

Mathematics for Quantum Computing

$$2x + (y - x) = 2x + (y + (-x)) \quad \text{Notation} \rightarrow \begin{array}{l} \text{Additive} \\ \text{Inverse} \end{array}$$

↳ $2/(-2) = -2$
 Matter Anti-matter
 $2 + (-2) = 0 \rightarrow \text{Additive identity}$

$$a + b = 2x + ((-x) + y) \quad \text{Commutativity of group law}$$

$$= (2x + (-x)) + y \quad \text{Associativity}$$

$$= (2x + (-1)x) + y \quad \text{Notation}$$

$$= (2 + (-1))x + y \quad \text{Distributive property}$$

$$= 1x + y \quad \text{Closure}$$

$$x \in \mathbb{Z} \quad \text{A} \subseteq \mathbb{Z}$$

$$1x \in \mathbb{Z} \quad \text{B} \subseteq \mathbb{Z}$$

$$= x + y \quad \text{Multiplicative identity}$$

$$(c+d) \cdot e = c \cdot (d \cdot e) \quad \text{Associativity}$$

$$a \square e = a \quad \text{Identity of}$$

$$\text{that operation } = 0 \cdot 1 = 1 \cdot 0 \quad \text{closure property}$$

Algebraic Structure

$(R, *, *_1, *_2, *_3, \dots)$ \rightarrow Closure property (by default)
 Non-empty \downarrow Binary operations \rightarrow $a * b \in R$

$$a, b \in R$$

$$a * b \in R$$

A non-empty set, and an operation defined on that set, and that operation satisfies closure property, it is called a groupoid. And if it also satisfies associativity, it is called semi-group.

If the semigroup also has an identity element as well, it's called a Monoid.

If ~~the~~ each element also has an inverse, it's called a Group.

If the group is also commutative, it's called an Abelian Group.

- $(R, +, \cdot) \Rightarrow$ A set equipped with two operations; $+, \cdot$
- ① Closure under addition
 $a, b \in R$
 $a+b \in R$
 - ② Associativity under addition
 $a, b, c \in R$
 $(a+b)+c = a+(b+c)$
 - ③ Identity under addition
 $\exists 0 \in R$ such that
 $a+0 = a = 0+a$
 - ④ Additive inverse
 $a \in R \exists -a \in R$
such that,
 $a+(-a) = 0 = (-a)+a$
 - ⑤ Commutativity under Addition
 $a, b \in R$ such that
 $a+b = b+a$
 - ⑥ Closure under multiplication
 $a, b \in R$
 $\Rightarrow ab \in R$
 - ⑦ Associativity under multiplication
 $a, b, c \in R$
 $(a \cdot b) \cdot c = a \cdot (b \cdot c)$
 - ⑧ Identity under Multiplication
 $a \in R \exists 1 \in R$ such that
 $a \cdot 1 = 1 \cdot a = a$
 - ⑨ Multiplicative inverse
 $a \in R \exists a^{-1} \in R$ such that
 $a \cdot (a^{-1}) = a^{-1} \cdot a = 1$
 - ⑩ Commutativity under Multiplication
 $a, b \in R$
 $a \cdot b = b \cdot a$
 - ⑪ Distributive property
 $(a+b) \cdot c = ac + bc$
 $a \cdot (b+c) = ab + ac$
 - ⑫ $0 \neq 1$
(This is just to exclude some trivial cases)
(Identity element must be unique)

If $(R, +, \cdot)$ satisfies is an Abelian group under addition and has closure and associative multiplication and satisfies the distributive property, this algebraic structure is called a RING.

If a ring also satisfies the identity property under multiplication, it is called Ring with Unity.

If a Ring with unity also satisfies commutativity under Multiplication, it is called Commutative Ring with Unity.

An Abelian group under Addition and Multiplication (satisfying all 10 Rules) + satisfies Distributive prop. + ② Rule

$\boxed{V(F)}$ Field
↓
Vector Space

$F \rightarrow \mathbb{R}^*$, \mathbb{C}^* , \mathbb{Q}^*
↓
 $(\mathbb{R} - \{0\})$ $(\mathbb{C} - \{0\})$ $(\mathbb{Q} - \{0\})$

Note: There are ∞ Extension Fields of \mathbb{Q}
★ $\mathbb{Q}(\alpha)$ is an Extension Field of $\mathbb{Q} \Rightarrow$ ALGEBRAIC EXTENSION

Extending \mathbb{Q} :

$\sqrt{2}$ is not a rational no.

A Field containing both \mathbb{Q} and $\sqrt{2}$:

$\mathbb{Q}(\alpha) \rightarrow$ Extended Field of \mathbb{Q} .

Any irrational no.

$\mathbb{Q}(\sqrt{2})$
+ Abelian
• Abelian
Distributive & $0 \neq 1$
Extension Field of \mathbb{Q}

Linear Transformation and Vector Spaces

$$\begin{array}{l} x+y=6 \\ 2x+y=3 \\ \hline -x=3 \end{array}$$

Unique soln.

$$\begin{array}{l} x+y=6 \\ x+y=4 \\ \hline \text{No soln.} \end{array}$$

Two parallel lines

$$\begin{array}{l} x+y=6 \\ 2x+2y=12 \\ \hline \text{Infinite solns.} \end{array}$$

Lines coincide

(x, y)

(0, 0)

(0, 1)

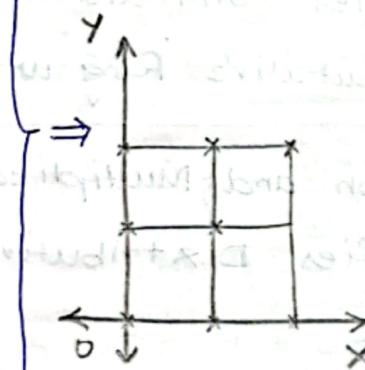
(1, 0)

(1, 1)

(2, 0)

(0, 2)

(2, 2)



Let's say: $x+y=x$ (x, y)

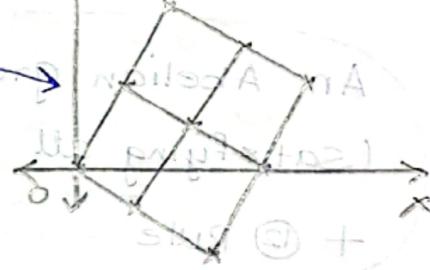
$$2x+y=4$$

(0, 0)

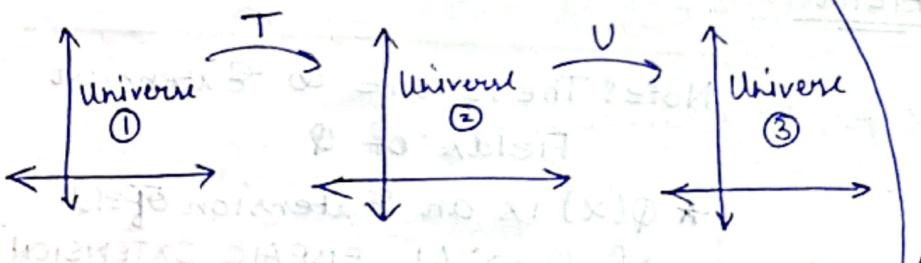
(1, 1)

(1, 2)

Graphs are drawn just for understanding and might not be actual graphs.



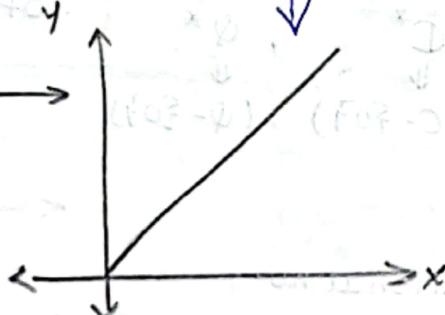
The graph that was a straight square has now become a tilted rectangle
⇒ TRANSFORMATION



$$\begin{array}{l} x+2y=x \\ 2x+4y=y \end{array}$$

$$\boxed{2(x+2y)=y}$$

$$2x=y$$



Notation:

Cayley

$$\begin{pmatrix} 1 & 1 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 6 \\ 4 \end{pmatrix}$$

"Matrix Notation"

$$x+2y+3=0$$

$$x-3y+3=3$$

$$2x-y+5=5$$

$$\boxed{\begin{pmatrix} 1 & 2 & 1 \\ 1 & -3 & 1 \\ 2 & -1 & 1 \end{pmatrix}} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 5 \end{pmatrix}$$

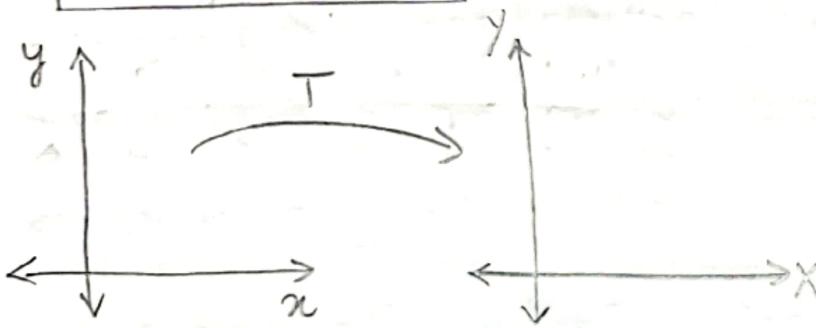
Responsible for transformation

$$x+y = x$$

$$2x-y = y$$

$$T(x, y) = (x, y)$$

$$(or) T(x, y) = (x+y, 2x-y)$$



Shifting from
this plane to ...

