

# Mathematical Tools for Quantum Computing



CPSC 4470/5470

Introduction to Quantum Computing

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# Mathematical Model of Quantum Systems

**Four Principles** to model quantum systems mathematically:

**1. Superposition:**

The state of a qubit is a unit complex vector in the two-dimensional *Hilbert Space*.

**2. Composition:**

The joint state of many (independent) quantum systems is the *tensor product* of component states.

**3. Transformation:**

Time evolution of a quantum system is a *unitary process*.

**4. Measurement:**

Measuring a quantum state causes its superposition to *collapse/project* to one of its basis states randomly.

**Von Neumann:** “*In mathematics, you don’t understand things. You just get used to them.*”

# Superposition State of A Qubit


The **state of a qubit** is represented by a unit vector in the two-dimensional complex vector space (**Hilbert Space**  $\mathbb{C}^2$ ). In the Dirac notation:


$$|\psi\rangle = \alpha|0\rangle + \beta|1\rangle = \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$$


where  $\alpha, \beta \in \mathbb{C}$ , satisfying that its 2-norm:  $|\alpha|^2 + |\beta|^2 = 1$ .

Here,  $\alpha$  and  $\beta$  are called the **probability amplitudes** on the (classical) basis,  $|0\rangle$  and  $|1\rangle$ , respectively.

*(can be negative, even complex numbers)*


$$|\psi\rangle = 0.6|0\rangle - 0.8|1\rangle = \begin{bmatrix} 0.6 \\ -0.8 \end{bmatrix}$$


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


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

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

Not valid:  $\frac{1}{4}|0\rangle + \frac{i}{2}|1\rangle = \begin{bmatrix} 1/4 \\ i/2 \end{bmatrix}$

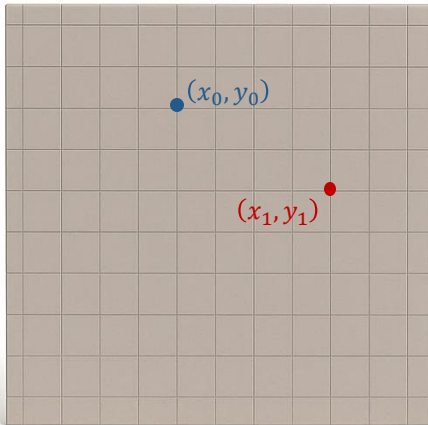
# Vector Spaces

**Vector space:** a (special) set of vectors.

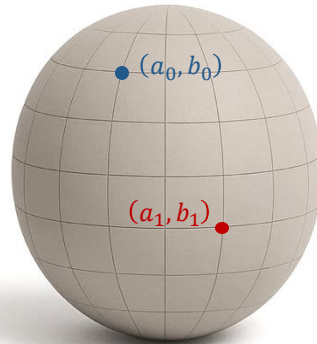
**Example:** Position vectors for points on a 2D surface.

$\mathbb{R}^2$ : Two-dimensional real vector space

Position on a flat surface: X-y plane



Surface of a sphere: Longitude-latitude



Two real-valued coordinates.

**Algebra:** that allows us to take **linear combinations**.

- Scalar multiplication:  $c \cdot \vec{a} = \begin{bmatrix} ca_0 \\ ca_1 \end{bmatrix}$  for some  $c \in \mathbb{R}$
- Vector addition:  $\vec{a} + \vec{b} = \begin{bmatrix} a_0 + b_0 \\ a_1 + b_1 \end{bmatrix}$

“Vectors after algebra remain in the same space.”

