dots + 0|v_k\rangle \in \text{span}(\mathbf{s})$.

2. Let $|v\rangle, |u\rangle \in \text{span}(\mathbf{s})$ be arbitrary.

Since $|v\rangle, |u\rangle \in \text{span}(\mathbf{s})$, there exists scalars c_1, c_2, \dots, c_k and d_1, d_2, \dots, d_k such that

$$\begin{aligned} |v\rangle &= c_1 |v_1\rangle, \dots, c_k |v_k\rangle \\ |u\rangle &= d_1 |v_1\rangle, \dots, d_k |v_k\rangle \end{aligned}$$

then,

$$\begin{aligned} |v\rangle + |u\rangle &= (c_1 |v_1\rangle, \dots, c_k |v_k\rangle) + (d_1 |v_1\rangle, \dots, d_k |v_k\rangle) \\ &= (c_1 + d_1) |v_1\rangle + \dots + (c_k + d_k) |v_k\rangle \in \text{span}(\mathbf{s}) \end{aligned}$$

This is true because all of these are linear combinations consisting of a complex constant and a ket vector, which coincides with the **Linear Combination Definition 1.2**.

3. Let $|w\rangle \in \text{span}(\mathbf{s})$ and $c \in \mathbb{C}$. If this is true, then there exists $\lambda_1, \lambda_2, \dots, \lambda_k$ such that $|w\rangle = \lambda_1 |v_1\rangle, \dots, \lambda_k |v_k\rangle$. Consider:

$$\begin{aligned} c|w\rangle &= c|w\rangle \\ &= C(\lambda_1 |v_1\rangle, \dots, \lambda_k |v_k\rangle) \\ &= (C\lambda_1) |v_1\rangle + \dots + (C\lambda_k) |v_k\rangle \end{aligned} \tag{1.6}$$

Equation 1.6 shows that after distributing **C** in, we have another linear combination, which must exist inside of $\text{span}(\mathbf{s})$, and since $\text{span}(\mathbf{s})$ satisfies 1,2,3, $\text{span}(\mathbf{s})$ must be a subspace. ■

Theorem 1.8 *Basis Theorem*

If \mathbf{S} is a set of n linearly independent vectors in \mathbb{C}^n , then $\text{span}(\mathbf{S}) = \mathbb{C}^n$.

Definition 1.9 A **basis** of a subspace $M \subseteq \mathbb{C}^n$ is a set of vectors \mathbf{S} such that

1. \mathbf{S} is linearly independent.

2. $\text{span}(\mathbf{S}) = M$, or in plain English "S spans M" ◆

Definition 1.10 The **dimension** of a space is the number of vectors in it. ◆

Example 1.11 Standard basis of different complex vector spaces.

- Standard Basis for \mathbb{C}^2 : $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$
- Standard Basis for \mathbb{C}^n : $\begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$

◆

Exercise 1.12 (Vector Space Exercise) Below are some exercises designed to put the concepts in this section to work, in order to further reinforce learning.

1. Consider $S = \{|v_1\rangle, |v_2\rangle, |v_3\rangle\}$, where $|v_1\rangle = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, |v_2\rangle = \begin{bmatrix} i \\ 1 \end{bmatrix}, |v_3\rangle = \begin{bmatrix} 0 \\ i \end{bmatrix}$

- (a) Give a linear combination of the vectors in S .

$$\lambda |v\rangle = 1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + 1 \begin{bmatrix} i \\ 1 \end{bmatrix} - (1 + 200i) \begin{bmatrix} 0 \\ i \end{bmatrix} \quad (1.7)$$

- (b) Determine if $\begin{bmatrix} 1+i \\ 200-i \end{bmatrix}$ is in $\text{Span}(S)$.

$$\alpha |v_1\rangle + \beta |v_2\rangle + \gamma |v_3\rangle = \begin{bmatrix} 1+i \\ 200-i \end{bmatrix}, \quad \alpha = 1, \beta = 1, \gamma = ?$$

$$1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + 1 \begin{bmatrix} i \\ 1 \end{bmatrix} - \gamma \begin{bmatrix} 0 \\ i \end{bmatrix} = \begin{bmatrix} 1+i \\ 200-i \end{bmatrix} \quad \text{Choose alpha and beta to obtain } \begin{bmatrix} 1+i \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1+i \\ 0 \end{bmatrix} - \gamma \begin{bmatrix} 0 \\ i \end{bmatrix} = \begin{bmatrix} 1+i \\ 200-i \end{bmatrix} \quad \text{Now we must isolate } \gamma$$

$$\begin{bmatrix} 0 \\ \gamma * i \end{bmatrix} = \begin{bmatrix} 0 \\ -200+i \end{bmatrix} \quad \text{We now set the matrix as an equality}$$

$$\gamma * i = -200 + i$$

$$\gamma = -(1 + 200i)$$

$$1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + 1 \begin{bmatrix} i \\ 1 \end{bmatrix} - (1 + 200i) \begin{bmatrix} 0 \\ i \end{bmatrix} = \begin{bmatrix} 1+i \\ 200-i \end{bmatrix}$$

Since we were able to create a linear combination of the vectors that satisfy the given vector, it is in $\text{Span}(S)$.

- (c) Describe $\text{Span}(S)$ geometrically.

$\text{Span}(S)$ is every possible vector $\in \mathbb{C}^2$

◆

Exercise 1.13 (Linear Independence Exercise) Find the condition under which the following two vectors are linearly independent.

$$|v_1\rangle = \begin{bmatrix} x \\ y \\ 3 \end{bmatrix} \in \mathbb{R}, \quad |v_2\rangle = \begin{bmatrix} 2 \\ x-y \\ 1 \end{bmatrix} \in \mathbb{R}$$

$$\begin{bmatrix} x \\ y \\ 3 \end{bmatrix} = \alpha \begin{bmatrix} 2 \\ x-y \\ 1 \end{bmatrix} \tag{1.8}$$

First, we must make this linearly dependent so we know when x and y fail linear independence

$$\alpha = 3 \quad \rightarrow \quad \begin{bmatrix} x \\ y \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 2 \\ x-y \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \\ 3 \end{bmatrix} = \begin{bmatrix} 6 \\ 3(x-y) \\ 3 \end{bmatrix}$$

Now we need to make this an equality to solve for x and y.

$$x = 6 \quad \rightarrow \quad y = 3(x - y)$$

$$y = 3x - 3y$$

$$4y = 18$$

$$y = 9/2 \text{ (or 4.5)}$$

◆

Exercise 1.14 (Basis Exercise) Show that the set formed by the following vectors is a basis for C^3

$$|V_1\rangle = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, |V_2\rangle = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, |V_3\rangle = \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}$$

There are two things we must prove in order for a set of vectors to form a basis: linear independence and an equal number of vectors as the dimension of the space.

First we will prove linear independence:

1. Linear Independence Test

$$\begin{aligned} & \begin{bmatrix} 1 & 1 & 1 & 0 \\ 1 & 0 & -1 & 0 \\ 1 & 1 & -1 & 0 \end{bmatrix} \text{ First, we build augmented matrix} \\ \xrightarrow{R_2-R_3} & \begin{bmatrix} 1 & 1 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 1 & -1 & 0 \end{bmatrix} \\ \xrightarrow{R_3-R_1} & \begin{bmatrix} 1 & 1 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -2 & 0 \end{bmatrix} \\ \xrightarrow{-1*R_2, -\frac{1}{2}*R_3} & \begin{bmatrix} 1 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \\ \xrightarrow{\begin{matrix} R_1-R_2 \\ R_1-R_3 \end{matrix}} & \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \end{aligned} \tag{1.9}$$

Thus, we have proved that if the only way to obtain the $[0]$ matrix is through $\lambda_1, \dots, \lambda_k = 0$ as mentioned in [Definition 1.4](#), we know that the vectors are **linearly independent**.

2. Dimension Test

Now we must prove that the dimension of the space is equal to the number of the vectors, and since we are working with C^3 and we have $|V_1\rangle, |V_2\rangle, |V_3\rangle$ we know that our space has three dimensions and we have three vectors, therefore we have just proven that **the given set of vectors forms a basis for C^3** . \blacklozenge

1.4 Inner/Outer Products, and Norms of Vectors

Definition 1.15 The **dual space of \mathbb{C}^n** (\mathbb{C}^{n*}) is a space of row vectors where there are n entries from \mathbb{C} . ♦

Definition 1.16 A **transpose** is an operation that every column into a row vector and vice versa without changing the order. ♦

Example 1.17 (Example of transpose operation)

$$|V\rangle^\dagger = \begin{bmatrix} 7 \\ 8i \\ \pi + 3i \\ 0 \end{bmatrix}^\dagger \quad (1.10)$$

$$\langle V| = [7 \quad 8i \quad \pi + 3i \quad 0] \quad (1.11)$$

It can be seen that the only difference between **1.10** and **1.11** is that we have converted the column to a row, this is accomplished through the **dagger** (\dagger) being applied to a matrix or vector. ♦

Definition 1.18 Given $|V\rangle, |W\rangle$, the **inner product** of $|V\rangle, |W\rangle$ is $\langle W|V\rangle$ which can also be written as $\langle W| * |V\rangle$.