this plane

$$G(x, y) = (x, -y) = (x, y)$$

$$H(x, y) = (y, -x) = (x', y')$$

$$\text{If } G(x, y) = (x, -y)$$

$$x = x + 0y$$

$$y = 0x - 1y$$

$$\Rightarrow G = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$\text{If } H(x, y) = (y, -x)$$

$$x' = 0x + 1y$$

$$y' = -1x + 0y$$

$$\Rightarrow H = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

$$G(x', y') = (x', -y')$$

$$= (y, -(-x))$$

$$= (y, x)$$

$$\Rightarrow GH(x, y) = (y, x)$$

$$\underline{\underline{\text{Ex:- } GH(1, 2)}}$$

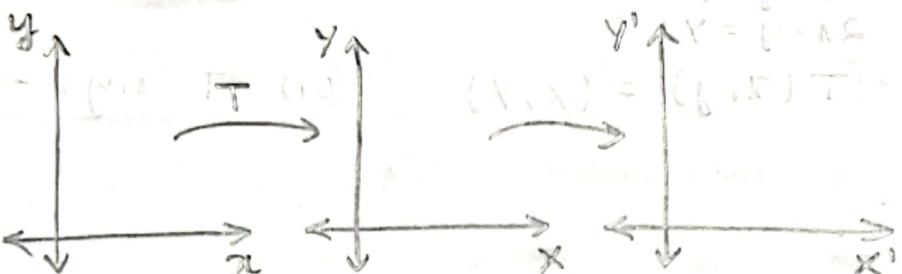
$$\Rightarrow H(1, 2) = (2, -1)$$

$$\Rightarrow G(2, -1) = (2, 1)$$

$$\begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

$$x = ax + by$$

$$y = cx + dy$$



$$x' = Ax + By$$

$$y' = Cx + Dy$$

$$\Rightarrow x' = A(ax+by) + B(cx+dy)$$

$$= Aax + Abay + Bcx + Bdny$$

$$x' = (Aa+Bc)x + (Ab+Bd)y$$

$$\Rightarrow y' = C(ax+by) + D(cx+dy)$$

$$= Cax + Cby + Dcx + Ddy$$

$$y' = (Ca+Dc)x + (Cb+Dd)y$$

$$\begin{pmatrix} A & B \\ C & D \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} Aa+Bc & Ab+Bd \\ Ca+Dc & Cb+Dd \end{pmatrix} \Rightarrow \text{Matrix Multiplication}$$

From the previous example of G and H transformations;

$$G(x, y) = (x, -y)$$

$$H(x, y) = (y, -x)$$

$$GH = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

$$GH = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ x \end{pmatrix}$$

Abstract Formalism:

$$(a, b) \in \mathbb{R}^2 \quad (c, d) \in \mathbb{R}^2$$

$$(a, b) + (c, d) \in \mathbb{R}^2 \quad [\text{closure}]$$

$$(a, b) + (c, d) = (a+c, b+d)$$

$\mathbb{R}^2 \rightarrow 2\text{-dimensional plane}$
(SPACE)
 $r \rightarrow \text{Scalar}$
(FIELD)
↓
no./component

$$T[(a, b) + (c, d)] = T(a, b) + T(c, d)$$

$$T(r(a, b)) = rT(a, b)$$

⇒ Linear Transformation

A space with infinite dimensions ⇒ HILBERT SPACE

Extended Fields

* a_1, a_2, a_3, \dots, L
 $\underbrace{a_1, a_2, a_3, \dots}_\text{Sequence}, L$

$L \rightarrow$ Limit of the sequence

$$\mathbb{Q} + L = \mathbb{R}$$

\downarrow
Field

* Extended \mathbb{R}

$$\mathbb{R}(i)$$

$i \rightarrow \text{Imaginary no.}$
 $= \sqrt{-1}$

$\hookrightarrow \mathbb{C}$
Field

Extension of $\mathbb{C} = \mathbb{C}$

$$\mathbb{C}(x) = \begin{cases} f(x) \\ g(x) \end{cases} \mid f, g \text{ are polynomials with coeff's in } \mathbb{C}$$

↳ Rational functions

$$\text{Ex: } \mathbb{C}(x) = \frac{ix+3}{3-ix^2}$$

$\mathbb{C}(x) \Rightarrow$ Extended \mathbb{Q}

VECTOR SPACE

A vector space is a set of vectors for which addition and scalar multiplication are defined such that if u and v are vectors in the space and k is a scalar, then $u+v$ is in the space and ku is in the space.

What is space in Mathematics?

→ A non-empty set having some mathematical structure

Ex:- $A = \{1, 2, 3\}$ \Rightarrow ^{Ordinary} Simple Set (3 elements)

$(\mathbb{R}, +) \Rightarrow$ This set follows certain theorems & definitions

\therefore This is a space

Definitions: ① Closure
② Associativity } Group
③ Identity
④ Inverse

Theorems: ① If G is a group wrt $*$, then identity element is unique.

② If G is a group wrt $*$, then inverse element is unique.

Ex:- $X = \mathbb{R}$, $d(x, y) = |x - y|$

① Let X be a non-empty set and

$d: X \times X \rightarrow \mathbb{R}$ (Basically, if any two elements satisfy these 3 conditions,

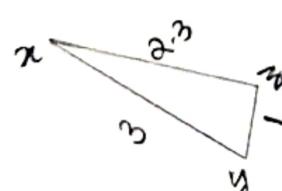
(i) $d(x, y) \geq 0$ and $d(x, y) = 0 \Leftrightarrow x = y$ they will come under metric space)

(ii) $d(x, y) = d(y, x)$

(iii) $d(x, z) \leq d(x, y) + d(y, z)$ (Triangular inequality)

Then X is metric space (\because it's equipped with a machine 'd' that calculates distance)

Mathematical Structure



Examples of vector spaces:

① $u = x+1$, $v = 2x-3$ (Polynomials of first degree)

$u+v = 3x-2 \Rightarrow$ (Polynomial of first degree)

$5u = 5(x+1) = 5x+5 \Rightarrow$ (Polynomial of first degree)

② $u = \begin{pmatrix} 2 & 1 \\ 1 & 3 \end{pmatrix}$, $v = \begin{pmatrix} -1 & 0 \\ 1 & -1 \end{pmatrix}$

③ $u = \int f(x) dx$, $v = \int g(x) dx$

④ $u = \frac{d}{dx} f(x)$, $v = \frac{d}{dx} g(x)$

⑤ $u = 1$, $v = 2$

Axioms that a vector Space must fulfill:

Let V be a vector space and F a field of scalars if

$u, v, w \in V$ and $a, b \in F$, then

1] $u+v = v+u \rightarrow$ Commutative ppty of addition

2] $(u+v)+w = u+(v+w) \rightarrow$ Associativity

3] There should be zero vector 0 , such that

$0+u = u+0 = u \rightarrow$ Additive identity

4] For every $u \in V$, $\exists u'$ such that $u+u' = 0 \rightarrow$ Additive inverse

5] $\forall u \in V, 1u = u$

6] $a \cdot (bu) = (ab) \cdot u$

7] $(a+b)u = au + bu$

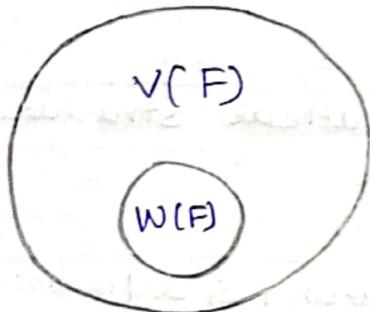
8] $a(u+v) = au + av$

9] There should be a zero vector s.t. $0u = 0$

Subspace

$V(F) \rightarrow$ Subset of $V(F) \Rightarrow W(F)$ ↓ also

Vector Space



$W(F)$ is a subspace of $V(F)$

Formal Definitions:

1] Internal Composition:

A composition $*$ is an internal composition in a non-empty set V if $\forall \alpha, \beta \in V$ there exists a unique $\alpha * \beta \in V$

2] External Composition:

$F \rightarrow$ Field

↪ Scalar (number)

$a \in F$ and $\alpha \in V$

A composition \circ is an external composition in a non-empty set V over a non-empty field F such that

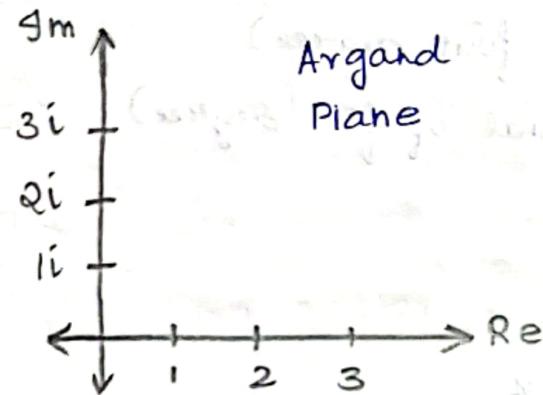
$\forall a \in F$ (and) $\alpha \in V$

$\Rightarrow a \circ \alpha \in V$

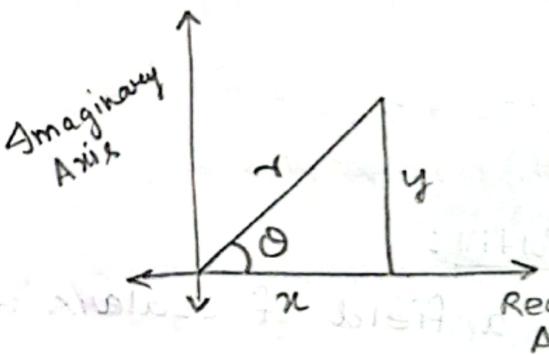
"It is also unique"