$\mathbb{R}^n$ : n-dimensional real vector space:

all vectors with n real components:  $\vec{a} = \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix} \in \mathbb{R}^n$

# More about Vectors

**Properties** (derive on board):

- **Length** of a vector (as in  $\ell_2$ -norm or Euclidean norm):  $\vec{a} = \begin{bmatrix} a_0 \\ a_1 \end{bmatrix}$

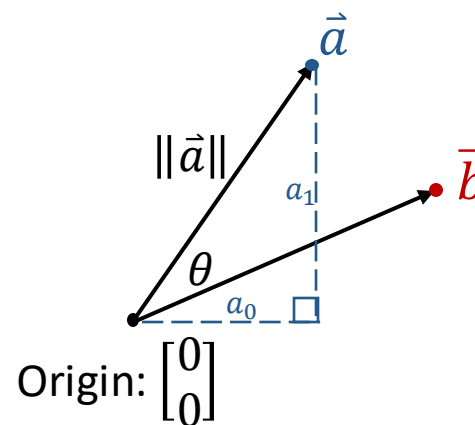
$$\|\vec{a}\|_2^2 = \vec{a}^T \vec{a}$$

$\vec{a}^T$ : Transpose of  $\vec{a}$

- **Angle** between two vectors:  $\vec{a} = \begin{bmatrix} a_0 \\ a_1 \end{bmatrix}$  and  $\vec{b} = \begin{bmatrix} b_0 \\ b_1 \end{bmatrix}$

$$\|\vec{a}\| \cdot \|\vec{b}\| \cos \theta = \vec{a}^T \vec{b}$$

**Inner product** of  $\vec{a}$  and  $\vec{b}$ :  $\vec{a}^T \vec{b}$



**Hilbert space:** inner-product vector space.

- Well-defined “geometry” (incl. length and angle).

Do they work for **complex vectors**?

# What about Complex Vectors?

A **complex number**  $\alpha \in \mathbb{C}$  is of form:

$$\alpha = \underbrace{a_0}_{\text{Real component}} + \underbrace{a_1 i}_{\text{Imaginary component}}$$

Where  $a_0, a_1 \in \mathbb{R}$  and  $i^2 = -1$ .

A vector in a two-dimensional complex vector space:

$$|\psi\rangle = \begin{bmatrix} \alpha_0 \\ \alpha_1 \end{bmatrix} \in \mathbb{C}^2$$

**Can we use the same definition for length and angle?**

- Length:

$$\vec{a}^T \vec{a} = \|\vec{a}\|_2^2$$

- Angle:

$$\vec{a}^T \vec{b} = \|\vec{a}\| \cdot \|\vec{b}\| \cos \theta$$

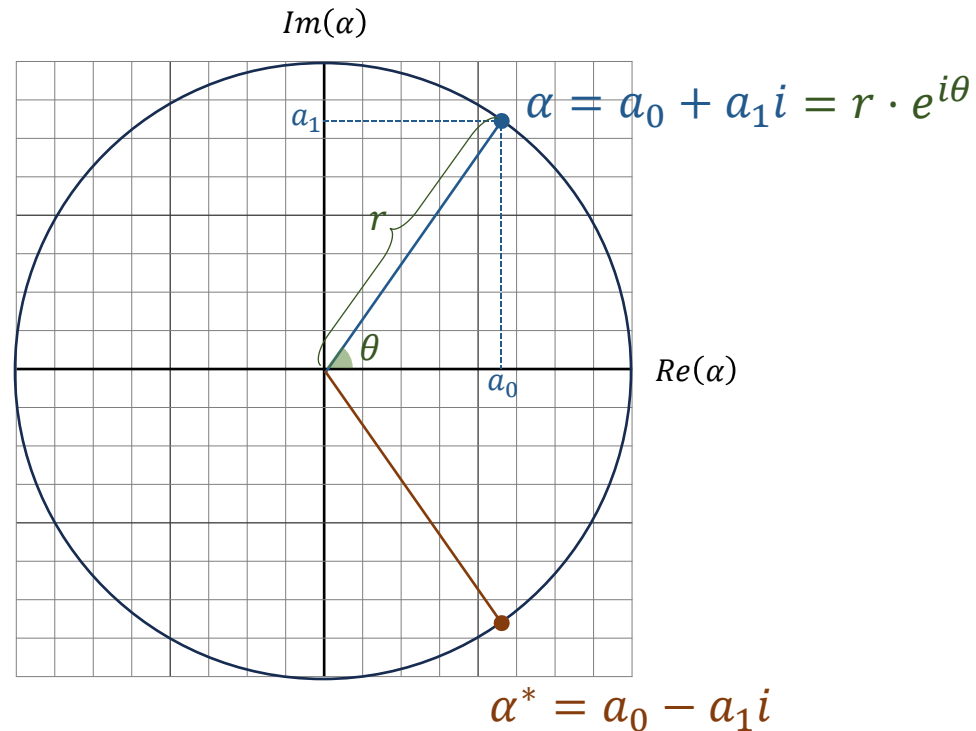
What could be problematic?