It should be noted that the inner product of a bra and ket vector will always yield a constant. ♦

Example 1.19 (Example of an Inner Product on \mathbb{R}^n)

$$|V\rangle = \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}, |W\rangle = \begin{bmatrix} 5 \\ -2 \\ 1 \end{bmatrix} \quad (1.12)$$

$$\langle W|V\rangle = [5 \quad -2 \quad 1] \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix} = 5 - 2 + 3 = 6 \quad (1.13)$$

It should be noted that the inner product on \mathbb{R}^n is just the usual dot product. ♦

Example 1.20 (Example of an Inner Product on \mathbb{C}^n)

$$|V\rangle = \begin{bmatrix} i \\ 1 \end{bmatrix}, |W\rangle = \begin{bmatrix} -i \\ -1 \end{bmatrix} \quad (1.14)$$

$$\langle W|V\rangle = [-i \quad -1] \begin{bmatrix} i \\ 1 \end{bmatrix} = 1 - 1 = 0 \quad (1.15)$$

*The inner product yielding 0 means that the two vectors are **orthogonal**.* ♦

Definition 1.21 The norm (or magnitude) of $|V\rangle \in \mathbb{C}^n$ is

$$\| |V\rangle \| := \sqrt{\langle V|V \rangle} \quad (1.16)$$

◆

Example 1.22 (Example of taking the norm of a complex vector)

$$v = [1, 1] \in \mathbb{R}^2$$

$$\| |V\rangle \| := \sqrt{\begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix}} = \sqrt{1+1} = \sqrt{2}$$

Two points of note:

1. Can $\langle W|V \rangle$ be complex? **YES**
2. Can $\langle V|V \rangle = \| |V\rangle \|^2 \in \mathbb{C}$? **YES**

◆

Definition 1.23 A set of nonzero vectors $S = \{|V_1\rangle, |V_2\rangle, \dots, |V_k\rangle\} \subseteq \mathbb{C}^n$ is called an **orthogonal set** if $\langle v_i | v_j \rangle = 0$ if $i \neq j$.

◆

Theorem 1.24 (Orthogonality and Linear Independence Theorem) *In this theorem, we will use what we have learned so far in order to gain a deeper understanding into the properties of orthogonality in a vector space.*

If $S = |V_1\rangle, \dots, |V_k\rangle$ is an orthogonal vector set of nonzero vectors in \mathbb{C}^n , then S is linearly independent.

Proof. Let $c_1, \dots, c_k \in \mathbb{C}$ s.t. $c_1 |v_1\rangle, c_2 |v_2\rangle, \dots, c_k |v_k\rangle = |0\rangle$.

Goal: Show $c_1, \dots, c_k = 0$ using assumption S is orthogonal set.

Let $j \in \{1, \dots, k\}$, then

$$\begin{aligned} \langle v_j | (c_1 |v_1\rangle, \dots, c_k |v_k\rangle) &= \langle v_j | 0 \rangle \\ c_1 \langle v_j | v_1 \rangle + \dots + c_k \langle v_j | v_k \rangle & \\ c_j \langle v_j | v_j \rangle &= \text{Since } S \text{ is an orthogonal set} \end{aligned} \quad (1.17)$$

Since $\langle v_j | v_j \rangle$ can't possibly be 0 because we know that $j \neq 0$, it means that c_k must be zero, which as stated in the **Linear Independence Definition**, means that S is linearly independent. ■

Definition 1.25 A basis is **orthonormal** if two conditions are satisfied:

1. S is an orthogonal set of n unit vectors.
2. $\text{Span}(S) \in \mathbb{C}^n$

In plain English, an orthonormal basis is a basis in which there are exactly the number of vectors as the dimension of the space (from the basis definition) as well as all of those vectors being orthogonal to each other.

◆

Exercise 1.26 (Basis Practice) In this example, there will be multiple transformations completed to gain a better understanding of what a basis is, and what it means to be orthogonal.

1.

$$|b_1\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad |b_2\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad (1.18)$$

(a) Show that \mathcal{B} is an orthonormal basis for \mathbb{C}^2 .

i. Check orthogonality:

$$\langle b_1 | b_2 \rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 1 - 1 = 0$$

These two vectors are orthogonal since their inner product is 0.

ii. Check that they are unit vectors: (is the norm one?)

$$\begin{aligned} \| |b_1\rangle \| &= \sqrt{\langle b_1 | b_1 \rangle} = \frac{1}{\sqrt{2}} \sqrt{1 + 1} = \frac{\sqrt{2}}{\sqrt{2}} = 1 \\ \| |b_2\rangle \| &= \sqrt{\langle b_2 | b_2 \rangle} = \frac{1}{\sqrt{2}} \sqrt{1^2 + (-1)^2} = \frac{\sqrt{2}}{\sqrt{2}} = 1 \end{aligned}$$

This means that, since the vectors are orthogonal and unit vectors, they are orthonormal, forming an orthonormal basis for \mathbb{C}^2 .

- (b) Find the coordinates of $|x\rangle = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$ relative to basis \mathcal{B} (i.e. find the scalars such that...)

$$c_1, c_2 \in \mathbb{C} \text{ such that } c_1 |b_1\rangle + c_2 |b_2\rangle = |x\rangle$$

$$\langle b_1 | (c_1 |b_1\rangle + c_2 |b_2\rangle) = \langle b_1 | x \rangle$$

$$c_1 \langle b_1 | b_1 \rangle + c_2 \langle b_1 | b_2 \rangle = \langle b_1 | x \rangle$$

$$b_1, b_2 \text{ are orthogonal so } \langle b_1 | b_2 \rangle = 0 \text{ so}$$

$$c_1 = \langle b_1 | x \rangle$$

$$c_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} -2 \\ 1 \end{bmatrix} = \frac{-1}{\sqrt{2}}$$

$$\langle b_2 | (c_1 |b_1\rangle + c_2 |b_2\rangle) = \langle b_2 | x \rangle$$

$$c_1 \langle b_2 | b_1 \rangle + c_2 \langle b_2 | b_2 \rangle = \langle b_2 | x \rangle$$

$$b_1, b_2 \text{ are orthogonal so } \langle b_2 | b_1 \rangle = 0 \text{ so}$$

◆

$$c_2 = \langle b_2 | x \rangle$$

$$c_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \end{bmatrix} \begin{bmatrix} -2 \\ 1 \end{bmatrix} = \frac{-3}{\sqrt{2}}$$

Definition 1.27 Define $P_i := |f_i\rangle\langle f_i|$ to be the **projection operator** onto $\text{span}(|f_i\rangle)$ ♦

Proposition 1.28 Let $\mathcal{B} = \{|f_1\rangle, \dots, |f_n\rangle\}$ be an orthonormal basis for \mathbb{C}^n then...

1. $P_i(|v\rangle) \in \text{span}(|f_i\rangle)$
2. $|v\rangle - P_i|v\rangle$ is orthogonal to $|f_i\rangle$
3. $P_i^2 = P_i \star P_i = P_i$ Projection will still be the same.
4. $P_i P_j = 0$ shadow on one plane will be "nothing" in the perspective of another plane.
5. $P_i^\dagger = P_i$ self-adjoint
6. $\sum_{i=1}^n P_i = I_n$

Example 1.29 (Outer Product Example) Given $P_2 = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$, what is $P_1 + P_2$

$$\begin{aligned}
 P_1 &= |f_1\rangle\langle f_1| \\
 &= \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \star \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \end{bmatrix} \\
 &= \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \\
 P_1 + P_2 &= \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} \\
 &= \frac{1}{2} \begin{bmatrix} 0 & 2 \\ 2 & 0 \end{bmatrix} \\
 &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}
 \end{aligned}$$

This is an interesting result, it shows that if you add all of the projections in a space you will get the identity matrix, proof available elsewhere as I am too lazy to type it out. ♦

Why is projection useful?: It allows us to find c_1, \dots, c_k for $|x\rangle = c_1|f_1\rangle + \dots + c_k|f_k\rangle$ when all $|f\rangle$ are orthogonal to each other.

Theorem 1.30 (The Gram Schmidt Process) *Let $|b_1\rangle, \dots, |b_k\rangle$ be the basis for a subspace M in \mathbb{C}^n . Define the set of vectors $|f_1\rangle, \dots, |f_k\rangle$ as follows:*

Step 1:

Define $|f_1\rangle := |b_1\rangle$.

Step $i+1$:

for $i = 1:k-1$

$$|f_{i+1}\rangle := |b_{i+1}\rangle - \left(\sum_{j=1}^i \frac{|f_j\rangle \langle f_j|}{\langle f_j | f_j \rangle} \right) |b_{i+1}\rangle = |b_{i+1}\rangle - \left(\sum_{j=1}^i \frac{\langle f_j | b_{i+1} \rangle}{\langle f_j | f_j \rangle} \right) |f_j\rangle \quad (1.19)$$

Done

Now normalization step

for $i = 1 : k$

$$|f_i\rangle := \frac{1}{\| |f_i\rangle \|} |f_i\rangle \quad (1.20)$$

Done, now we know that $|f_1\rangle, \dots, |f_k\rangle$ is a orthonormal basis for M

A special thank you for Dr. Hamidi and Dr. Ismert for giving this psuedocode function for Gram Schmidt, I have pulled from it heavily in this page as it is the best way to explain this code.