Complex Numbers:

$\sqrt{-1} = i \rightarrow$ Imaginary Number



$$z = x + iy \rightarrow \text{Complex No.}$$



$$\cos \theta = \frac{x}{r} \quad \sin \theta = \frac{y}{r}$$

$$\Rightarrow x = r \cos \theta \quad \Rightarrow y = r \sin \theta$$

Complex number $z = x + iy$

$$\Rightarrow z = r \cos \theta + iy \sin \theta$$

$$z = r(\cos \theta + i \sin \theta)$$

$$z = r e^{i\theta} \quad \text{Euler notation}$$

* In calculations, we use complex nos. but the final answer we always obtain will be a Real Number

$$z_1 = r_1 e^{i\theta_1}$$

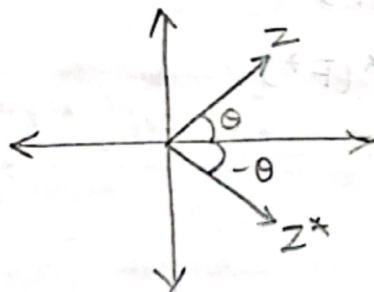
$$z_2 = r_2 e^{i\theta_2}$$

$$\begin{aligned} z_1 z_2 &= (r_1 e^{i\theta_1})(r_2 e^{i\theta_2}) \\ &= (r_1 r_2) (e^{i(\theta_1 + \theta_2)}) \end{aligned}$$

* $V(F) \rightarrow$ Complex Vector Space (\because We will take \mathbb{C} as F)

* For every complex number z , there exists a z^* that is z 's complex conjugate

If $z = x + iy$, then $z^* = x - iy$



$$zz^* = (x+iy)(x-iy)$$

$$= (re^{i\theta})(re^{-i\theta})$$

$$= r^2(e^0)$$

$$\underline{zz^* = r^2}$$

z^* → Complex No.

$z \& z^*$ → Dual no. System

\therefore A $z \exists$ a unique z^*

* Special Complex Number → PHASE FACTOR

where $r=1$

$$z = e^{i\theta}$$

$$z = \cos\theta + i\sin\theta$$

Notations: z Complex No. $\{z\}$ set of all complex nos. $\in \mathbb{C}$

z^* Complex Conjugate $\{z^*\}$ set of all complex nos. $\in \mathbb{C}$

$|\Psi\rangle$ Ket Vector

$\langle\Psi|$ Bra Vector

$\{\text{vectors}\}$

Dual No. System

$|\Psi\rangle \in V(F)$

$\langle\Psi| \in V^*(F^*)$

Dual Space / Complex Conjugate Vector Space

$\langle\phi|\Psi\rangle$ Inner product \rightarrow Number

$|\phi\rangle \otimes |\Psi\rangle$ Tensor product

A^* Complex conjugate of a matrix

A^T Transpose

A^+ Hermitian Conjugate $(A^T)^*$

Topological Space:

→ Most General Spaces

→ Set + Topology (Collection of open sets) satisfying

① Union of arbitrary open sets is open

② Intersection of finite no. of open sets is open

→ Ex: Functional Spaces, Polynomial Spaces

Vector Spaces

Set of elements
(vectors)

↓
2 Axioms

Abstract Structure

w/o any inherent
notion of vector
distance or angle

b/w vectors

Normed Spaces

Mainly used to define
length of vector
(OR)

$\|\cdot\|$ is also used to

calculate distance
b/w two vectors

(OR)

$\|\cdot\|$ assigns a non-negative number to each pair vectors

elements of V w/o any inherent notion of actual distance

Function $\|\cdot\| : V \rightarrow F$ satisfying following properties

① Non-negativity

$$|\psi| \geq 0$$

② Homogeneity

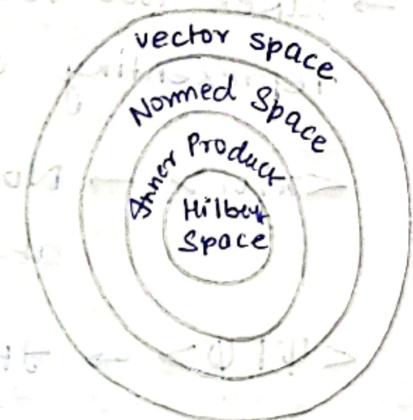
$$v \in V, \alpha \in F$$

$$\|\alpha v\| = |\alpha| \|v\|$$

③ Triangular inequality

$$|\psi + \phi| \leq |\psi| + |\phi| \quad (\text{Consider -ve vectors})$$

$$d(x, y) = \|x - y\| \rightarrow \text{Dist. b/w two vectors}$$



Inner Product Space (IPS):

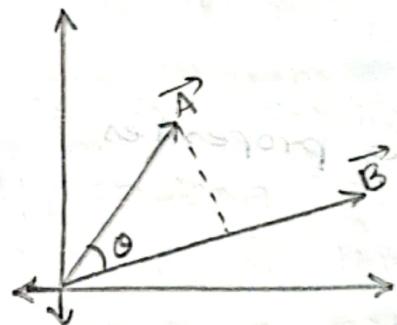
→ Vector Space + Inner Product

→ Takes two vectors and assigns a number (R or C), representing angles or correlation b/w them

$\langle x, y \rangle \rightarrow$ No / Scalar that encodes the information about their alignment or how similar they are

$\langle \psi | \phi \rangle \rightarrow$ IPS (The answer in generally will be complex)

→ "Geometrically, IP measures how much one vector points in the direction of the other vector and how their magnitudes contribute to their alignment."



$$\langle A | B \rangle = \overline{A} \cdot \overline{B} = |A| |B| \cos \theta$$

(How much \vec{A} is contributing in the direction of \vec{B})

{This is a kind of Inner Product, usually defined for use in Physics}

→ Relation between Norm and Inner Product:

$$\|x\| = \sqrt{\langle x, x \rangle}$$

$$\text{Ex:- } \overrightarrow{A} \cdot \overrightarrow{A} = |A| |A| \cos 0^\circ = |A|^2 = \|A\|^2$$

$$\Rightarrow |A| = \sqrt{A \cdot A}$$

Definition - $\langle \cdot, \cdot \rangle : V \times V \rightarrow F$ (Only +ve no.s) satisfying

① Linearity

$$\langle ax_1 + bx_2, y \rangle = a \langle x_1, y \rangle + b \langle x_2, y \rangle$$

② Conjugate Symmetry

$$\langle x, y \rangle = \overline{\langle y, x \rangle}$$

③ Positive definiteness

$$\langle x, x \rangle \geq 0 \text{ & } \langle x, x \rangle = 0 \text{ iff } x=0$$

ϵ (Extremely small no.
but $\neq 0$)

Hilbert Spaces

Complete IPS

↓
Properties of IPS + Completeness

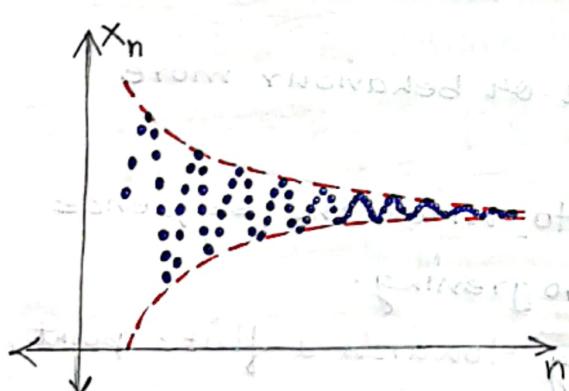
(Every Cauchy Sequence has a limit
in Hilbert Space)

Cauchy Sequence: एक दोस्ती समीकरण में एक point के
एवं consecutive elements के बीच की distance अद्यत ही नहीं हो
जाता है (निरूपित)

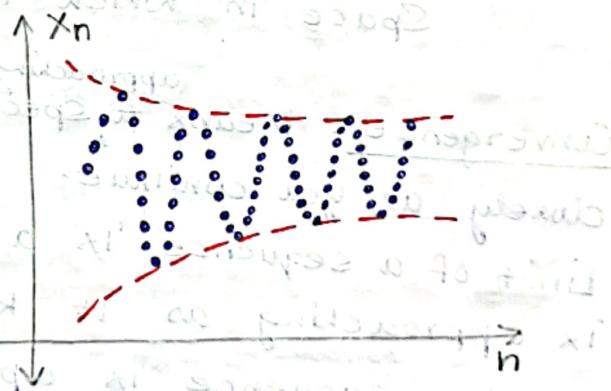
$$\left\{ \frac{1}{n} \right\} = \left\{ \frac{1}{1}, \frac{1}{2}, \frac{1}{3}, \frac{1}{4}, \dots \right\}$$

→ After a point, sequence becomes more and more
tightly clustered.

Cauchy Sequence: A sequence $\{x_n\}$ in (metric space
or Normed Space) $\forall \epsilon > 0 \exists N \in \mathbb{N}$ such that
 $|x_m - x_n| < \epsilon$



A plot of Cauchy Sequence



A sequence that's not Cauchy

Completeness: Normed Space X is complete if every
Cauchy Sequence in that space has a limit within the
Same space.