Example: Length of  $|\psi\rangle = \begin{bmatrix} 1 \\ i \end{bmatrix}$ ?

# Complex Numbers

A **complex number**  $\alpha \in \mathbb{C}$  is of form:  $\alpha = a_0 + a_1 i$ , where  $a_0, a_1 \in \mathbb{R}$  and  $i^2 = -1$ .

**Complex Plane:**



**Cartesian coordinate**  $(a_0, a_1)$ : (real, imaginary)  
 $\alpha = a_0 + a_1 i$

**Euler's formula**  $e^{i\theta} = \cos \theta + i \sin \theta$

**Polar coordinate**  $(r, \theta)$ : (magnitude, phase)  
 $\alpha = r \cos \theta + r \sin \theta i = r \cdot e^{i\theta}$  "length" "angle"

"Length": Magnitude  $r = |\alpha| = \sqrt{a_0^2 + a_1^2} = \sqrt{\alpha^* \alpha}$

"Angle": Complex phase  $\theta = \arctan \frac{a_1}{a_0}$

**Notation:**  $\alpha^*$  is the **complex conjugate** of  $\alpha$   
 $\alpha^* = a_0 - a_1 i = r \cdot e^{-i\theta}$

# Back to Complex Vectors

For a **quantum state**  $|\psi\rangle = \begin{bmatrix} \alpha_0 \\ \alpha_1 \end{bmatrix}$  :  
What are  $\langle\psi|\psi\rangle$  ,  $\langle 0|\psi\rangle$  and  $\langle 1|\psi\rangle$ ?

## Algebra:

- Scalar multiplication:  $c \cdot |\psi\rangle = \begin{bmatrix} c\alpha_0 \\ c\alpha_1 \end{bmatrix}$  for some  $c \in \mathbb{C}$
- Vector addition:  $|\psi\rangle + |\varphi\rangle = \begin{bmatrix} \alpha_0 + \beta_0 \\ \alpha_1 + \beta_1 \end{bmatrix}$

## Properties (derive on board):

- Length of a vector:  $|\psi\rangle = \begin{bmatrix} \alpha_0 \\ \alpha_1 \end{bmatrix}$

$$\| |\psi\rangle \|_2^2 = [\alpha_0^* \quad \alpha_1^*] \cdot \begin{bmatrix} \alpha_0 \\ \alpha_1 \end{bmatrix}$$

Example: Length of  $|\psi\rangle = \begin{bmatrix} 1 \\ i \end{bmatrix}$

- Angle between two vectors:  $|\psi\rangle = \begin{bmatrix} \alpha_0 \\ \alpha_1 \end{bmatrix}$  and  $|\varphi\rangle = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$

$$\cos \theta = [\alpha_0^* \quad \alpha_1^*] \cdot \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix}$$

“Ket”

## Dirac Notation:

$$\langle\psi| = (|\psi\rangle)^\dagger = (|\psi\rangle^*)^T = [\alpha_0^* \quad \alpha_1^*]$$

↑  
“Bra”

Adjoint of a vector: conjugate transpose

$$\langle\psi| \cdot |\psi\rangle = \langle\psi|\psi\rangle$$

Inner product: “Bra-ket”

$$\langle\psi| \cdot |\varphi\rangle = \langle\psi|\varphi\rangle$$



# What about more qubits?

Two coins:



$$\Pr(0_A) = 0.36, \Pr(1_A) = 0.64 \quad \Pr(0_B) = 0.5, \Pr(1_B) = 0.5$$

Assuming they are independent:  $\Pr(0_A 0_B) = \Pr(0_A) \cdot \Pr(0_B)$

$$\begin{aligned} \Pr(0_A 0_B) &= 0.36 \cdot 0.5, & \Pr(0_A 1_B) &= 0.36 \cdot 0.5, \\ \Pr(1_A 0_B) &= 0.64 \cdot 0.5, & \Pr(1_A 1_B) &= 0.64 \cdot 0.5, \end{aligned}$$