2 Matrices and Linear Transformations

2.1 Matrices

This section assumes a basic knowledge of matrix operations, and should be a basic overview.

Definition 2.1 $M_{mn}(\mathbb{C})$ = set of all $m \times n$ matrices with complex entries. \blacklozenge

Definition 2.2 $M_n(\mathbb{C}) \equiv M_{nn}(\mathbb{C})$ meaning that it is just a square matrix with $n \times n$ size. \blacklozenge

Example 2.3 (Pauli Matrices)

$$\begin{aligned}\sigma_x &= \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \\ \sigma_y &= \begin{bmatrix} 0 & i \\ -i & 0 \end{bmatrix} \\ \sigma_z &= \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}\end{aligned} \in M_2 \quad (2.21)$$

These three matrices are called the **Pauli matrices** the the basic matrices we will use for quantum computations in the future. \blacklozenge

Definition 2.4 The **Identity Matrix** (I_n) is the matrix M_n whose columns are the standard basis $|e_1\rangle, \dots, |e_n\rangle$ for \mathbb{C}_n .

$$I_4 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (2.22) \quad \blacklozenge$$

Theorem 2.5 M_n has a well defined matrix multiplication, in general, if $A \in M_{mn}, B \in M_{nk}$.

The matrix $AB = [A|b_1\rangle \dots A|b_k\rangle]$

Example 2.6 (Pauli Matrices in action) For this example, we will be scaling the σ_y Pauli matrix.

$$\begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} 0 & -3i \\ i & 0 \end{bmatrix} = [1|a_1\rangle \quad 3|a_2\rangle] \quad (2.23)$$

This shows that in general, multiplying a row matrix by a diagonal matrix will scale the row matrix by the respective diagonal matrix value. \blacklozenge

2.2 Eigenvectors and Eigenvalues

2.2.1 Eigenvalues/Eigenvectors

Definition 2.7 An **eigenvalue** for $A \in M_n$ is a complex number $\lambda \in \mathbb{C}$ such that there is a nonzero vector $|x\rangle \in \mathbb{C}^n$ satisfying

$$A|x\rangle = \lambda|x\rangle \quad (2.24)$$

◆

Definition 2.8 A matrix $A \in M_n$ is **diagonalizable** if and only if:

1. There exists a diagonal matrix D and an invertible matrix P such that $A = PDP^{-1}$ if and only if
2. there exists a basis for \mathbb{C}^n consisting of eigenvectors for A .

◆

Example 2.9 Consider:

$$A = \begin{bmatrix} I_2 & 0 \\ 0 & \sigma_y \end{bmatrix} \equiv \begin{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} & \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \\ \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} & \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} \end{bmatrix} \quad (2.25)$$

1. Get eigenvalues and some normalized eigenvectors:

$$\begin{aligned} & \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & -i \\ 0 & 0 & i & 0 \end{bmatrix} - \begin{bmatrix} \lambda & 0 & 0 & 0 \\ 0 & \lambda & 0 & 0 \\ 0 & 0 & \lambda & 0 \\ 0 & 0 & 0 & \lambda \end{bmatrix} \\ &= \begin{bmatrix} 1-\lambda & 0 & 0 & 0 \\ 0 & 1-\lambda & 0 & 0 \\ 0 & 0 & \lambda & -i \\ 0 & 0 & i & -\lambda \end{bmatrix} \\ &= (1-\lambda)(1-\lambda)[\lambda^2 - 1] = 0 \\ &\lambda = 1, 1, 1, -1 \end{aligned} \quad (2.26)$$

2. Now plug in eigenvalue and find eigenvector:

$$\begin{aligned}
 A|e_1\rangle &= |e_1\rangle \\
 A|e_2\rangle &= |e_2\rangle \\
 A - \lambda * I &= \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & -i \\ 0 & 0 & i & -1 \end{bmatrix} \\
 \xrightarrow{R_4 + iR_3} & \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & i \\ 0 & 0 & 0 & 0 \end{bmatrix} \\
 \xrightarrow[-1R_1]{R_1 \leftrightarrow R_3} & \begin{bmatrix} 0 & 0 & 1 & -i \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}
 \end{aligned}$$

Now we must use this and transfer it into an equation to solve for x_3, x_4

$$x_3 = ix_4$$

Choose $x_4 = 1$

$$|x\rangle = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ i \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ i \end{bmatrix}$$

(2.27)

These first two arrays are just chosen because there are no x_1, x_2 values in the eigenvector array, so we choose two unit vectors.

3. Finally, normalize the vector.

$$|x\rangle = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 0 \\ i \\ 1 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 0 \\ 1 \\ i \end{bmatrix}$$

4. Final answer

$$P = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} * i \\ 0 & 0 & \frac{1}{\sqrt{2}} * i & \frac{1}{\sqrt{2}} \end{bmatrix}, D = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}$$

◆

2.2.2 Eigenvectors and Eigenvalues Big Picture

1. Self-Adjoint (Hermitian)

$A \in M_n$ is Hermitian if

$$\begin{aligned} A^\dagger &= A \\ \sigma_x^\dagger &= \sigma_x \\ \sigma_y^\dagger &= \sigma_y \\ \sigma_z^\dagger &= \sigma_z \end{aligned} \tag{2.28}$$

Additionally, all Hermitian matrices are diagonalizable, meaning their eigenvalues form an orthonormal basis.

2. Normal Matrices

$$A * A^\dagger = A^\dagger * A \leftrightarrow \text{eigenvectors form orthonormal basis} \tag{2.29}$$

3. Positive semi-definite matrix M_n is:

$$\begin{aligned} \forall |x\rangle \in \mathbb{C}^n, \text{ we have} \\ \langle x|A|x\rangle \geq 0 \text{ where } A|x\rangle = y \end{aligned} \tag{2.30}$$

4. (Unitary Matrices) The following are all equivalent:

- $U \in M_n$ is unitary
- $U^\dagger = U^{-1}$ or $(U^{-1} * U = U * U^{-1} = I_n)$
- $U * U^\dagger = I_n$ or $U^\dagger * U = I_n$ *note: U^\dagger not always equal to U , but they are normal.*
- The columns of U form an orthonormal basis for \mathbb{C}^n .
- $\forall |x\rangle, |y\rangle \in \mathbb{C}^n$ and $\langle U_x|U_y\rangle = \langle x|y\rangle$ meaning unitary matrices are rotation matrices.

3 Introduction to Quantum Theory

Now that we have laid out the mathematical framework necessary to understand this subject, it is time to lay out the physics necessary to complete the basis (no pun intended) for Quantum Information Theory.

3.1 Quantum Theory Axioms

These axioms are humans attempts at creating rules and methods of abstractions to describe and understand

3.1.1 Axiom 1.1: A vector state x is a unit vector in a complex Hilbert space.

We want to figure out the photon's (particle) polarization, to figure this out we are going to shoot it at a vertically polarized filter. Classically, $|x\rangle$ should be polarized \uparrow or \rightarrow , if this is the case we expect two outcomes:

1. If $|x\rangle$ goes through filter, then $|x\rangle$ is \uparrow .
2. If $|x\rangle$ is deflected, then $|x\rangle$ is \rightarrow .

3.1.2 Axiom 1.2: Linear combinations (or superposition) of the physical states are allowed to act as x vectors.

Physical States (Choices made by humans to describe quantum mechanics):

$$\uparrow = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \rightarrow = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Something of note, these two vectors form a basis for \mathbb{C}^2 and all combinations are:

$$|x\rangle = \alpha \begin{bmatrix} 0 \\ 1 \end{bmatrix} + \beta \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{ where } \|\alpha\|^2 + \|\beta\|^2 = 1$$

To keep track of this choice, we make the matrix $A = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$ representing vertically polarized filter.

Another note, the e-vals of that matrix are $|v\rangle, |h\rangle$.

Example 3.1 $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ might be a nice to represent a horizontally polarized filter. ♦

3.1.3 Axiom 2: Observable

An observable of a state $|x\rangle$ corresponds to a Hermitian matrix A . $|x\rangle$ is in state $|u\rangle$, an e-vect for A with probability $|\langle u|x\rangle|^2$

Definition 3.2 Expectation value of A (or mean value) of observable associated to A after measurements with respect to many copies of $|\psi\rangle$ is **the weighted average of the expected outcomes**.

$$\langle A \rangle_\psi = \sum_{i=1}^n |\langle u_i | \psi \rangle|^2$$

$$|\psi\rangle = \sum_{i=1}^n c_i |u_i\rangle \text{ where } u_i \text{ is an e-basis from A}$$

♦

3.1.4 Axiom 3: The time dependence of a state is governed by the Schrödinger equation

$$i\hbar \frac{\delta |\psi\rangle}{\delta t} = H |\psi\rangle \quad (3.31)$$

\hbar is reduced Planck's constant

H is Hermitian matrix corresponding to energy of the system **Hamiltonian**.