[If points of sequence become closer, the point of
convergence should also be in the same space as the
space in which the sequence was defined.]

$$\underline{\text{Ex: }} E_n = \frac{-13.6}{n^2} Z^2$$

As $n \rightarrow \infty$, $E_n \xrightarrow[]{} 0$

Limit of the E_n sequence is 0

$\Rightarrow 0$ should also be in the space
in which E_n was defined

Completeness

Ex: [Not Complete]

Let's take a sequence in Rational Normed Space ~~R~~ (\mathbb{Q}), $n \in \mathbb{N}$

$$\{x_n\} = \left\{ 1 + \frac{1}{2!} + \frac{1}{3!} + \frac{1}{4!} + \dots + \frac{1}{n!} + \dots \right\}$$

This started from rational no.s but is approaching 'e'. 'e' is an Irrational no.

\Rightarrow Limit of this ^{Cauchy} sequence is not in the same space in which it was defined

approaching a specific value or behaviour more closely as you continue.

Convergence! Means a point to which the sequence is approaching as it keeps progressing.

If the sequence is approaching towards a finite point, it is called a convergent sequence.

Ex: We have a sequence $\{\frac{1}{n}\}$

$\lim_{n \rightarrow \infty} \frac{1}{n} = 0 \Rightarrow$ This sequence is converging to zero (Although it will never be equal to zero)

Analogy: Dart Board \rightarrow Every shot we hit on the dartboard wants to go as near to the centre as possible.

- ② Connectedness: Meaning the space can't be divided into non-overlapping or disjoint non-empty open sets.
- ③ Continuity: Smoothness of function
 \Rightarrow Small changes in input result in small changes in O.P.
- ④ Compactness: Space is bounded and closed, making it well behaved for convergence and containment.
 $(a, b) \rightarrow$ open \Rightarrow Not compact \rightarrow Can't be covered by finite no. of open sets
 $[a, b] \rightarrow$ closed \Rightarrow Compact
- ⑤ Subspace: If W is a subset of V and W follows all properties of V , then W is called Subspace of V .
- Quantum Mechanics
- Hilbert Space
- ∞ -Dimensional Vector Space
- \hookrightarrow Relates to the range of possible states a single particle can have / occupy, not the no. of particles
- $|\Psi\rangle \rightarrow$ State of a particle in a Hilbert Space
- \rightarrow For many Quantum systems, the Hilbert Space is ∞ -dimensional.
- \rightarrow Particle can be described by a potentially ∞ set of basis functions.
- \rightarrow Spin of e^- γ 2 or 3 dimensions
 Polarization of photon
- \rightarrow Position γ ∞ Dimensions
 Energy
 Momentum

* Infinite Dimension means the complexity of states available to a particle.

→ It allows a particle's state to be represented by

a superposition of ∞ no. of basis states.

Ex:- Different energy levels in a Quantum Harmonic Oscillator

→ The particle is represented by one vector in Hilbert Space.

→ Space is ∞ -dimensional due to the range of possible states available to the particle.

→ For many particles \Rightarrow Tensor product is used

Tensor product of ^{HS for} each particle

Space becomes very large but each particle in HS is still ∞ dimensional, if space is described in terms of continuous spectrum of ∞ states.

Let $V(\mathbb{C})$ be a vector space over field of \mathbb{C}

$V(\mathbb{C}) \rightarrow$ Complex Vector Space

$|\Psi\rangle \in V(\mathbb{C})$

↓
State of a quantum particle

Axioms followed by a Ket vector:

① $|\Psi\rangle + |\Phi\rangle \in V$ (closure under +)

② $|\Psi\rangle + |\Phi\rangle = |\Phi\rangle + |\Psi\rangle$ (commutativity under +)

③ $|\Psi\rangle + \{|\Phi\rangle + |\chi\rangle\} = \{|\Psi\rangle + |\Phi\rangle\} + |\chi\rangle$ (associativity under +)

④ $|\Psi\rangle + 0 = |\Psi\rangle$ (identity under +)

⑤ $|\Psi\rangle + (-|\Psi\rangle) = 0$ (inverse under +)

$$\begin{aligned} \textcircled{6} \quad |z\Psi\rangle &= z|\Psi\rangle \\ \textcircled{7} \quad z\{|A\rangle + |B\rangle\} &= z|A\rangle + z|B\rangle \\ (z+w)|A\rangle &= z|A\rangle + w|A\rangle \end{aligned} \quad \left. \begin{array}{l} \text{Linearity} \\ \text{Addition} \end{array} \right\}$$

Note: In ordinary space, multiplying by a complex no. doesn't make any sense.

Our final answer must come in \mathbb{R}

A ket vector can be defined as a function

$A(x)$, that is a continuous, complex valued function such that you can perform addition ($A_1(x) + A_2(x)$) and can perform multiplication by a scalar and that it follows the above 7 axioms, then $A(x)$ is a ket vector.

Ex:- Matrices

$$\begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix} + \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} = \begin{pmatrix} \alpha_1 + \beta_1 \\ \alpha_2 + \beta_2 \end{pmatrix}$$

$$z \begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix} = \begin{pmatrix} z\alpha_1 \\ z\alpha_2 \end{pmatrix}$$

Note:- We normally do not mix vectors of different dimensionality

For every ket vector there exists a bra vector

$$|\Psi\rangle = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix} \quad \langle \Psi | = (\alpha_1^*, \alpha_2^*, \dots, \alpha_n^*) \quad \begin{array}{l} \text{(Complex conjugate} \\ \text{vector space)} \end{array}$$

(Bra vector also satisfies 7 axioms)

$z \rightarrow$ complex no.

$z^* \rightarrow$ Dual version

$V(F) \rightarrow$ Complex
vector
space

$V^*(F^*) \rightarrow$ Complex conjugate
vector space

$\forall |a\rangle \in V \quad \exists \quad \langle a | \in V^*$

Note: ① If $\langle a \rangle$ is a bra vector corresponding to $|a\rangle$
 & $\langle b |$ is a bra vector corresponding to $|b\rangle$
 then $|a\rangle + |b\rangle \Rightarrow \langle a + b |$

* ② $z|A\rangle \neq \langle A|z$

$\therefore z \in \mathbb{C}$ and $|A\rangle \in V(\mathbb{C})$ (vectors in \mathbb{C})

$z^* \in \mathbb{C}^*$ and $\langle A | \in V(\mathbb{C}^*)$

$\therefore \langle A | \in V^*(\mathbb{C}^*)$

$z|A\rangle \Rightarrow \langle A|z^*$

Vector Space \rightarrow Normed Space

* Length of a vector:

$$z = x + iy$$

$$z^* = x - iy$$

$$zz^* = (x+iy)(x-iy)$$

$$= x^2 - ixy^2$$

$$\boxed{zz^* = x^2 + y^2}$$

$$|\Psi\rangle = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix}$$

$$\boxed{|\Psi\rangle = \sqrt{\alpha_1^2 + \alpha_2^2 + \alpha_3^2}}$$

* Vector with length = 1 \Rightarrow Unit Vector

$$\alpha\alpha^* = \alpha^2$$

$$\underline{\text{Ex: }} |\Psi\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$|\Psi\rangle = \sqrt{1^2 + 0^2} = \underline{\underline{1}}$$

Few more examples of unit vectors:

$$| \uparrow \rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$| \leftrightarrow \rangle = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$| \downarrow \rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$| \rightarrow \rangle = \begin{pmatrix} 1/2 \\ -\sqrt{3}/2 \end{pmatrix}$$

$$| \rightarrow \rangle = \begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix}$$

$$| \leftarrow \rangle = \begin{pmatrix} \sqrt{3}/2 \\ 1/2 \end{pmatrix}$$

A vector can be made a unit vector by —

$$|\Psi\rangle = \begin{pmatrix} 3 \\ 1 \end{pmatrix}$$

$$||\Psi|| = \sqrt{10}$$

Unit vector = $\frac{|\Psi\rangle}{||\Psi||}$
↓
Normalizing factor

$$\text{Unit vector} = \frac{1}{\sqrt{10}} \begin{pmatrix} 3 \\ 1 \end{pmatrix}$$

$$\text{unit vector} = \frac{1}{\sqrt{10}} \begin{pmatrix} 3 \\ 1 \end{pmatrix}$$

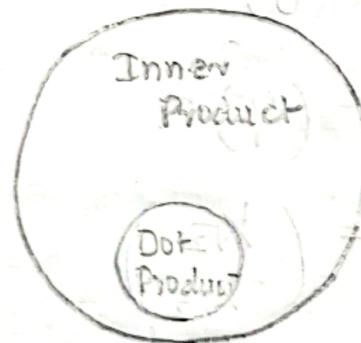
$$= \begin{pmatrix} 3/\sqrt{10} \\ 1/\sqrt{10} \end{pmatrix} \rightarrow \text{Scalar multiplication allowed because } |\Psi\rangle \text{ is an element of V's}$$

$$|\text{unit vector}| = \sqrt{\frac{9}{10} + \frac{1}{10}} = \underline{\underline{1}}$$

Vector Space \rightarrow Normed Space \rightarrow Inner product Space

* $\langle \phi | \psi \rangle = C$

This encodes information about how similar the two vectors (bra & ket) are.