We use a “joint state” to fully characterize the system:

Tensor product:

$$\begin{bmatrix} 0.36 \\ 0.64 \end{bmatrix} \otimes \begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix} = \begin{bmatrix} 0.36 \cdot \begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix} \\ 0.64 \cdot \begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 0.36 \cdot 0.5 \\ 0.36 \cdot 0.5 \\ 0.64 \cdot 0.5 \\ 0.64 \cdot 0.5 \end{bmatrix} \begin{matrix} \Pr(0_A 0_B) \\ \Pr(0_A 1_B) \\ \Pr(1_A 0_B) \\ \Pr(1_A 1_B) \end{matrix}$$

Two qubits:



$$|\psi_A\rangle = 0.6|0\rangle + 0.8|1\rangle$$

$$|\psi_B\rangle = \frac{1}{\sqrt{2}}|0\rangle - \frac{1}{\sqrt{2}}|1\rangle$$

Assuming they are independent:  $\langle 0_A 0_B | \psi_{AB} \rangle = \langle 0_A | \psi_A \rangle \cdot \langle 0_B | \psi_B \rangle$

$$|\psi_{AB}\rangle = 0.6 \cdot \frac{1}{\sqrt{2}}|00\rangle - 0.6 \cdot \frac{1}{\sqrt{2}}|01\rangle + 0.8 \cdot \frac{1}{\sqrt{2}}|10\rangle - 0.8 \cdot \frac{1}{\sqrt{2}}|11\rangle$$

$$\begin{bmatrix} 0.6 \\ 0.8 \end{bmatrix} \otimes \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 0.6 \cdot 1/\sqrt{2} \\ -0.6 \cdot 1/\sqrt{2} \\ 0.8 \cdot 1/\sqrt{2} \\ -0.8 \cdot 1/\sqrt{2} \end{bmatrix} \begin{matrix} |0_A 0_B\rangle \\ |0_A 1_B\rangle \\ |1_A 0_B\rangle \\ |1_A 1_B\rangle \end{matrix}$$

# High-Dimensional Complex Hilbert Space

The **joint state of n qubit** is a  $2^n$ -dimensional complex vector in the **Hilbert Space**  $\mathcal{H}$ , described by complex numbers,  $\alpha_i \in \mathbb{C}$ , for  $i \in \{0,1\}^n$ , satisfying that its 2-norm:  $\sum_{i \in \{0,1\}^n} |\alpha_i|^2 = 1$ . In the Dirac notation:

$$|\psi\rangle = \alpha_{00\dots 0}|00 \dots 0\rangle + \alpha_{00\dots 1}|00 \dots 1\rangle + \dots + \alpha_{11\dots 1}|11 \dots 1\rangle = \begin{bmatrix} \alpha_{00\dots 0} \\ \alpha_{00\dots 1} \\ \vdots \\ \alpha_{11\dots 1} \end{bmatrix} \in \mathcal{H} = \mathbb{C}^{2^n}$$

$$\text{Inner product: } \langle\psi|\psi\rangle = [\alpha_{00\dots 0}^* \quad \alpha_{00\dots 0}^* \quad \dots \quad \alpha_{00\dots 0}^*] \begin{bmatrix} \alpha_{00\dots 0} \\ \alpha_{00\dots 1} \\ \vdots \\ \alpha_{11\dots 1} \end{bmatrix} = \sum_{i \in \{0,1\}^n} \alpha_i^* \alpha_i = 1 \quad \text{Normalized.} \quad \checkmark$$

This is a linear combination over  $2^n$  **basis states**:  $|00 \dots 0\rangle, |00 \dots 1\rangle, \dots, |11 \dots 1\rangle$ .