When H is time invariant (constant), the Schrödinger equation becomes:

$$|\psi(t)\rangle = e^{\frac{-itH}{\hbar}} |\psi(0)\rangle \quad (3.32)$$

3.2 More Introductory Quantum Concepts

Now that the basic framework of what a quantum state is, it is now time to apply these concepts and build on them to further characterize what a quantum system is, and why it is important to us. Prepare yourself, because we are jumping right into an example applying the previous axioms to characterize a *real world* quantum system.

Example 3.3 Consider a physical system with Hamiltonian $H = \frac{\hbar}{2}\omega\sigma_x$ and suppose $|\psi(0)\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ (The first column of σ_z .)

- Find the wave function $|\psi(t)\rangle, t > 0$, which by the Shrödinger equation is:

$$\begin{aligned}
 |\psi(t)\rangle &= \exp \frac{-itH}{\hbar} * |\psi\rangle \\
 &= \exp \left(i \left(\frac{-t}{\hbar} \right) \left(\frac{\hbar}{2} \omega \sigma_x \right) \right) \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\
 &= \exp i \left(\frac{t}{2} \omega \right) \sigma_x \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \text{where } \frac{t}{2} \omega \text{ is } \alpha \\
 &= \exp i(\alpha) \sigma_x \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\
 &= \left[\cos\left(\frac{t}{2}\omega\right) I + i \sin\left(\frac{t}{2}\omega\right) \sigma_x \right] \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\
 &= \begin{bmatrix} \cos\left(\frac{t}{2}\omega\right) & i \sin\left(\frac{t}{2}\omega\right) \\ i \sin\left(\frac{t}{2}\omega\right) & \cos\left(\frac{t}{2}\omega\right) \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\
 |\psi(t)\rangle &= \begin{bmatrix} \cos\left(\frac{\omega t}{2}\right) \\ i \sin\left(\frac{\omega t}{2}\right) \end{bmatrix}
 \end{aligned} \tag{3.33}$$

- Find the probability for the system to have outcome +1 upon measurement of σ_z

$$\begin{aligned}
 P_2 |\psi(t)\rangle &= \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \cos\left(\frac{\omega t}{2}\right) \\ i \sin\left(\frac{\omega t}{2}\right) \end{bmatrix} = \begin{bmatrix} \cos\left(\frac{\omega t}{2}\right) \\ 0 \end{bmatrix} \\
 P_{\downarrow}(t) &= \left| \cos \frac{\omega t}{2} \right|^2 = \cos^2 \frac{\omega t}{2}
 \end{aligned} \tag{3.34}$$

As you can see, we have picked off the cos out of ψ and in order to compute the probability distribution of the wave, we square the final answer, this will look familiar to a probability and statistics class topic of expected values and distributions (because it is... Wow! never thought it would be useful again)

- Find the probability for the system to have outcome -1 upon measurement of σ_z

$$P_2 |\psi(t)\rangle = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \cos(\frac{\omega t}{2}) \\ i \sin \frac{\omega t}{2} \end{bmatrix} = \begin{bmatrix} 0 \\ i \sin \frac{\omega t}{2} \end{bmatrix} \quad (3.35)$$

$$P_{\downarrow}(t) = |i \sin \frac{\omega t}{2}|^2 = \sin^2 \frac{\omega t}{2}$$

Same thing as the one above, but notice that this one is picking out the sin in ψ instead using the other diagonal term in the P matrix.

- Find the expectation value under many measurements of σ_z :

$$\begin{aligned} \langle \sigma_x \rangle_{\psi} &= \langle \psi(t) | \sigma_z \psi(t) \rangle = (+1)(\cos^2 \frac{\omega t}{2}) + (-1)(\sin^2 \frac{\omega t}{2}) \\ &= (\cos^2 \frac{\omega t}{2}) - (\sin^2 \frac{\omega t}{2}) \end{aligned} \quad (3.36)$$

As you can see, we have set matrix σ_z against vector $\psi(t)$, giving us the overall expected outcome if we measured wave function $\psi(t)$ in the σ_z "direction". ♦

3.3 Multipartite Physical States

3.3.1 Tensor Products

Definition 3.4 Tensor Product Consider vector space H with an inner product $H = H_1 \otimes H_2$ where \otimes is the **tensor product** combining H_1, H_2 in such a way that things in H_1 effect H_2 and vice versa. \blacklozenge

A general vector of H is a linear combination of vectors $\{|v_1\rangle \otimes |v_2\rangle : |v_1\rangle \in H_1, |v_2\rangle \in H_2\}$

Now to take the tensor of two matrices where $A \in M_{mn}, B \in M_{pq}$:

$$A \otimes B = A = \begin{bmatrix} A_{11}B & A_{12}B & \dots & A_{1n}B \\ A_{21}B & A_{22}B & & \\ \vdots & \vdots & \ddots & \vdots \\ A_{m1} & \dots & & A_{KK} \end{bmatrix} \in M_{(mp)(nq)} \quad (3.37)$$

This equation may look overwhelming at first glance, but it can be seen that all it is doing is applying the *elements* of A to the *matrix* B . This means that you will get a matrix that any change in A affects B , and vice versa just like we hoped.

Example 3.5

$$\sigma_x \otimes i\sigma_y = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \otimes \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} = \begin{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} & \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \\ \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} & \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \end{bmatrix} \quad (3.38)$$

Understand that this is actually just a 4×4 matrix, and was only written with the extra brackets in order to provide clarity. \blacklozenge

Rules of tensor products Now that we have learned a little about tensor products, lets learn some basic rules that apply to this operation:

1. $(\lambda A) \otimes B = A \otimes (\lambda B)$, $\lambda \in \mathbb{C}$ (*Scaling one matrix affects the other and vice versa*)
2. In general $A \otimes B \neq B \otimes A$ (*Generally noncommutative, not always true though*)
3. $(A \otimes B)(C \otimes D) = AC \otimes BD$ (*Distributivity law*)
4. $A \otimes (B + C) = (A \otimes B) + (A \otimes C)$ (*Another distributivity law*)
5. $(A \otimes B) + (C \otimes D) \neq (A + C) \otimes (B + D)$ (*Nonassociative*)
6. $(A \otimes B)^\dagger = A^\dagger \otimes B^\dagger$ (*Daggers are distributive*)
7. If A and B are invertible, $(A \otimes B)^{-1} = A^{-1} \otimes B^{-1}$ (*Inverses are distributive*)

Example 3.6 (Tensor product of two ket vectors)

$$\begin{aligned} |1\rangle \otimes |1\rangle &=: |11\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \otimes \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\ |1\rangle \otimes |2\rangle &=: |12\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \otimes \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} \\ |2\rangle \otimes |1\rangle &=: |21\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \otimes \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} \\ |2\rangle \otimes |2\rangle &=: |22\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \otimes \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \end{aligned}$$

Note: $\{|11\rangle, |12\rangle, |21\rangle, |22\rangle\}$ forms a basis for \mathbb{C}^4 but these vectors live in $\mathbb{C}^2 \otimes \mathbb{C}^2 \approx \mathbb{C}^4$ ♦

This example showed a property of tensor products in which if $|v\rangle \in \mathbb{C}^m, |w\rangle \in \mathbb{C}^n$, then $|wv\rangle := |v\rangle \otimes |w\rangle \in \mathbb{C}^m \otimes \mathbb{C}^n \approx \mathbb{C}^{mn}$, so the result is the two dimensions multiplied.

Since we have done so much fun stuff this chapter, lets finish it off with something fun, a proof!

Theorem 3.7 (Tensor Product Basis)

If H_1 has ONB $\varepsilon_1 = \{|e_{1,1}\rangle, |e_{1,2}\rangle, \dots, |e_{1,m}\rangle\}$ and H_2 has ONB $\varepsilon_2 = \{|e_{2,1}\rangle, |e_{2,2}\rangle, \dots, |e_{2,n}\rangle\}$

Note: First subscript is the basis where the vector is formed.

Then $H := H_1 \otimes H_2$ has an orthonormal basis $\{|e_{1,i}\rangle \otimes |e_{2,j}\rangle : 1 \leq i \leq m, 1 \leq j \leq n\}$

Proof. Let $\sum_p \lambda_p |v_p\rangle \otimes |w_p\rangle \in H$

It suffices to show that just one elementary tensor $|v_p\rangle \otimes |w_p\rangle$ belongs to $\text{Span}(\varepsilon)$, so since ε_1 is a basis for H_1 and $|v_p\rangle \in H_1$, we can write

$$|v_p\rangle = c_1 |e_{1,1}\rangle + c_2 |e_{1,2}\rangle + \dots + c_m |e_{1,m}\rangle \text{ for some } c_1, \dots, c_m \in \mathbb{C}$$

$$|w_p\rangle = d_1 |e_{2,1}\rangle + d_2 |e_{2,2}\rangle + \dots + d_n |e_{2,n}\rangle \text{ for some } d_1, \dots, d_n \in \mathbb{C}$$

$$\begin{aligned} \text{So } |v_p\rangle \otimes |w_p\rangle &= (c_1 |e_{1,1}\rangle + c_2 |e_{1,2}\rangle + \dots + c_m |e_{1,m}\rangle) \otimes (d_1 |e_{2,1}\rangle + d_2 |e_{2,2}\rangle + \dots + d_n |e_{2,n}\rangle) \\ &= \left(\sum_{i=1}^m c_i |e_{1,i}\rangle \right) \otimes \left(\sum_{j=1}^n d_j |e_{2,j}\rangle \right) \\ &= \sum_{i=1}^m \sum_{j=1}^n c_i d_j |e_{1,i}\rangle \otimes |e_{2,j}\rangle \\ &= \sum_{i=1}^m \sum_{j=1}^n c_i d_j (|e_{1,i}\rangle \otimes |e_{2,j}\rangle) \in \text{Span}(\varepsilon) \text{ by tensor product fact 1} \end{aligned}$$

What is the inner product of H ?

$$\begin{aligned} \text{let } |v_1\rangle, |w_1\rangle \in H_1 \text{ and } |v_2\rangle, |w_2\rangle \in H_2 \\ \langle |v_1\rangle \otimes |v_2| | |w_1\rangle \otimes |w_2\rangle \rangle &:= \langle v_1 v_2 | w_1 w_2 \rangle \\ &= \langle v_1 | w_1 \rangle_{H_1} \langle v_2 | w_2 \rangle_{H_2} \text{ Inner products of first factors form first set and same for second} \end{aligned}$$

Finally we must check that ε is an orthonormal set.