* Follows two properties:

① Linearity:

$$\langle \psi | \{ |\phi \rangle + |\chi \rangle \} = \langle \psi | \phi \rangle + \langle \psi | \chi \rangle$$

② Conjugate property:

$$\langle A | B \rangle = \langle B | A \rangle^* \quad (\text{Complex conjugation})$$

* $\langle A | A \rangle$ is always a real number (Some exceptions)

Important points:

① Orthogonality:

* $\langle A | B \rangle = 0 \Rightarrow \langle A | \& | B \rangle$ are independent of each other / not similar

② Normalization:

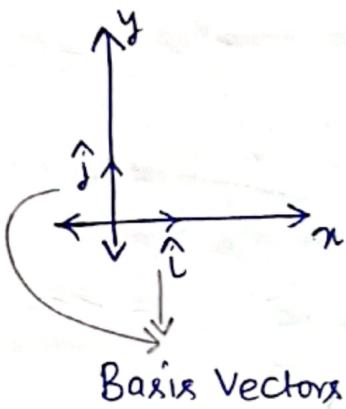
If $\langle A | A \rangle = 1$, then A is a normalized vector

③ Orthonormality:

$\langle A | B \rangle = \delta_{ij}$ if $0 \Rightarrow A \& B$ are different

$1 \Rightarrow A \& B$ are same

★ Basis = Spanning Set



$$\langle \hat{i} | \hat{i} \rangle = 1$$

$$\langle \hat{i} | \hat{j} \rangle = 0$$

$$\hat{j} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad (\hat{i}, \hat{j}) \text{ is a basis}$$

$$\hat{j} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad (\hat{i}, \hat{j}) = \langle \hat{i} | \hat{j} \rangle$$

$$|\psi\rangle = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} = \langle \hat{i} | \hat{i} \rangle + \langle \hat{j} | \hat{j} \rangle$$

$$|\psi\rangle = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} = \langle \hat{i} | \hat{i} \rangle + \langle \hat{j} | \hat{j} \rangle$$

Let $V(F)$ is a vector space over the field F

$\{ |i\rangle \}$ → Orthonormal Basis (In Quantum Mechanics, basis (Sequence of orthonormal vectors) must be orthonormal)

(Now, any vector can be written in terms of the basis)

$$|A\rangle = \sum_i \alpha_i |i\rangle \rightarrow ①$$

$$\vec{A} = 2\hat{i} + 3\hat{j} + 4\hat{k}$$

$$\vec{A} \cdot \hat{i} = 2\hat{i} \cdot \hat{i} + 3\hat{j} \cdot \hat{i} + 4\hat{k} \cdot \hat{i}$$

$$\vec{A} \cdot \hat{i} = 2$$

$$\Rightarrow \vec{A} = (\vec{A} \cdot \hat{i}) \hat{i} + (\vec{A} \cdot \hat{j}) \hat{j} + (\vec{A} \cdot \hat{k}) \hat{k}$$

$$+ (\vec{A} \cdot \hat{k}) \hat{k}$$

$$\langle j | A \rangle = \langle j | \sum_i \alpha_i | i \rangle$$

$$\downarrow$$

$$\langle j | A \rangle = \sum_i \alpha_i \langle j | i \rangle$$

Collapse + noisy result

Substituting in eqn ①

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

Some example questions:

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

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

Standard Basis → used for spin of e^{\pm}

$$\langle 1 | 1 \rangle = (1 \ 0) \begin{pmatrix} 1 \\ 0 \end{pmatrix} = 1$$

$$(1)_{\text{left}} + (1)_{\text{right}} = (2)$$

$$\langle \downarrow | \downarrow \rangle = (0 \ 1) \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \underline{1}$$

$$\langle \uparrow | \downarrow \rangle = (1 \ 0) \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \underline{0}$$

$$② |\rightarrow\rangle = \begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix}$$

$$③ |\uparrow\rangle = \begin{pmatrix} 1/\sqrt{2} \\ -\sqrt{3}/2 \end{pmatrix}$$

$$|\leftarrow\rangle = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}$$

$$|\downarrow\rangle = \begin{pmatrix} \sqrt{3}/2 \\ 1/2 \end{pmatrix}$$

$$\text{H.W } \langle \rightarrow | \rightarrow \rangle = \langle \leftarrow | \leftarrow \rangle = \langle \uparrow | \uparrow \rangle = \langle \downarrow | \downarrow \rangle = 1$$

$$\langle \leftarrow | \rightarrow \rangle = \langle \downarrow | \uparrow \rangle = 0$$

* Orthonormal Basis \rightarrow Span the complete Vector Space

\Rightarrow Using orthonormal basis, any vector in the vector space can be written.

Linear Combination:

Any vector from a vector space can be written as the linear combination of the basis vectors of the vector space.

$$|v\rangle \in V(F)$$

$\{|\uparrow\rangle, |\downarrow\rangle\} \rightarrow$ Orthonormal Basis

$$\text{Ex- } ① |v\rangle = \begin{pmatrix} c \\ d \end{pmatrix}$$

$$|v\rangle = x_1 |\uparrow\rangle + x_2 |\downarrow\rangle$$

$$\begin{pmatrix} c \\ d \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$\langle v | \langle a | b \rangle \leq \langle a | b \rangle$$

$$\text{① } \langle v | \langle a | b \rangle \leq \langle a | b \rangle$$

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} = \langle \uparrow | \circledast$$

$$\begin{pmatrix} 0 \\ 1 \end{pmatrix} = \langle \downarrow | \circledast$$

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} (\circledast, \circledast) = \langle \uparrow | \circledast$$

$$\begin{pmatrix} c \\ d \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad \text{is called linear combination of } \mathbf{v}_1, \mathbf{v}_2$$

$$\Rightarrow x_1 = c$$

$$\text{and } x_2 = d \quad \text{is called linear combination of } \mathbf{v}_1, \mathbf{v}_2$$

② Another example of basis:

$$| \rightarrow = \begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix} \quad | \leftarrow = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix} \quad v = \langle \sqrt{1}, \sqrt{2} \rangle$$

$$\boxed{\text{Let } |v\rangle = x_1 | \rightarrow + x_2 | \leftarrow}$$

Multiply both sides by $\langle \cdot |$

$$\Rightarrow \langle \cdot | v \rangle = x_1 \cancel{\langle \cdot | \rightarrow \rangle} + x_2 \cancel{\langle \cdot | \leftarrow \rangle}$$

$$\Rightarrow x_2 = \langle \cdot | v \rangle$$

$$= \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \end{pmatrix} \begin{pmatrix} c \\ d \end{pmatrix}$$

$$= \frac{c}{\sqrt{2}} + \frac{d}{\sqrt{2}}$$

$$= \frac{c+d}{\sqrt{2}}$$

$$\text{Hence } x_1 = \langle \cdot | v \rangle$$

$$= \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} \end{pmatrix} \begin{pmatrix} c \\ d \end{pmatrix}$$

$$= \frac{c-d}{\sqrt{2}}$$

$$\Rightarrow |v\rangle = \left(\frac{c-d}{\sqrt{2}} \right) | \rightarrow + \left(\frac{c+d}{\sqrt{2}} \right) | \leftarrow$$

$$= \left(\frac{c-d+2c+2d}{\sqrt{2}} \right) | \rightarrow + \left(\frac{c+d+2c+2d}{\sqrt{2}} \right) | \leftarrow$$

* Considering ∞ -dimensional vector space:

$$|\Psi\rangle = a_1|b_1\rangle + a_2|b_2\rangle + \dots + a_n|b_n\rangle \quad \text{Eqn ①}$$

$x^{by} < b_1 |$ on b.s

$$\langle b_1 | \Psi \rangle = a_1 \langle b_1 | b_1 \rangle + a_2 \cancel{\langle b_1 | b_2 \rangle} + \dots + a_n \cancel{\langle b_1 | b_n \rangle}$$

(collapse)

$$\langle b_1 | \Psi \rangle = a_1$$

\Rightarrow Eqn ① becomes

$$|\Psi\rangle = \langle b_1 | \Psi \rangle |b_1\rangle + \langle b_2 | \Psi \rangle |b_2\rangle + \dots + \langle b_n | \Psi \rangle |b_n\rangle$$

~~Set up basis and b' x~~

Probability Amplitude

$\underbrace{\langle b_2 | \Psi \rangle}_{\text{to } |b_2\rangle} \langle \Psi | b_2 \rangle^* \rightarrow$ Probability of $|\Psi\rangle$ jumping

Square of Probability Amplitude

* Linear Dependence:

A set of vectors $\{|v_1\rangle, |v_2\rangle, \dots, |v_n\rangle\}$ is linearly dependent if \exists scalars c_1, c_2, \dots, c_n such that at least one of them is non-zero and $c_1|v_1\rangle + c_2|v_2\rangle + \dots + c_n|v_n\rangle = 0$, then the set of vectors are linearly dependent.

If $c_1|v_1\rangle + c_2|v_2\rangle + \dots + c_n|v_n\rangle = 0$ if $c_1 = c_2 = \dots = c_n = 0$, then the set of vectors is called linearly independent.

Ex: $(2 -1), (3 2), (4 3) \rightarrow$ LI?

$$c_1(2 -1) + c_2(3 2) + c_3(4 3) = (0, 0)$$

$$(2c_1 + 3c_2 + 4c_3 \quad -c_1 + 2c_2 + 3c_3) =$$

* Basis

- ↳ Spans the entire VS \Rightarrow Orthonormality
- ↳ Linear Independence

Note: Any two sets of linearly independent elements which span the Vector Space, they have same number of elements.

* A vector space can have ∞ no. of distinct basis vectors, but they all have same no. of elements.