# Change of Basis?



$$|\psi\rangle = 0.6|0\rangle + 0.8|1\rangle$$

Standard basis: 
$$\begin{cases} |0\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ |1\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \end{cases}$$

Define an alternative basis:

$$\begin{cases} |+\rangle = \frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle \\ |-\rangle = \frac{1}{\sqrt{2}}|0\rangle - \frac{1}{\sqrt{2}}|1\rangle \end{cases}$$

Derive on board:

$$|\psi\rangle = ?|+\rangle + ?|-\rangle$$

What about the following basis for  $|\psi\rangle$ ?

$$\begin{cases} |v\rangle = |0\rangle + i|1\rangle \\ |w\rangle = |0\rangle - i|1\rangle \end{cases}$$

What are the criteria for a “good” basis?

# Orthonormal Basis

**Span:** A set of vectors  $|v_0\rangle, |v_1\rangle, \dots, |v_{n-1}\rangle$  *spans* the vector space  $S$ , if for any vector  $|w\rangle \in S$ , there exists  $\alpha_0, \dots, \alpha_{n-1} \in \mathbb{C}$ :

$$|w\rangle = \underbrace{\alpha_0|v_0\rangle + \alpha_1|v_1\rangle + \dots + \alpha_{n-1}|v_{n-1}\rangle}_{\text{Linear combination (with complex coefficients)}}$$

**Linear dependence:** A set of (non-zero) vectors are *linearly dependent* if there exists  $\alpha_0, \dots, \alpha_{n-1}$  not all zero:

$$0 = \alpha_0|v_0\rangle + \alpha_1|v_1\rangle + \dots + \alpha_{n-1}|v_{n-1}\rangle$$

**Basis:** *linearly independent* vectors *spans* the vector space  $S$ .

**Orthonormal basis:**

- Length: A set of unit vectors
- Angle: mutually orthogonal

$$\langle v_j | v_k \rangle = \begin{cases} 1, & \text{if } j = k \text{ (unit length)} \\ 0, & \text{if } j \neq k \text{ (orthogonal)} \end{cases}$$

**Examples:** two-qubit basis

$$|00\rangle = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, |01\rangle = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, |10\rangle = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}, |11\rangle = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$|+\rangle \otimes |+\rangle \rightarrow |++\rangle = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, |+-\rangle = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix}, |-+\rangle = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ -1 \\ -1 \end{bmatrix}, |--\rangle = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}$$

# The Process of Evolving a Probabilistic State

A **Markov Process** is a stochastic process over a state space  $S$  such that:

$$\Pr[S_{t+1} = s' | S_t = s, S_{t-1} = s_{t-1}, \dots, S_0 = s_0] = \Pr[S_{t+1} = s' | S_t = s]$$

For all state  $s_t \in S$  at time  $t$ .

**Transition probability** (from  $s$  to  $s'$ ):

$$\Pr[S_{t+1} = s' | S_t = s]$$

**Example:**

**A 3-bit Markov Model** for CPSC 4470 student:

- Ⓟ Phone out? (0 = away, 1 = out)
- ⓓ Discussing? (0 = quiet, 1 = participating)
- Ⓢ Understanding now? (0 = lost, 1 = get)

**Eight possible states ( $s \in S$ ):**

- 0 ⓅⓓⓈ Silently lost
- 1 ⓅⓓⓈ Zen absorber
- 2 ⓅⓓⓈ Eager-but-lost
- 3 ⓅⓓⓈ Classroom king
- 4 ⓅⓓⓈ Doomscrolling
- 5 ⓅⓓⓈ Overconfident texter
- 6 ⓅⓓⓈ Clueless chaos
- 7 ⓅⓓⓈ Mythical Multitasker

A **transition matrix** to fully characterize the process.