Let $|e_{1,i}\rangle \otimes |e_{2,j}\rangle, |e_{1,k}\rangle \otimes |e_{2,l}\rangle \in \varepsilon$

Then $\langle e_{1,i} e_{2,j} | e_{1,k} e_{2,l} \rangle = \langle e_{1,i} | e_{1,k} \rangle_{H_1} * \langle e_{2,j} | e_{2,l} \rangle_{H_2}$

That means this breaks down to when $i=k$ and $j=l$ otherwise they are orthogonal so

$= \langle e_{1,i} e_{2,j} | e_{1,i} e_{2,j} \rangle$ which must be 1 since they are parallel so ε is an orthonormal set.

(3.39)

■

3.3.2 Quantum Multipartite Systems

$$H = H_1 \otimes H_2 \otimes \cdots \otimes H_n \text{ is a multipartite system} \quad (3.40)$$

*Note: if n is two then the system is **bipartite***

Definition 3.8 (Bipartite state) $|\psi\rangle$ state in $H_1 \otimes H_2$ is called a **bipartite state** ♦

Definition 3.9 (Separability) A vector $|\psi\rangle \in H = H_1 \otimes H_2$ is **separable** (or an elementary tensor) if $|\psi\rangle = |\psi_1\rangle \otimes |\psi_2\rangle$ for some $|\psi_1\rangle \in H_1, |\psi_2\rangle \in H_2$ ♦

In plain English, separability states that there are vectors that can be tensored together to create the final tensor, an example below will illustrate this better.

Example 3.10 Is $\begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \in \mathbb{C}^4 (\approx \mathbb{C}^2 \otimes \mathbb{C}^2)$ an elementary tensor?

$$\begin{bmatrix} 0 \\ 1 \end{bmatrix} \otimes \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 * \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ 1 * \begin{bmatrix} 1 \\ 0 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} \quad (3.41)$$

It *is* separable. ♦

Definition 3.11 A vector that is *not* separable is called **entangled**. ♦

Remark 3.12 A state $\psi = C_{11}|11\rangle + C_{12}|12\rangle + C_{21}|21\rangle + C_{22}|22\rangle$ is separable only if (in \mathbb{C}^2) rank is one (i.e. only if its rows are scalar multiples of each other), where the coeff matrix is $\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$. ♦

Most of the time, finding if $|\psi\rangle \in H_1 \otimes H_2$ is separable is not as easy as the exercises as above, this is because most of the time we cannot look at a matrix and infer the vectors being tensored together, this is where the next section comes into play, and we use a singular value decomposition.

3.3.3 Singular Value Decomposition

First things first, we must know that everything that is about to be shown is only applicable in bipartite systems, anything larger than that is an impossible, and yes that really means impossible, task in most situations because of the properties of quantum systems and their probabilistic nature. If you can figure out an algorithmic way to solve the decomposition of multipartite systems, you will be very famous and more than likely have a decomposition named after you. Now that we have laid that out there, lets continue with singular value decompositions.

The Singular Value Decomposition takes on the form $C = U\Sigma V^\dagger$, this will mean very little, but keep this in mind as you move through the example as this is the motivation to the steps below.

Definition 3.13 If $A \in M_{mn}$, then the **singular values** for A are the square roots of the e-vals for $A^\dagger A \in M_{nn}$. \blacklozenge

Fact: For any $A \in M_{mn}$, $A^\dagger A$ is positive semi-definite, meaning it is Hermetian and its eigenvalues are non-negative.

Example 3.14 Find the singular value decomposition (SVD) for $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ i & i \end{bmatrix}$

1. Compute eigenvalues and eigenvectors

$$A^\dagger A = \begin{bmatrix} 1 & 0 & -i \\ 1 & 0 & -i \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ -i & -i \end{bmatrix} = \begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix} \quad (3.42)$$

Find the eigenvalues

$$\det(A^\dagger A - \lambda I) = 0 \rightarrow \boxed{\lambda = 0, 4}$$

Find the eigenvectors

$$\lambda = 0 \rightarrow A^\dagger A |x\rangle = |0\rangle \rightsquigarrow |x\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\lambda = 4 \rightarrow A^\dagger A |x\rangle = 4|x\rangle \rightsquigarrow |x\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Now we must take the spectral decomposition to get the e-vec matrix (see below). We are allowed to take the spectral decomposition because we know that $A^\dagger A$ is positive semi definite.

$$A^\dagger A = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix} \left(\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \right)^\dagger \quad (3.43)$$

2. Make sure V 's columns are in decreasing order with respect to eigenvalues.

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

$$\Sigma = \begin{bmatrix} 2 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \leftarrow \begin{bmatrix} \sqrt{\lambda_1} & 0 \\ 0 & \sqrt{\lambda_2} \\ 0 & 0 \end{bmatrix} \text{ (This is generally how it is)}$$

3. **Construct more e-vectors** $A|\lambda_1\rangle, A|\lambda_2\rangle$

$$A|\lambda_1\rangle = \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ i & i \end{bmatrix} \left(\frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} \frac{2}{\sqrt{2}} \\ 0 \\ \frac{2i}{\sqrt{2}} \end{bmatrix} = \sqrt{2} \begin{bmatrix} 1 \\ 0 \\ i \end{bmatrix}$$

$$A|\lambda_2\rangle = \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ i & i \end{bmatrix} \left(\frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right) = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \text{ This one is useless as there is no new info}$$

4. **Now use** $|u_1\rangle = \sqrt{2} \begin{bmatrix} 1 \\ 0 \\ i \end{bmatrix}$ **to create an ONB for** \mathbb{C}^3 .

$$\{|u_1\rangle, |u_2\rangle, |u_3\rangle\} \rightsquigarrow \{|u_1\rangle, |2\rangle, |3\rangle\} \text{ where } |2\rangle = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \text{ and } |3\rangle = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

This is linearly independent but not orthogonal, we must use Gram-Schmidt to change that:

5. **Apply the Gram-Schmidt Process to the basis** My poor fingers tire of this, therefore I implore you to go look at the Gram-Schmidt example above to understand how this process works. If you don't believe me, feel free to try it for yourself and email me telling me of my stupidity if it is indeed wrong. Thank you ☺.

Final Orthonormal Basis:

$$|u_1\rangle = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ \frac{i}{\sqrt{2}} \end{bmatrix}$$

$$|u_2\rangle = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

$$|u_3\rangle = \begin{bmatrix} \frac{i}{\sqrt{2}} \\ 0 \\ \frac{1}{\sqrt{2}} \end{bmatrix}$$

6. Form U vector with orthonormal basis calculated above

$$U = \begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{i}{\sqrt{2}} \\ 0 & 1 & 0 \\ \frac{i}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \end{bmatrix} \text{ is unitary} \quad (3.44)$$

7. Finally, assemble the SVD!

$$C = U\Sigma V^\dagger = \begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{i}{\sqrt{2}} \\ 0 & 1 & 0 \\ \frac{i}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}^\dagger \quad (3.45)$$

Whew, that was a hard process.. but now that we have this set of matrices, you can see that we can recover A through these, super useful! \blacklozenge

3.3.4 Schmidt Decomposition

Theorem 3.15 *Let $H = H_1 \otimes H_2$ where $\dim H_1 = m < \text{infinity}$ and $\dim H_2 = n < \text{infinity}$, every vector $\psi \in H$ admits a Schmidt Decomposition:*

$$\psi = \sum_{j=1}^r S_j |u_j\rangle \otimes |v_j\rangle \quad (3.46)$$

where $s_j > 0$ are the **Schmidt Coefficients** satisfying $\sum_{j=1}^r S_j^2 = 1$, $\{|u_j\rangle\} \leq H_1$ and $\{|v_j\rangle\} \leq H_2$ are orthonormal, and $r \leq \min\{m, n\}$ is called the **Schmidt number**.