Ex: $| \uparrow \rangle, | \downarrow \rangle$ $| \uparrow \rangle, | \downarrow \rangle$ $| \leftarrow \rangle, | \rightarrow \rangle$

* No. of elements in Basis set = Dimension of Vector Space

* In Quantum Computing \rightarrow Finite Dim. Space
Quantum Mechanics \rightarrow ∞ Dim. Space

* Ordered Basis:

$$\{ | \uparrow \rangle, | \downarrow \rangle \} \neq \{ | \downarrow \rangle, | \uparrow \rangle \}$$

MATRICES

$$A = \begin{bmatrix} a & d & g \\ b & e & h \\ c & f & i \end{bmatrix}_{3 \times 3}$$

① Square Matrix

$$\langle a \rangle = [a \ b \ c]$$

② Row Matrix

$$| a \rangle = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

④ Identity Matrix

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$AI = A = IA \quad (AB \neq BA \text{ in general})$$

⑤ Transpose

$$A^T = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$

⑥ Trace of Matrix

$$\text{Tr}(A) = \text{Sum of diagonal elements} \\ = a + e + i$$

⑦ Symmetric Matrix:

$$A^T = A$$

$$\underline{\text{Ex:}} \quad A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\text{Since } A^T = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

and both the given matrix are equal to its own transpose.

⑧ Hermitian Matrix

$$A^H = A$$

$$A^H = (\bar{A})^T$$

$$\underline{\text{Ex:}} \quad A = \begin{pmatrix} 3 & 1-i \\ 1+i & -2 \end{pmatrix}$$

$$\bar{A} = \begin{pmatrix} 3 & 1+i \\ 1-i & -2 \end{pmatrix}$$

$$(\bar{A})^T = \begin{pmatrix} 3 & 1-i \\ 1+i & -2 \end{pmatrix}$$

$$= \underline{\underline{A}}$$

Note: Diagonal elements of HM is always real

⑨ Orthogonal Matrices

$$AA^T = I = A^T A$$

$$\underline{\text{Ex:}} \quad ① A = \begin{pmatrix} \cos x & \sin x \\ -\sin x & \cos x \end{pmatrix}$$

$$A^T = \begin{pmatrix} \cos x & -\sin x \\ \sin x & \cos x \end{pmatrix}$$

$$AA^T = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

Rows & columns form an orthonormal set of vectors:

$$\begin{pmatrix} \cos x \\ \sin x \end{pmatrix} (-\sin x \cos x) = -\cos x \sin x + \sin x \cos x = 0$$

$$\begin{pmatrix} \cos x \\ \sin x \end{pmatrix} (\cos x \sin x) = \cos^2 x + \sin^2 x = 1$$

$$\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \underline{\underline{I}}$$

$$\underline{\text{Ex: ②}} \quad A = \begin{pmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \end{pmatrix} \quad A^T = \begin{pmatrix} \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \end{pmatrix}$$

$$A A^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

⑩ Unitary Matrices

$$AA^T = I = A^T A$$

$$\underline{\text{Ex: ①}} \quad A = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ i\frac{1}{\sqrt{2}} & -i\frac{1}{\sqrt{2}} \end{pmatrix}$$

$$A^T = \begin{pmatrix} \frac{1}{\sqrt{2}} & -i\frac{1}{\sqrt{2}} \\ i\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$AA^T = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\underline{\text{Ex: ②}} \quad B = \begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix} \Rightarrow B^T = \begin{bmatrix} 0 & i \\ i & 0 \end{bmatrix}$$

$$Q. A = \{ |b_1\rangle, |b_2\rangle, \dots, |b_n\rangle \}$$

Check if A forms basis set

① Check normalization

$$\langle b_1 | b_1 \rangle = 1$$

② Orthogonality

$$A^T A = I$$

$$\begin{bmatrix} \langle b_1 | \\ \langle b_2 | \\ \vdots \\ \langle b_n | \end{bmatrix}_{n \times 1} \begin{bmatrix} |b_1\rangle & |b_2\rangle & \dots & |b_n\rangle \end{bmatrix} =$$

$$= \begin{bmatrix} \langle b_1 | b_1 \rangle & \langle b_1 | b_2 \rangle & \dots & \langle b_1 | b_n \rangle \\ \langle b_2 | b_1 \rangle & \langle b_2 | b_2 \rangle & \dots & \langle b_2 | b_n \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle b_n | b_1 \rangle & \langle b_n | b_2 \rangle & \dots & \langle b_n | b_n \rangle \end{bmatrix}_{n \times n}$$

$$= \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}_{n \times n} = I$$

LINEAR OPERATORS

- States in Quantum Mechanics are represented by vectors in V-S
- Physical observables are represented by Linear Operators in Vector Space.

↓
For now we will take it as an axiom

- Energy, position, angular momentum, momentum
- Linear operators are also associated with a vector space but they are not vectors.

(Think of Linear operators as a Machine that transforms)

- Not every operator is linear

- If anything goes into the linear operator (machine), then something must come out.

If it doesn't give an output, it isn't a linear operator.

Axioms for linearity:

$$\textcircled{1} \quad M\{z|A\rangle\} = zM|A\rangle$$

$$\textcircled{2} \quad M\{|A\rangle + |B\rangle\} = M|A\rangle + M|B\rangle$$

* Row and column representation of vectors and linear transformations is dependent on the chosen basis

$$|A\rangle = \sum_i a_i |i\rangle$$

$$M|A\rangle = |B\rangle$$

~~Multiplying both sides with $\langle k |$~~

$$M \sum_i a_i |i\rangle = \sum_i b_i |i\rangle$$

~~Multiplying both sides with $\langle k |$~~

$$\sum_i a_i \langle k | M | i \rangle = \sum_i b_i \langle k | i \rangle$$

$\langle k | M | i \rangle$
vector

Complex
Number

$$\sum_i a_i m_{ki} = b_k$$

Now m_{ki} is an element of the matrix M , where M is the Linear Operator.

Dimensions of M depend on the basis vectors we choose.

$$\begin{pmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{pmatrix} \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

$\downarrow \quad \downarrow \quad \downarrow$

$$M |A\rangle = |B\rangle$$

Note:- Relationship b/w vectors and operators is independent of basis we choose when dealing with abstract notations.

* Identity Operator

$$I_v |v\rangle = |v\rangle, |v\rangle \in V(F)$$

* Zero Operator

$$0 |v\rangle = 0, |v\rangle \in V(F)$$

* $V: A \rightarrow B, W: B \rightarrow C$ where V and W both are linear operators

$$AB(|v\rangle) = A(B|v\rangle)$$

Linear operators \equiv Matrices (But for this we have to have specific input and output basis of vector space)

EIGEN VALUES & EIGEN VECTORS:

→ In general, a linear operator changes the direction of vector (input)

→ But there are certain Linear Operators that preserve the direction of Input Vector

$$M|\lambda\rangle = \lambda|\lambda\rangle$$

↓
complex no.

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \cos \theta & \sin \theta \\ \sin \theta & -\cos \theta \end{pmatrix} \begin{pmatrix} \lambda \\ 0 \end{pmatrix} = \lambda \begin{pmatrix} \cos \theta & \sin \theta \\ \sin \theta & -\cos \theta \end{pmatrix} \begin{pmatrix} \lambda \\ 0 \end{pmatrix}$$

Ex: ① $M = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}$ $|v\rangle = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

$$M|v\rangle = \begin{pmatrix} 3 \\ 3 \end{pmatrix} = 3 \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$M|v\rangle = 3|v\rangle$$

↓
Eigenvalue

② $|u\rangle = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$

$$M|u\rangle = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$= -1 \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$M|u\rangle = \underline{-1}|u\rangle$$

③ $|u'\rangle = \begin{pmatrix} 1 \\ i \end{pmatrix}$ $N = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$

$$N|u'\rangle = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ i \end{pmatrix} = \begin{pmatrix} -i \\ 1 \end{pmatrix} = -i \begin{pmatrix} 1 \\ i \end{pmatrix}$$

$$\Rightarrow N|u'\rangle = \underline{-i|u'\rangle}$$

$= -i \begin{pmatrix} 1 \\ i \end{pmatrix}$

* $M|\psi\rangle \neq \langle\psi|M$

we must find the complex conjugate of M

Dual space में जाते ही conjugate भी जाते हैं
 $z|A\rangle = |B\rangle \Rightarrow \langle A|z^* = \langle B|$

Finding complex conjugation of M

$$M|A\rangle = |B\rangle$$

$$M \sum_i a_i |i\rangle = \sum_i b_i |i\rangle$$

$$\sum_i a_i \langle M|i\rangle = \sum_i b_i \langle i|i\rangle$$

$$\sum_i a_i \langle j|M|i\rangle = \sum_i b_i \langle j|i\rangle$$

$$\sum_i a_i m_{ji} = b_j$$

In dual space;

$$\sum_i a_i^* m_{ji}^* = b_j^*$$

$$\Rightarrow M|A\rangle = |B\rangle$$

↓

$$\langle A|M^+ = \langle B|$$

$$M = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \Rightarrow M^+ = \begin{pmatrix} a^* & c^* \\ b^* & d^* \end{pmatrix}$$

M^+ is Hermitian conjugate

Note: Outcome of an experiment is always a real no.