|    |     | FROM |     |     |     |      |      |      |     |
|----|-----|------|-----|-----|-----|------|------|------|-----|
|    |     | 000  | 001 | 010 | 011 | 100  | 101  | 110  | 111 |
| TO | 000 | 0.5  | 0.2 |     |     | 0.25 |      |      |     |
|    | 001 | 0.3  | 0.5 |     | 0.3 |      | 0.25 |      |     |
|    | 010 | 0.2  |     | 0.5 | 0.2 |      |      | 0.25 |     |
|    | 011 |      | 0.3 | 0.5 | 0.5 |      |      |      | 0.3 |
|    | 100 |      |     |     |     | 0.5  |      |      |     |
|    | 101 |      |     |     |     |      | 0.5  |      | 0.2 |
|    | 110 |      |     |     |     | 0.15 |      | 0.5  |     |
|    | 111 |      |     |     |     | 0.1  | 0.25 | 0.25 | 0.5 |

**Col-k, Row-j of the matrix:**

$$T[k, j] = \Pr[S_{t+1} = k | S_t = j]$$

- Non-negative elements.
- Columns sum to 1.

# The Process of Evolving a Probabilistic State

Probabilistic mixture of possible states:

- 0 (P) (D) (U) Silently lost
- 1 (P) (D) (U) Zen absorber
- 2 (P) (D) (U) Eager-but-lost
- 3 (P) (D) (U) Classroom king
- 4 (P) (D) (U) Doomscrolling
- 5 (P) (D) (U) Overconfident texter
- 6 (P) (D) (U) Clueless chaos
- 7 (P) (D) (U) Mythical Multitasker

$$\vec{p}_t = \begin{bmatrix} \Pr[S_t = 0] \\ \Pr[S_t = 1] \\ \Pr[S_t = 2] \\ \Pr[S_t = 3] \\ \Pr[S_t = 4] \\ \Pr[S_t = 5] \\ \Pr[S_t = 6] \\ \Pr[S_t = 7] \end{bmatrix}$$

Evolution of probabilistic state:

$$\vec{p}_{t+1} = T \cdot \vec{p}_t$$

Transition matrix

Three stages of the class:

Output  
state

$T_0$ : "Staying engaged"

Input  
state

$$\begin{bmatrix} 0.2 \\ 0.5 \\ 0.3 \end{bmatrix} = \begin{bmatrix} 0.5 & 0.2 & & & 0.25 \\ 0.3 & 0.5 & & 0.3 & & 0.25 \\ 0.2 & & 0.5 & 0.2 & & 0.25 \\ & 0.3 & 0.5 & 0.5 & & \\ & & & & 0.5 & \\ & & & & 0.15 & 0.5 \\ & & & & 0.1 & 0.25 & 0.25 & 0.5 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$T_1$ : "Drifting"

$$\begin{bmatrix} 0.22 \\ 0.46 \\ 0.1 \\ 0.17 \\ 0.05 \end{bmatrix} = \begin{bmatrix} 0.6 & 0.2 & 0.2 & & 0.2 \\ 0.2 & 0.6 & & 0.4 & \\ 0.2 & & 0.6 & 0.2 & \\ & 0.1 & 0.2 & 0.4 & \\ & 0.1 & & & 0.7 & 0.2 \\ & & & & 0.4 & 0.2 \\ & & & & 0.1 & 0.6 & 0.2 \\ & & & & 0.2 & 0.2 & 0.4 \end{bmatrix} \cdot \begin{bmatrix} 0.2 \\ 0.5 \\ 0 \\ 0.3 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$T_2$ : "Fatigue + bounce back"

$$\begin{bmatrix} 0.1925 \\ 0.4475 \\ 0.128 \\ 0.207 \\ 0.02 \\ 0.005 \end{bmatrix} = \begin{bmatrix} 0.4 & 0.2 & & & 0.25 \\ 0.4 & 0.6 & 0.2 & 0.3 & 0.25 & 0.25 \\ 0.2 & & 0.5 & 0.2 & & 0.1 \\ & 0.2 & 0.3 & 0.5 & & 0.3 \\ & & & & 0.4 & 0.15 \\ & & & & 0.1 & 0.5 & 0.2 \\ & & & & & 0.5 & \\ & & & & & 0.25 & 0.25 & 0.5 \end{bmatrix} \cdot \begin{bmatrix} 0.22 \\ 0.46 \\ 0.1 \\ 0.17 \\ 0.05 \end{bmatrix}$$