Note: r is just the rank of the coeff matrix

Remark: $|\psi\rangle \in H_1 \otimes H_2$ is separable iff it's Schmidt number is 1.

3.4 Mixed States as Density Matrices

In this section, we will learn how to describe quantum systems with density matrices, which we will find has many advantages compared to the ways we have learned to describe systems thus far, so before we get into any of the math, let us discuss the motivation more specifically.

3.4.1 Motivation for density matrices

Quantum particles are waves, and as such, there is some specificity when it comes to describing those particles with respect to time. This would be a problem if two of the same particles were measured, there would be no guarantee that the particle would behave in the same way because of this phase shift with respect to time. With a density matrix, we encompass the phase shift within the projection (math for this shown below) in creating one, meaning that we can describe particles regardless of their phase shift. Now let's get into how to make one.

3.4.2 Creating a density matrix

Example 3.16 Suppose $|\psi\rangle \in \mathbb{C}^2$ is the state of some quantum system. That means that $\| |\psi\rangle \| = 1$ and if we were hoping to measure vertical or horizontal polarization.

Let being vertically polarized be represented by:

$$|1\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \text{and} \quad |0\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

These values should come from the eigenstates from chosen Hermitian matrix which represents the apparatus.

$$\begin{aligned} A &= 0|0\rangle\langle 0| + 1|1\rangle\langle 1| = |1\rangle\langle 1| \\ &= \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \text{ This can be assigned as the "apparatus" } \quad \blacklozenge \end{aligned}$$

To glean information about ψ regarding its polarization, we use the equation

$$\begin{aligned} \psi &= \alpha|0\rangle + \beta|1\rangle \\ \text{where } \|\psi\| &= \sqrt{\langle \psi | \psi \rangle} = \sqrt{\|\alpha\|^2 + \|\beta\|^2} = 1 \end{aligned}$$

We interpret this expression of ψ in the $\{|0\rangle, |1\rangle\}$ -base as:

$$|\psi\rangle \text{ is in } |0\rangle \text{ with } \|\alpha\|^2 \text{ probability and } |\psi\rangle \text{ is in } |1\rangle \text{ with } \|\beta\|^2 \text{ probability}$$

Recall: $|\psi_1\rangle = \alpha|0\rangle + \beta|1\rangle$ when we take this, and let it develop over time, the Schrödinger equation states that time development creates a phase shift in the particle vector.

This leads to $|\psi_2\rangle = e^{i\theta}\alpha|0\rangle + e^{i\theta}\beta|1\rangle$ and it is in $|0\rangle$ with $\|e^{i\theta}\alpha\|^2$ probability and is in $|1\rangle$ with

$\|e^{i\theta}\beta\|^2$ probability.

*Note: Something interesting is that based on this is that $|\psi_1\rangle \neq |\psi_2\rangle$ as **vectors** in \mathbb{C}^2 but $|\psi_1\rangle = |\psi_2\rangle$ as **states** of the system in "quantum land".*

As we can see, it is going to be really hard to understand what state a particle is in this format, but this is where density matrices come in.

Consider:

$$\begin{aligned}\rho_{\psi_1} &= |\psi_1\rangle \langle \psi_1| = (\alpha|0\rangle + \beta|1\rangle)(\alpha^*\langle 0| + \beta^*\langle 1|) \\ &= |\alpha|^2|0\rangle\langle 0| + \alpha\beta^*|0\rangle\langle 1| + \alpha^*\beta|1\rangle\langle 0| + |\beta|^2|1\rangle\langle 1| \\ \rho_{\psi_1} &= \begin{bmatrix} |\alpha|^2 & \alpha\beta^* \\ \alpha^*\beta & |\beta|^2 \end{bmatrix}\end{aligned}$$

Now let's try the same thing with $|\psi_2\rangle$

$$\begin{aligned}\rho_{\psi_2} &= |\psi_2\rangle \langle \psi_2| = e^{i\theta}\alpha(e^{i\theta}\alpha)^* = e^{i\theta}\alpha * e^{-i\theta}\alpha^* = \alpha\alpha^* = |\alpha|^2 \dots \\ &= \begin{bmatrix} |\alpha|^2 & \alpha\beta^* \\ \alpha^*\beta & |\beta|^2 \end{bmatrix} \checkmark\end{aligned}$$

If you continue this you will see that they are the same matrix, meaning that using these matrices we can eliminate the time dependence of the vectors, and still keep the same data about the states.

3.4.3 Density Matrix Observations

Example 3.17 Given $|\psi\rangle = \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$ or $|\psi\rangle = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}$

Find ρ_ψ (w.r.t. $\{|0\rangle, |1\rangle\}$):

$$\rho_\psi = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix} \quad (3.47) \quad \blacklozenge$$

Now, a couple observations about density matrices:

- $\rho_\psi \in M_n(\mathbb{C})$
- ρ_ψ has trace 1 (i.e. $\text{Tr}(\rho_\psi) = 1$ where $\text{Tr}()$ is the sum of the diagonal matrices)
- ρ_ψ is **Hermetian**, and in fact furthermore **positive semi-definite**.

Definition 3.18 (Trace) The **trace** of an nxn matrix is the sum of its diagonal entries. \blacklozenge

Definition 3.19 (Density Matrix) A **density matrix** $\rho \in M_n$ is a Hermetian, positive semi-definite matrix such that $\text{tr}(\rho) = 1$. \blacklozenge

3.4.4 Density Matrix Axioms

- Axiom 1': A physical state of a system, whose Hilbert Space \mathbb{H} , is completely determined by its associated density matrix.
- Axiom 2': The mean value of an observable A is $\langle A \rangle_\psi = \text{tr}(\rho A)$
- Axiom 3': The time evolution of density matrix is given by the Louisvillie - von Neumann equation

$$i\hbar \frac{d}{dt} \rho = [H, \rho] \quad (3.48)$$

3.4.5 What about bipartite?

I can already hear the cries from the judgmental people reading this document, "we will never have a single particle system". You are completely right, but we have to walk before we run young padawan, be patient. In this next section, we will discuss systems of tensor products of multiple states, and how they are represented in density matrices.

Definition 3.20 Let $\rho \in M_n(\mathbb{C}) \otimes M_m(\mathbb{C})$ be a density matrix.

Recall: $\rho = \sum_{i=1}^N c_i \underbrace{A_{1,i}}_{\in M_n(\mathbb{C})} \otimes \underbrace{A_{2,i}}_{\in M_m(\mathbb{C})}$

1. ρ is called "uncorrelated" if:

$$\rho = \rho_1 \otimes \rho_2 \text{ where } \rho_i \text{ is a density matrix. (i.e. simple factorization)}$$

2. ρ is "separable" if:

$$\rho = \sum_{i=1}^N p_i \rho_{1,i} \otimes \rho_{2,i}. \text{ (i.e. factorable, but not actually two elementary tensors, actually a sum of multiple)}$$

3. If ρ is not "separable", so we call ρ inseparable. (i.e. NO factorization)

Example 3.21 Find the tensor product of the quantum vector.

$$\begin{aligned}
 |\psi\rangle &= |02\rangle \in \mathbb{C}^2 \otimes \mathbb{C}^3 \\
 &= \begin{bmatrix} 1 \\ 0 \end{bmatrix} \otimes \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \in \mathbb{C}^6
 \end{aligned} \tag{3.49}$$

$$\rho = |02\rangle \langle 02| = \underbrace{\begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{\text{"Tells the row"}} \underbrace{\begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 \end{bmatrix}}_{\text{"Tells the column"}} = \underbrace{\begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}}_{\text{"3rd row, 3rd col"}} \tag{3.50}$$

Definition 3.22 Partial Trace The partial trace of A over \mathbb{C}^m is

$$A_1 = \sum_{i=1}^N c_i \underbrace{A_{1,i}}_{\text{elementary tensor}} \otimes \underbrace{A_{2,i}}_{\text{elementary tensors}} \tag{3.51}$$

What is the point of the partial trace?

The partial trace gives you information about the specific particle instead of the overall system as you are looking at the elementary tensors of the system.

4 Qubits and Information Theory

Finally! I know, there has been large amounts of math up until this point, and there will continue to be, but with all of this math I hope that you have not lost sight of the purpose of this document: quantum computing! All of this math was designed to give you some perspective on what is happening in these computers, and this is the section in which we will connect these math concepts to the quantum concepts integral to quantum computing, but first lets start with a quick review of classical computers.

4.1 Classical and Quantum Computer Comparisons

As a computer engineer, this is where I start to get *super* interested, no offense to the mathematicians out there, but I find that placing these concepts to a real world parallel makes everything more interesting, so this first section will be a slight history lesson on how computers have worked since their inception, and how quantum computers solve many of the problems of classical computers, as well as create a whole new set of problems.

4.1.1 Classical Computer Basics

- Boolean bit is either a 1 or 0.
- Data is transmit in strings of bits.