② Observables in Quantum Mechanics are Hermitian operators.

$$M = M^+$$

⇒ All eigen values are real

Proof: $M|\lambda\rangle = \lambda|\lambda\rangle$

$$\langle \lambda|M^+ = \lambda^* \langle \lambda|$$

Applying $|\lambda\rangle$ on both sides

$$\langle \lambda|M^+|\lambda\rangle = \lambda^* \langle \lambda|\lambda\rangle$$

$$\Theta \langle \lambda|M|\lambda\rangle = \Theta \lambda \langle \lambda|\lambda\rangle \quad (\because M \text{ is Hermitian})$$

$$0 = \langle \lambda|\lambda\rangle(\lambda^* - \lambda)$$

$$\Rightarrow M^+ = M$$

$$\Rightarrow 0 = (\lambda^* - \lambda) (\langle \lambda | \lambda \rangle)$$

λ^* is eigen value of A

$$\Rightarrow \lambda^* = \lambda$$

\Rightarrow All eigen values are real

$$|\lambda| = \sqrt{\lambda^* \lambda} = \sqrt{\lambda \lambda} = \lambda$$

FOURIER TRANSFORM

↳ A way to see a funcⁿ from different perspective

Fourier Series:

→ Fourier series is used to express a complex periodic function into the sum of sines and cosines, each with its own amplitude.

$$f(x) = f(x + T) \quad \text{Time period}$$

Periodic function

(Here sines and cosines work as basis)

$$\text{Ex: } f(x) = 1 \sin \omega t + \frac{1}{2} \sin 2\omega t + \frac{3}{2} \cos \omega t$$

$$\omega = 2\pi\nu$$

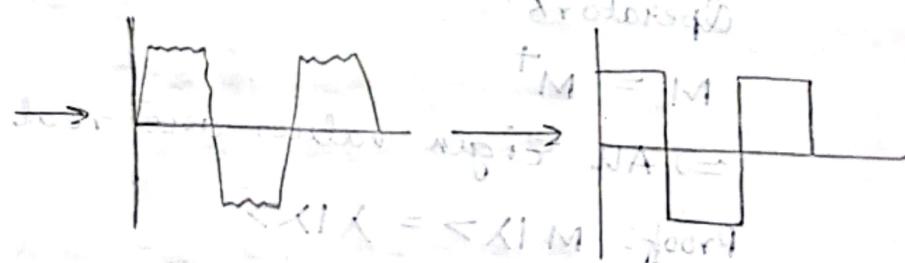
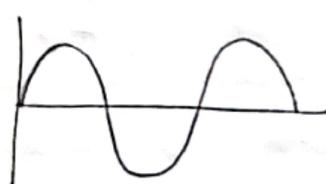
$$\downarrow$$

Angular velocity

Frequency (no. of waves per second)

$$\omega = \frac{\theta}{t} \rightarrow \text{Angular displacement}$$

$$\Rightarrow \theta = \omega t$$

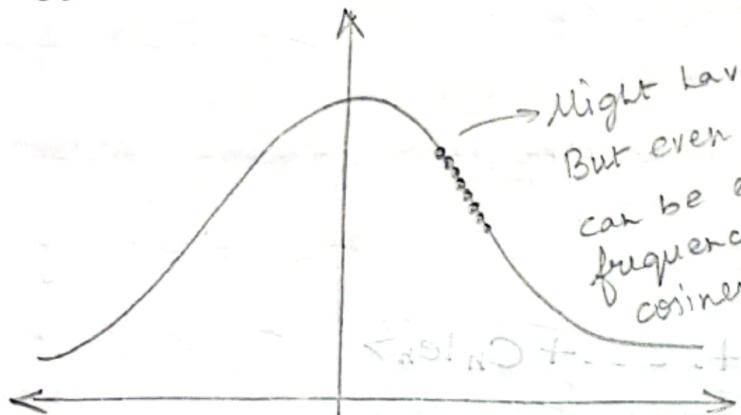


Heisenberg's uncertainty principle \rightarrow Heart of Q.M

Fourier Transform \rightarrow Heart of Heisenberg principle

Fourier Transform:

→ Expressing a non-periodic function in terms of its frequency content.



Might have to consider ∞ points
But even this non-periodic function
can be expressed in terms of
frequency content (using sines & cosines or e)

$$e^{i\theta} = \cos \theta + i \sin \theta$$

$$e^{i\pi} = -1$$

$$e^{i\pi} + 1 = 0$$

↳ When using Fourier Transform
we express func' in terms
of exponential (e).

→ A non-periodic function can be thought of as having a continuous spectrum of frequencies.



can be represented as a continuous sum (integral) of
sines, cosines or exponential ($e^{i\theta}$)
(even maybe by using almost all frequencies)

L^2 Space

→ Square Integrable functions

$$\int_{-\infty}^{\infty} |f|^2 dx < \infty$$

(Finite)

In Linear Algebra,

$$|v\rangle = c_1|e_1\rangle + c_2|e_2\rangle + \dots + c_n|e_n\rangle$$

$|e_i\rangle$ → Basis element

c_i → Coefficients

These coefficients tell us how much each basis element contributes to regenerate the original vector

In Fourier Series,

$f(t)$ → Periodic function

$$f(t) = a_0 + \sum_{n=1}^{\infty} [a_n \sin(n\omega_0 t) + b_n \cos(n\omega_0 t)]$$

* Why use just sines and cosines?

Orthogonality

Sine and cosine are independent of each other

(Like orthogonal vectors in v.s)

Completeness

Set of sine & cosine

is a complete set

Can generate a wide

variety of functions, specially from L^2 space

$$* e^{i\omega t} = \cos \omega t + i \sin \omega t$$

$$* \cos \theta = \frac{e^{i\theta} + e^{-i\theta}}{2}$$

$$= \frac{\cos \theta + i \sin \theta + \cos \theta - i \sin \theta}{2}$$

$$= \cos \theta$$

$$* \sin \theta = \frac{e^{i\theta} - e^{-i\theta}}{2i}$$

$$\Rightarrow f(t) = a_0 + \sum_{n=1}^{\infty} [a_n \sin(n\omega t) + b_n \cos(n\omega t)]$$

$$\Rightarrow f(t) = a_0 + \sum_{n=1}^{\infty} \left[a_n \left(\frac{e^{i\theta} - e^{-i\theta}}{2i} \right) + b_n \left(\frac{e^{i\theta} + e^{-i\theta}}{2} \right) \right]$$

$$= a_0 + \sum_{n=1}^{\infty} \left[a_n \left(\frac{e^{in\omega t} - e^{-in\omega t}}{2i} \right) + b_n \left(\frac{e^{in\omega t} + e^{-in\omega t}}{2} \right) \right]$$

$$= a_0 + \sum_{n=1}^{\infty} \left[\frac{a_n e^{in\omega t}}{2i} - \frac{a_n e^{-in\omega t}}{2i} + \frac{b_n e^{in\omega t}}{2} + \frac{b_n e^{-in\omega t}}{2} \right]$$

$$= a_0 + \sum_{n=1}^{\infty} \left[\frac{a_n + i b_n}{2i} e^{in\omega t} + \frac{i b_n - a_n}{2i} e^{-in\omega t} \right]$$

$$= a_0 + \sum_{n=1}^{\infty} \left[\frac{i a_n - b_n}{-2} e^{in\omega t} + \frac{-b_n - i a_n}{-2} e^{-in\omega t} \right]$$

$$= a_0 + \sum_{n=1}^{\infty} \left[\frac{b_n - i a_n}{2} e^{in\omega t} + \frac{b_n + i a_n}{2} e^{-in\omega t} \right]$$

$$= a_0 + \sum_{n=1}^{\infty} \left[C_n e^{in\omega t} + C_{-n} e^{-in\omega t} \right]$$

$$f(t) = \sum_{n=-\infty}^{\infty} C_n e^{in\omega t}$$

where, $C_n \rightarrow$ Complex Exponential Fourier coefficient

* Derive $C_n = \frac{1}{T} \int_0^T f(t) e^{-in\omega t} dt$

$$WKT \quad f(t) = \sum_{n=-\infty}^{\infty} C_n e^{in\omega t}$$

Soln. Let $f(t)$ be a periodic function with period T

$$f(t) = \sum_{k=-\infty}^{\infty} C_k e^{ik\omega t}$$

① \times by $e^{-in\omega t}$ (for a fixed integer n) and integrate

over [a period 0 to T]

$$\Rightarrow \left[\int_0^T e^{-in\omega t} \cdot f(t) dt = \sum_{k=-\infty}^{\infty} C_k \int_0^T e^{i(k-n)\omega t} dt \right] _{t=0}^{t=T} = (f) + \dots$$

$$\left[\int_0^T e^{-in\omega t} \cdot f(t) dt = \int_0^T \left(\sum_{k=-\infty}^{\infty} C_k e^{i(k-n)\omega t} \right) dt \right] _{t=0}^{t=T} =$$

② Exchange sum and integral (Fubini's theorem):

$$\left[\int_0^T e^{-in\omega t} \cdot f(t) dt = \sum_{k=-\infty}^{\infty} C_k \int_0^T e^{i(k-n)\omega t} dt \right] _{t=0}^{t=T} =$$

③ Use orthogonality of complex exponentials

* Any two exponentials $e^{ik\omega t}$ and $e^{in\omega t}$ are orthogonal over a period T .

\Rightarrow Inner product (integral of their product over one period) is zero if $k \neq n$ (they have different frequencies)

\Rightarrow Only when $k = n$ (same frequency), their inner product is non-zero ($= T$).

$$\Rightarrow \int_0^T e^{i(k-n)\omega t} dt = \begin{cases} T, & \text{if } k=n \\ 0, & \text{if } k \neq n \end{cases}$$

* If $k \neq n$:

$$\Rightarrow \int_0^T e^{i(k-n)w_0 t} f(t) dt = \frac{[e^{i(k-n)w_0 T}]_0}{i(k-n)w_0}$$
$$= \frac{e^{i(k-n)w_0 T} - e^0}{i(k-n)w_0}$$
$$= \frac{e^{i(k-n)\frac{2\pi}{T} \cdot T} - 1}{i(k-n)w_0}$$
$$= \frac{e^{i(k-n)2\pi} - 1}{i(k-n)w_0}$$
$$= \frac{\cos(2\pi(k-n)) + i \sin(2\pi(k-n)) - 1}{i(k-n)w_0}$$
$$= \frac{1 + 0 - 1}{i(k-n)w_0}$$
$$= \underline{\underline{0}}$$

$$\cos(n\pi) = (-1)^n$$

$$\sin(n\pi) = 0$$

\Rightarrow When $k = n$

$$④ \int_0^T e^{-inw_0 t} f(t) dt = \sum_{k=-\infty}^{\infty} C_k \int_0^T 1 \cdot dt$$

$$\int_0^T e^{-inw_0 t} f(t) dt = C_n \cdot T$$

(All terms vanish except when $k = n$)

$$⑤ C_n = \frac{1}{T} \int_0^T e^{-inw_0 t} f(t) dt$$

* Why $e^{i\omega t}$ is chosen?