# The Process of Evolving a Superposition State

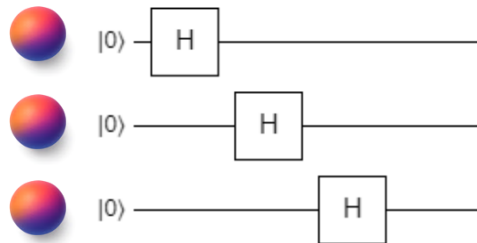
Superposition of possible states:

$$|\psi_t\rangle = \begin{bmatrix} \alpha_{00\dots0} \\ \alpha_{00\dots1} \\ \vdots \\ \alpha_{11\dots1} \end{bmatrix}$$

Evolution of superposition state:

$$|\psi_{t+1}\rangle = U \cdot |\psi_t\rangle$$

Unitary matrix



Apply three quantum gates:

Output state

$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$U_0 = H \otimes I \otimes I$$

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

Input state

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$U_1 = I \otimes H \otimes I$$

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

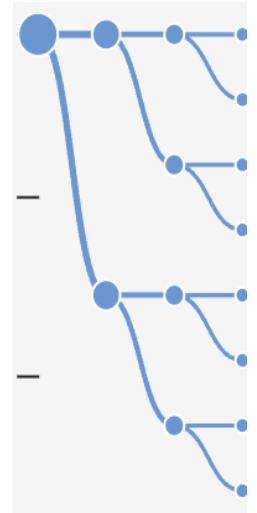
$$\cdot \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\frac{1}{2\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$U_2 = I \otimes I \otimes H$$

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

$$\cdot \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$



# Geometry-Preserving Transformations

**Inner product** gives the length and angle of vectors:

Real vectors:  $\vec{v}_j^T \vec{v}_k$

Complex vectors:  $\langle v_j | v_k \rangle$

Transformations that preserves “geometry”: lengths and angles.

- Orthonormal basis stays orthonormal.
- Shapes don’t get stretched.

|            | Orthogonal Matrix        | Unitary Matrix              |
|------------|--------------------------|-----------------------------|
| Definition | $Q^T Q = I$              | $U^\dagger U = I$           |
| Inverse    | $Q^{-1} = Q^T$           | $U^{-1} = U^\dagger$        |
| Columns    | Real orthonormal vectors | Complex orthonormal vectors |
| Example    | 2D rotation matrix       | Quantum gates               |

**Principle #3: Transformation**

(More details in Lecture 5.)



# Understanding Projections

Derive on board:

$$\Pi_0 = \frac{|0\rangle\langle 0|}{\langle 0|0\rangle}, \text{ What is } \Pi_0|\psi\rangle?$$

## Principle #4: Measurements

(More details in Lecture 6.)

**Projection** (as a “linear operator”: mapping from one vector space to another)

Example:

3D object casting shadows onto 2D plane.

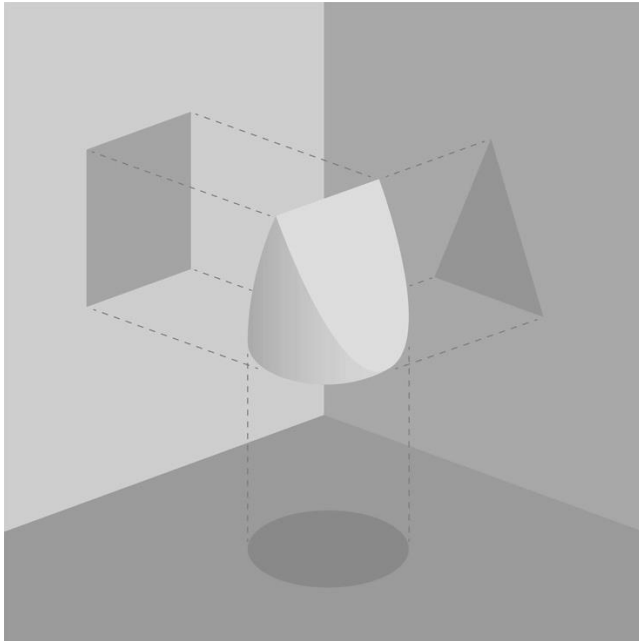
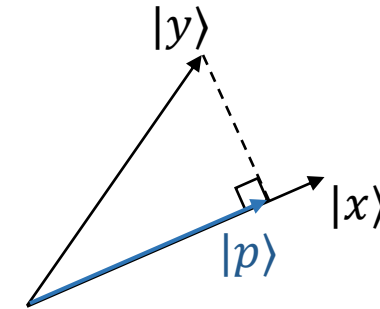