Example 4.1 A bit of length 8 is a byte. 00100100 in utf-8 is "\$". ☺

◆

4.1.2 Quantum Computer Basics

Definition 4.2 Qubit A **qubit** is a unit vector in \mathbb{C}^2

$$|\psi\rangle = a|0\rangle + b|1\rangle \text{ such that } \|a\|^2 + \|b\|^2 = 1 \quad (4.52)$$

◆

Remark 4.3 Some other facts about qubits

- To extract info from a qubit, make a measurement which results in **collapse** to a $|0\rangle$ or $|1\rangle$.
- Qutrits (\mathbb{C}^3) and qudits (\mathbb{C}^n) also exist, but are not discussed in this paper.

◆

Definition 4.4 A group or system of n qubits is called a quantum register.

◆

4.1.3 What problems does a quantum computer solve?

As you can see, for a quantum computer, there is little crossover, but the main point is that while a classical computer can represent information as a 0 or 1, a quantum computer can represent the same information with a 0, 1, or anything in between, with the particle in "superposition". An example of this contrast of the two computers will follow.

Example 4.5 Classical bit stream Suppose we have a bit stream, xyzw, of length 4. Also suppose that each bit in this bitstream is determined completely independent of each other. This would mean that there is a maximum of 2^4 possibilities, and we are choosing 4 out of those to represent our bitstream. This is quite a bit of data, that can be stored in those four bits, but this pales in comparison when looking at a quantum register. The example below illustrates that situation. ♦

Example 4.6 Suppose we can describe a quantum register of n qubits analogous to classical framework... meaning we can write each qubit as

$$\begin{aligned} a_i |0\rangle + b_i |1\rangle \text{ for } i = 1, \dots, n. \\ (a_1 |0\rangle + b_1 |1\rangle) \otimes (a_2 |0\rangle + b_2 |1\rangle) \otimes \dots \otimes (a_n |0\rangle + b_n |1\rangle) \in \mathbb{C}^{2^n} \end{aligned} \quad (4.53)$$

The main problem here is that we have superposition, we need to include the tangled vectors along with the pure states. this means that a state of the system should be

$$|\psi\rangle = \sum a |i_1\rangle \otimes |i_2\rangle \otimes |i_3\rangle \otimes \dots \otimes |i_n\rangle \quad (4.54)$$

Written differently, this is a linear combination of basis vectors:

$$\{|0000\rangle, |0001\rangle, |0010\rangle, \dots, |1111\rangle\} \quad (4.55)$$

As you can see, in the same four quantum bits, we can store infinitely more data in the same four bits because we can have some linear combination of the same bits, all because of superposition. This is only one of the powerful aspects of quantum computing, more will be discussed in future paragraphs. ♦

4.1.4 More common quantum states

Lets look at a different basis, far different from our standard basis we use.

Example 4.7 Suppose we have two qubits ($\mathbb{C}^2 \otimes \mathbb{C}^2$), the set

$$\begin{aligned} \{ |\phi^+\rangle = \frac{1}{\sqrt{2}}(|00\rangle + |11\rangle), \quad |\phi^-\rangle = \frac{1}{\sqrt{2}}(|00\rangle - |11\rangle), \\ |\psi^+\rangle = \frac{1}{\sqrt{2}}(|01\rangle + |10\rangle), \quad |\psi^-\rangle = \frac{1}{\sqrt{2}}(|01\rangle - |10\rangle) \} \end{aligned} \quad (4.56)$$

This is called the Bell Basis, another orthonormal basis for $\mathbb{C}^2 \otimes \mathbb{C}^2$. Something about this basis is very different from other basis we have seen in the past, can you see it? None of these states are separable, *each of the Bell states is **entangled**!* This will become more important when we further understand why superposition is so powerful, but I will save that definition when we start to talk about cryptography. ♦

4.1.5 Error recovery and avoidance

If there is one thing that we all know, it is that there are two things that are actually perfect: sweet baby Jesus and Grandma's pecan pie. Quantum bits are neither of these, so they are in fact not perfect. Therefore we need avenues to safeguarding and detecting against imperfections in the line, whether that be noise, some delay, or a software reception error. Let us first look at how this is solved in a classical manner then we can approach how it is solved in the quantum realm.

Classical repetition codes Suppose a user sends 1011 over a channel, but Sally, the person receiving the message, gets 1001.

We must be able to identify that there has been an error in the third bit. One of the simplest ways to avoid this is by sending four of each bit, like this: 1111 0000 1111 1111.

Now suppose that Sally receives 1111 0100 1110 1110. She can take the bit that shows up most commonly in these four bit strings, and choose that as the bit to be used in that slot. The more bits there are, the lower the chance that there will be an error in the message transmission.

Woohoo, that is awesome, surely it is that simple to do the same in the quantum realm right? **WRONG, SO SO WRONG, HOW NIAVE OF YOU.** This fails for a couple reasons:

1. Measurements cause collapse, meaning that as soon as we measure that bit, it is useless to us, at least in a quantum sense.
2. All bits are distinct to one another, so sending "the same bit" carries a different weight when thinking about quantum bits compared to classical bits.

So now we know that there is no way to send redundant information because of the properties of particles in superposition, but there must be a way to interact with quantum bits without collapsing them right? There is, it is through quantum gates, which simply put are unitary transformations applied to qubits. Just because we can use quantum gates to interact with qubits does not mean that we are able to freely clone systems with these gates to accomplish this redundancy, in fact, this is the opposite of what we can do, a theory called the no-cloning theorem disproves this. I may add it if I have time, if not, sorry. We will see how my fingers feel after I finish this chapter.

4.2 Qubits and Quantum Gates

Looking back, it seems that we jumped straight into this wild world of qubits and quantum gates, but I do not think there was every a simple definition of what a qubit is, so lets review that before jumping into another section.

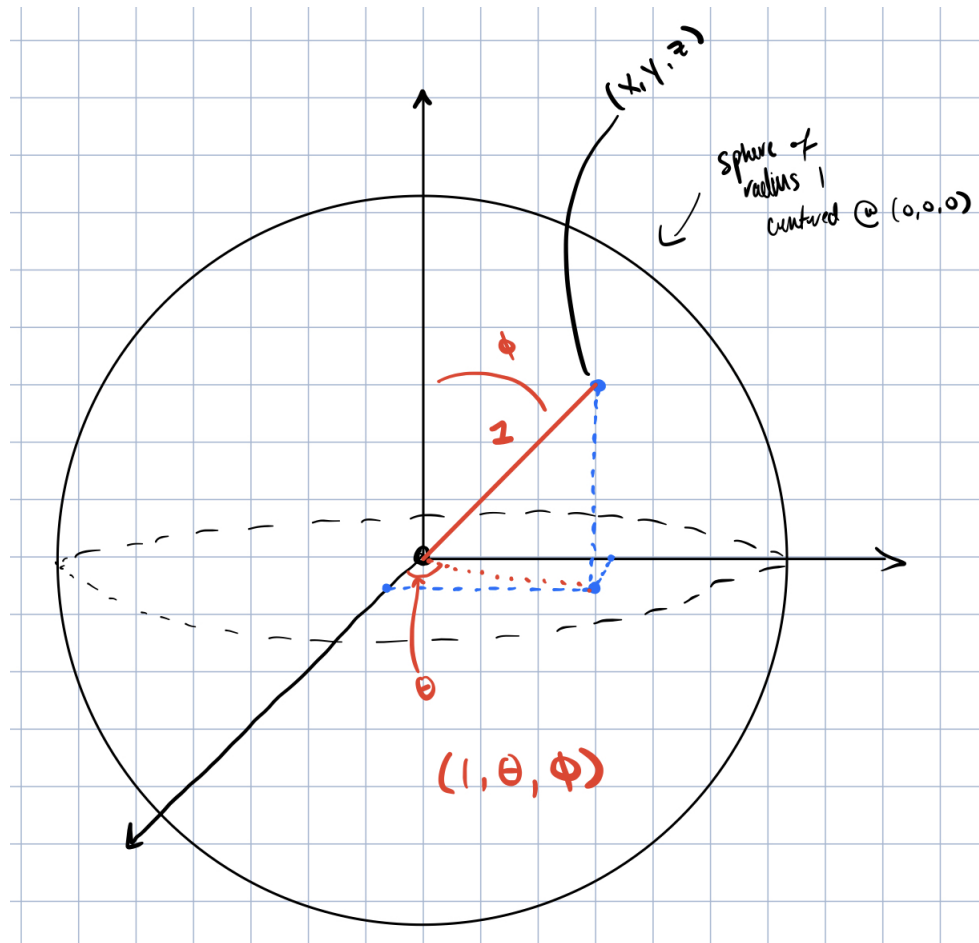
Qubit Review A qubit is simply a linear combination of two of our standard basis vectors, written in math it is

$$|\psi\rangle = \alpha \underbrace{|0\rangle}_{\begin{bmatrix} 1 \\ 0 \end{bmatrix}} + \beta \underbrace{|1\rangle}_{\begin{bmatrix} 0 \\ 1 \end{bmatrix}} \text{ where } \alpha = \alpha_1 + \alpha_2 i, \beta = \beta_1 + \beta_2 i \quad (4.57)$$

Well this is all well and good, but we really need another way to visualize this that can get rid of these pesky complex numbers. This should ring some bells for any of you math enthusiasts out there, it is time to break out some 3d polar coordinates! In this case, we plan to use the Bloch Sphere.