Orthogonality

$e^{i\omega t}, e^{-i\omega t}$

All are independent of each other

Completeness

Any well behaved funcⁿ can be expressed as the linear combination of $e^{i\omega t}$

Periodic nature

It traces a circle in the complex plane as t varies

$$f(t) = \sum_{n=-\infty}^{\infty} c_n e^{i\omega n t}$$

$$|v\rangle = c_1 |e_1\rangle + c_2 |e_2\rangle + \dots$$

$$\tilde{f}(t) = \int_{-\infty}^{\infty} \tilde{f}(w) e^{i\omega t} dw$$

$e^{i\omega t}$ → basis funcⁿ's (combination of sines and cosines)

$\tilde{f}(w)$ → Coefficient (tells us how much a particular $e^{i\omega t}$ contributes to regenerate the original function)

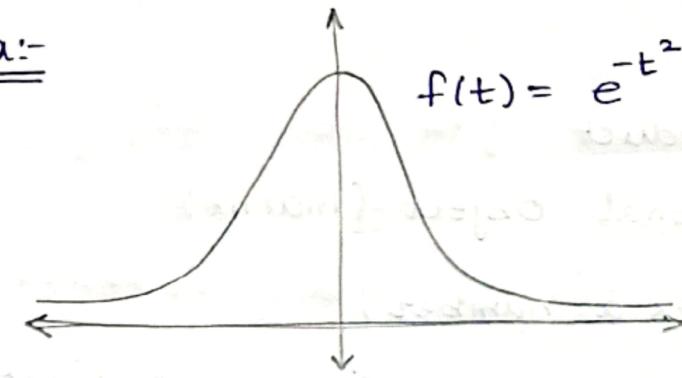
$$\tilde{f}(w) = \int_{-\infty}^{\infty} f(t) e^{i\omega t} dt$$

Original function:

$$\rightarrow f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \tilde{f}(w) e^{i\omega t} dw$$

Normalizing factor

Ex:-



$$f(t) = e^{-t^2}$$

Amplitude of freq component in original function

$$\tilde{f}(w) = \int_{-\infty}^{\infty} f(t) \cdot e^{-iwt} \cdot dt$$

$$= \int_{-\infty}^{\infty} e^{-t^2} \cdot e^{-iwt} \cdot dt$$

$$\tilde{f}(w) = \frac{1}{\sqrt{\pi}} e^{-\frac{w^2}{4}}$$

$$\Rightarrow e^{-t^2} = \int_{-\infty}^{\infty} \underbrace{\frac{1}{\sqrt{\pi}} e^{-\frac{w^2}{4}}}_{\text{coeff}} \underbrace{e^{iwt}}_{\text{basis function}} \cdot dt$$

Like \hat{i}, \hat{j} and \hat{k} span all 3D space

e^{iwt} can span all functions of L^2 space (especially)

$$\tilde{f}(w) = \int_{-\infty}^{\infty} f(t) e^{-iwt} \cdot dt$$

→ This transforms time domain funcⁿ to frequency domain function

Inverse fourier transform;

$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \tilde{f}(w) e^{iwt} \cdot dw$$

→ This transforms frequency domain funcⁿ to time domain funcⁿ

Outer Product

- Specific type of tensor product
- Results a higher dimensional object (Matrix)
- Unlike inner product results a number,
Outer Product results a matrix (higher dimensional object)

→ Ex: $|\psi\rangle\langle\phi|$

↳ Higher dimensional space

$|u\rangle \in \mathbb{R}^n$

$|v\rangle \in \mathbb{R}^m$

$|u\rangle\langle v| \in \mathbb{R}^{n \times m}$

$$\begin{pmatrix} \cdot \\ \vdots \\ \cdot \end{pmatrix}_{n \times 1} \begin{pmatrix} \cdot & \dots & \cdot \\ \vdots & \ddots & \vdots \\ \cdot & \dots & \cdot \end{pmatrix}_{1 \times m} = \begin{pmatrix} \cdot & \dots & \cdot \\ \vdots & \ddots & \vdots \\ \cdot & \dots & \cdot \end{pmatrix}_{n \times m}$$

$$\begin{pmatrix} \cdot \\ \vdots \\ \cdot \end{pmatrix}_{n \times 1} \begin{pmatrix} \cdot & \dots & \cdot \\ \vdots & \ddots & \vdots \\ \cdot & \dots & \cdot \end{pmatrix}_{m \times 1} = \begin{pmatrix} \cdot & \dots & \cdot \\ \vdots & \ddots & \vdots \\ \cdot & \dots & \cdot \end{pmatrix}_{n \times m}$$

$$\begin{pmatrix} \cdot \\ \vdots \\ \cdot \end{pmatrix}_{n \times 1} \begin{pmatrix} \cdot & \dots & \cdot \\ \vdots & \ddots & \vdots \\ \cdot & \dots & \cdot \end{pmatrix}_{m \times 1} = \begin{pmatrix} \cdot & \dots & \cdot \\ \vdots & \ddots & \vdots \\ \cdot & \dots & \cdot \end{pmatrix}_{n \times m}$$

→ Outer product preserves the full structure of both the vectors
(vectors do not lose their properties)

→ It's a way of "spreading out" the interaction between
two vectors in higher dimensional space.

→ Unlike dot product which collapses into a single number.

→ In QM, it is used to represent Projection Operator.

$|\psi\rangle$ → Represents state of a particle in Hilbert Space.

$|\psi\rangle\langle\phi|$ → Outer product

$$\text{Ex: } (\langle \psi | \phi \rangle) |x\rangle = |\psi\rangle \langle \phi | x \rangle$$

→ First $\langle \phi |$ is projected on $|x\rangle$ and then the vector $|\psi\rangle$

$|\psi\rangle$ is scaled using the number obtained after
projection ($\langle \phi | x \rangle$) whereas when A is used A is

→ Projection operator performs two tasks; projection + scaling

Representation of Linear Operator as Outer Product

$$|u\rangle \in \text{IPS} \quad V(F)$$

$$|v\rangle \in \text{IPS} \quad V(F)$$

$$(\langle u | v \rangle) |w\rangle = \cancel{\langle v | w \rangle} |u\rangle$$

↓

$$|u\rangle \langle v|$$

acting on

$$|w\rangle$$

↓

Scaling $|u\rangle$
by $\langle v | w \rangle$

$$(\langle u | \cancel{\langle w | v \rangle}) A (\langle v | \cancel{\langle w | z \rangle}) =$$

$$\langle w | A | z \rangle$$

→ We will define OP such that both definitions are equivalent

→ Considering $\sum a_i |u_i\rangle \langle w_i|$ as a linear operator acting
on $|v\rangle$ then $(\sum a_i |u_i\rangle \langle w_i|) |v\rangle$ gives $\sum a_i \langle w_i | v \rangle |u_i\rangle$

→ Let $|i\rangle$ → Orthonormal Basis for VS $V(F)$

⇒ Any vector $|v\rangle \in V(F)$ can be written as

$$|v\rangle = \sum v_i |i\rangle, \quad v_i \in \mathbb{C}$$

$$\langle i | v \rangle = v_i$$

(Assume $|v\rangle$ is an orthogonal
vector)

↳ Dot product = 0, Perpendicular,

Linear independence

$$(\sum |i\rangle \langle i|) |v\rangle = \sum |i\rangle \langle i | v \rangle |i\rangle = \sum v_i |i\rangle = |v\rangle$$

$$\Rightarrow (\sum |i\rangle \langle i|) = I \rightarrow \text{Completeness Relation}$$

Completeness Relation is used to give the means of representing a Linear Operator as Outer Product.

Let A be a Linear operator mapping vectors from Vector Space V to Vector Space W

$$A: V \rightarrow W$$

$|v_i\rangle \rightarrow$ Orthonormal basis of V

$|w_j\rangle \rightarrow$ Orthonormal basis of W

$$A = I_V A I_W$$

$$= \left(\sum_i |v_i\rangle \langle v_i| \right) A \left(\sum_j |w_j\rangle \langle w_j| \right)$$

$$A = \sum_{ij} |v_i\rangle \langle w_j| \quad \langle v_i | A | w_j \rangle \quad (\text{Can combine because indices are different (i and j)})$$

Outer product representation of A

So far, Linear Operators can be represented as:

① Abstract Notation ② Matrix Notation ③ Outer product

2. Cauchy-Schwarz Inequality

→ Tells an important point about Hilbert Space

→ Any two vectors that belong to the Hilbert

$$|u\rangle \& |v\rangle \in H.S$$

$$|\langle u | v \rangle|^2 \leq \langle \langle u | u \rangle \langle v | v \rangle \rangle \leq \langle \langle v | v \rangle \langle u | u \rangle \rangle$$