Image credit: Peter Hermes Furian / Alamy Stock

Projecting  $|y\rangle$  onto the line direction of  $|x\rangle$ .



Projected state:  $|p\rangle = ?$

**Projection is a matrix:**

$$|p\rangle = \Pi_x |y\rangle, \Pi_x = \frac{|x\rangle\langle x|}{\langle x|x\rangle}$$

Outer product: a matrix

Inner product: a scalar

**Properties:**

1.  $\Pi$  is symmetric. (Hermitian:  $\Pi^\dagger = \Pi$ )
2. Project twice is the same as once. ( $\Pi^2 = \Pi$ )

# Hermitian Matrix in QM

**Symmetric matrix** (real):

$$S^T = S$$

**Hermitian matrix** (complex):

$$H^\dagger = H$$

- Hermitian matrix has **real eigenvalues**.
- Corresponding to *physical observable* with real-valued quantity.

Why?

Eigenvalue equation for a linear operator  $A$ :

$$A|v_j\rangle = \lambda_j|v_j\rangle$$

where  $|v_j\rangle$  is the (non-zero) **eigenvector**,  
and  $\lambda_j$  is a complex number known as the **eigenvalue**.

Derive on board (Pauli Matrices):

“Pauli Z operator”  $\sigma_Z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$

Standard basis!  
 $\{|0\rangle, |1\rangle\}$

- Eigenvalues:  $\lambda_0 = 1$  and  $\lambda_1 = -1$
- Eigenvectors:  $|v_0\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $|v_1\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

“Pauli Y operator”  $\sigma_Y = \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix}$

$\{|+i\rangle, |-i\rangle\}$  basis!

- Eigenvalues:  $\lambda_0 = 1$  and  $\lambda_1 = -1$
- Eigenvectors:  $|v_0\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ i \end{bmatrix}$  and  $|v_1\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -i \end{bmatrix}$

“Pauli X operator”  $\sigma_X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

$\{|+\rangle, |-\rangle\}$  basis!

- Eigenvalues:  $\lambda_0 = 1$  and  $\lambda_1 = -1$
- Eigenvectors:  $|v_0\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $|v_1\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

# Spectral Theorem

For a linear operator that is normal ( $A^\dagger A = A A^\dagger$ ), we can write it in the **spectral decomposition**:

$$A = \sum_j \lambda_j |v_j\rangle\langle v_j|$$

where  $\lambda_j$  are the eigenvalues, and  $|v_j\rangle$  are the corresponding (orthonormal) eigenvectors.

**Examples:**

$$\sigma_Z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} = (+1)|0\rangle\langle 0| + (-1)|1\rangle\langle 1| \quad \sigma_X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = (+1)|+\rangle\langle +| + (-1)|-\rangle\langle -|$$

**Applications:**

- Power of a matrix:

$$A^8 = \left( \sum_j \lambda_j |v_j\rangle\langle v_j| \right)^8 = \sum_j \lambda_j^8 |v_j\rangle\langle v_j|$$

- Exponential of a matrix:

$$e^A \equiv \sum_{k=0}^{\infty} \frac{1}{k!} A^k = \sum_j e^{\lambda_j} |v_j\rangle\langle v_j|$$

**Example:**

$$e^{i\theta\sigma_Z} = e^{i\theta}|0\rangle\langle 0| + e^{-i\theta}|1\rangle\langle 1| = \begin{bmatrix} e^{i\theta} & 0 \\ 0 & e^{-i\theta} \end{bmatrix}$$