4.2.1 Bloch Sphere Introduction

As mentioned above, the Bloch Sphere is a polar coordinate system used to describe quantum particles, while minimizing the complex plane interaction. I think the easiest way to understand it is to see a picture of it then dive into the math, so check this out:



As you can see, the coordinate system we use for this is ($\underbrace{1}_{\text{Magnitude}}$, $\underbrace{\theta}_{\text{y-axis rotation}}$, $\underbrace{\phi}_{\text{Downward z-direction}}$), as you can see, the highest magnitude of a point is 1, meaning that it lies on the outer rim of the graph, with 100% of the system "accounted for". The following two entries in the coordinates will be described in depth as they are much more complicated.

What is ψ when in polar form?

$$\begin{aligned}
|\psi\rangle &= \mathbb{C}^2 \\
|\psi\rangle &= \alpha|0\rangle + \beta|1\rangle \\
|\psi\rangle &= \underbrace{|\alpha|e^{i\gamma_1}}_{\text{Polar form of } \alpha} |0\rangle + \underbrace{|\beta|e^{i\gamma_2}}_{\text{Polar form of } \beta} |1\rangle \\
|\psi\rangle &= \boxed{\underbrace{e^{i\gamma_1}}_{\text{Global phase shift}} (|\alpha||0\rangle + |\beta| \underbrace{e^{i(\gamma_2-\gamma_1)}}_{\text{Relative phase shift}} |1\rangle)}
\end{aligned} \tag{4.58}$$

What is θ, ϕ ? Now that we have the polar form of the state vector ψ , we can derive what θ and ϕ are. Let us first define

$$\underbrace{\theta}_{[0, 2\pi]} = \gamma_2 - \gamma_1 \tag{4.59}$$

Simplifying what this math is saying is take the global phase shift, set this as the "perspective", then we can look at how far out of phase our β is and choose that as our relative phase shift. This means we can completely eliminate the imaginary component from *alpha* as we are setting that vector as the point of view. Now lets find ϕ .

Find $\underbrace{\phi}_{[0, \pi]}$ such that $(\cos(\frac{\phi}{2}), \sin(\frac{\theta + \phi}{2})) = (|\alpha|, |\beta|)$, that means that

$$|\psi\rangle \underbrace{\approx}_{\text{"Up to a global phase shift"}} \cos(\frac{\phi}{2})|0\rangle + \sin(\frac{\theta}{2})e^{i\theta}|1\rangle \tag{4.60}$$

Lets look at an example of what this equation looks like with actual numbers.

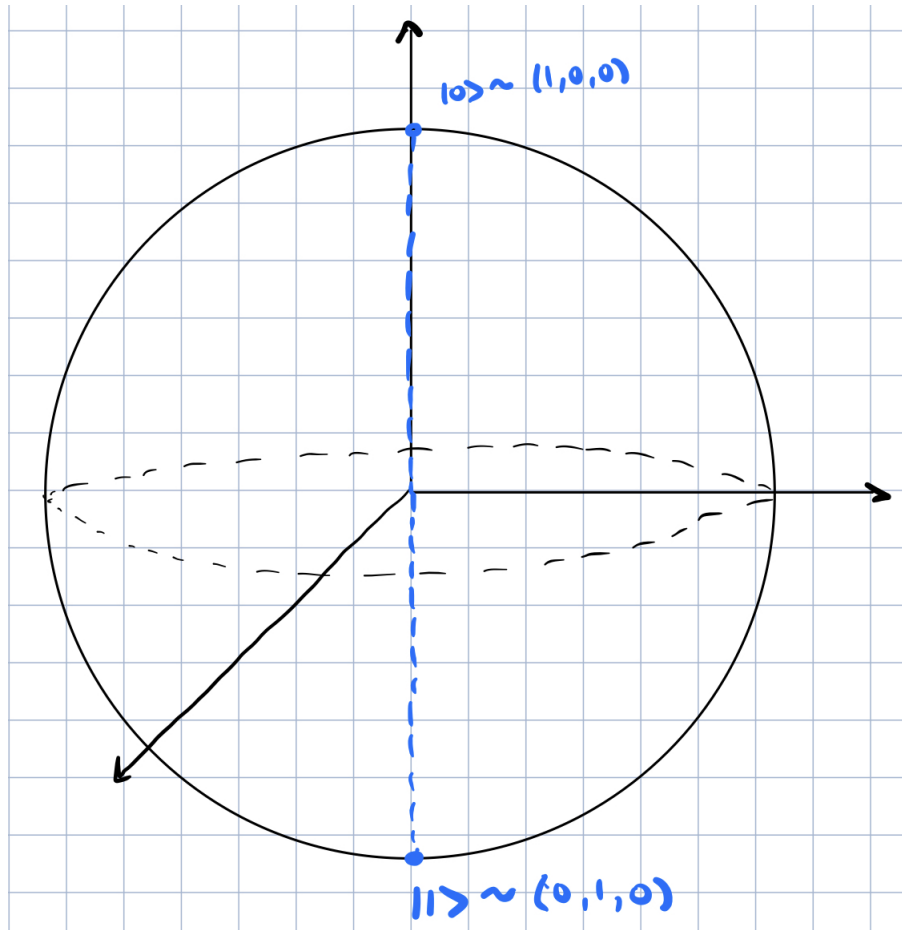
Example 4.8

$$\begin{aligned}
|0\rangle &\rightarrow \alpha = 1, \beta = 0 \\
\cos(\frac{\phi}{2}) &= 1 \quad \sin(\frac{\phi}{2}) = 0 \\
&\rightarrow \phi = 0 \\
&\rightarrow \theta = 0 \\
|0\rangle &\approx (1, 0, 0)
\end{aligned} \tag{4.61}$$

$$\begin{aligned}
|1\rangle &\rightarrow \alpha = 0, \beta = 1 \\
\cos(\frac{\phi}{2}) &= 0 \quad \sin(\frac{\phi}{2}) = 1 \\
&\rightarrow \phi = \pi \\
&\rightarrow \theta = 0 \\
|1\rangle &\approx (0, 1, 0)
\end{aligned} \tag{4.62}$$

As you can see in both of these we do not have any theta, that is because these are both pure states, meaning there is no relative phase shift, or more so it does not matter because we are able to get 0 or 1 from the sin and cos, this will change with states that are in some state of superposition, meaning that relative phase shift matters. ♦

Before we move on lets look at what that looks like on the Bloch Sphere:



As you can see, they are on the two "north/south" poles on the sphere, further reinforcing why that theta does not matter: because that spin on the y axis would not change how the vector looks! Very interesting!

4.3 Quantum Gates

Well we know what these particles look like, and we know how they are represented, but we still don't know how to do anything with them! Good thing we are about to learn about quantum gates. As we mentioned before, these gates are very important with doing things with the particles, more importantly without collapsing them. Lets get into it.

4.3.1 Classical Gate Overview

I have little desire to completely rehash classical logic gates, but in general it is something that changes a 0 to a 1, or swap, or clone. A couple quick examples are below.

Example 4.9

$$\begin{aligned} I : 0 &\rightarrow 0 \\ NOT : 0 &\rightarrow 1 \end{aligned} \tag{4.63}$$

As you can see, the identity passess the value of the bit through, while the not gate flips the value. This may seem elementary at first but combining these into more complex systems can lead to large and very complex operations. ♦

4.3.2 Quantum Gate Introduction

So real quick let us recall what a quantum gate is, it is just a unitary transformation (i.e. unitary matrices applied to our qubits). Something interesting is because they are unitary transformations, this means that all of these operations can be reversed. Lets look at a couple common ones.

Example 4.10

$$\begin{aligned} U : |0\rangle &\rightarrow \alpha |0\rangle + \beta |1\rangle \\ \underbrace{\sigma_x = X}_{\text{QNOT}} : |0\rangle, |1\rangle &\rightarrow 1 \\ -i\sigma_Y = Y &= |0\rangle, |1\rangle \rightarrow |-1\rangle, |0\rangle \\ \sigma_z = Z &= |0\rangle, |1\rangle \rightarrow |0\rangle, |-1\rangle \\ H &= |0\rangle, |1\rangle \rightarrow |\sigma_x^+\rangle, |\sigma_x^-\rangle \end{aligned} \tag{4.64}$$

As you can see we are using the Pauli spin matrices to make these quantum gates, but one of these gates may seem foreign, and rightfully so. "H" is the Hadamard gate, which if you can see, puts the particle into a state of complete superposition. This is an important gate that is used constantly for this exact purpose. ♦

4.3.3 Measuring Qubits

Recall: Suppose the single qubit system is in state $|x\rangle = a|0\rangle + b|1\rangle$, find the probability the state is $|0\rangle$ (+1 outcome) or $|1\rangle$ (-1 outcome) upon measurement of σ_z

Idea: Let's try and use the spectral projection associated to $\sigma_z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$:

$$\begin{aligned} &= 1 \underbrace{|0\rangle\langle 0|}_{M_0} + (-1) \underbrace{|1\rangle\langle 1|}_{M_1} \\ p(0) &= a^2 = \langle x | M_0 | x \rangle \\ p(1) &= b^2 = \langle x | M_1 | x \rangle \end{aligned} \tag{4.65}$$