

Linear algebra in Dirac notation

1) Vector spaces: Let V be a (complex) vector space.

A non-zero vector $v \in V$ will be denoted by

$|v\rangle$: vector in ket notation

Ex $\mathbb{C}^n = \{ (z_1, \dots, z_n) \mid z_i \in \mathbb{C} \}$

vectors, are denoted by

$$|v\rangle = \begin{pmatrix} z_1 \\ \vdots \\ z_n \end{pmatrix}$$

Rem Zero vector is denoted by 0 .

2) The span of $\{ |v_1\rangle, \dots, |v_n\rangle \}$

is the subspace of V given by

$$\text{span} \{ |v_1\rangle, \dots, |v_n\rangle \} = \left\{ |v\rangle \mid \sum_i a_i |v_i\rangle \quad a_i \in \mathbb{C} \right\}$$

If $\text{span} \{ |v_1\rangle, \dots, |v_n\rangle \} = V$ then

$\{ |v_1\rangle, \dots, |v_n\rangle \}$ is called a spanning set

3) A set $\{|v_1\rangle, \dots, |v_n\rangle\}$ is linearly dependent if there exist $a_1, \dots, a_n \in \mathbb{F}$, not all 0, such that

$$\sum_i a_i |v_i\rangle = 0$$

4) A set of vectors is linearly independent if it is not linearly dependent.

Fact Any two linearly independent spanning sets have the same number of vectors.

5) Such a set is called a basis and this number is called the dimension of the vector space, denoted by $\dim(V)$.

We will only be concerned with finite-dimensional vector spaces.

Ex A basis for \mathbb{C}^n

$$\left\{ \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \dots, \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix} \right\}$$

We will write

$$|e_i\rangle = \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix} \leftarrow \text{ith}$$

For \mathbb{C}^2 we write $\{|0\rangle, |1\rangle\}$

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

Linear operators

1) A linear operator between two vector spaces V and W is a function

$$A: V \rightarrow W \quad |v\rangle \mapsto A|v\rangle$$

such that

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

Ex Identity operator

$$I_V: V \rightarrow V \quad |v\rangle \mapsto |v\rangle$$

Zero operator

$$O_V: V \rightarrow V \quad |v\rangle \mapsto 0$$

We will write $L(V, W)$ for the set of linear operators $A: V \rightarrow W$.

Fact $L(V, W)$ is a complex vector space

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

$$(\alpha A)|v\rangle = \alpha A|v\rangle \quad \alpha \in \mathbb{C}$$

2) Given U, V, W vector spaces we can define composition

$$L(U, V) \times L(W, U) \rightarrow L(W, V)$$

$$(A: U \rightarrow V, B: W \rightarrow U) \mapsto AB: W \rightarrow V \\ |w\rangle \mapsto A(B|w\rangle)$$

3) We say V and W are isomorphic if there exists $A: V \rightarrow W$ and $B: W \rightarrow V$ such that $AB = I_W$ and $BA = I_V$.

Notation: $V \cong W$

Ex Let V be vector space of dim n .

$\{ |v_1\rangle, \dots, |v_n\rangle \}$ basis of V .

We can define $A \in L(V, \mathbb{C}^n)$

$$\begin{aligned} A|v\rangle &= A \left(\sum_i a_i |v_i\rangle \right) \\ &= \sum_i a_i \underbrace{A|v_i\rangle}_{\text{define to be } |e_i\rangle} = \begin{pmatrix} a_1 \\ \vdots \\ a_i \\ \vdots \\ a_n \end{pmatrix} \leftarrow \text{it's} \\ &= \sum_i a_i |e_i\rangle \end{aligned}$$

We can define $B \in L(\mathbb{C}^n, V)$

$$B|e_i\rangle = |v_i\rangle$$

Observe that $AB = I_{\mathbb{C}^n}$ and $BA = I_V$.

Therefore $V \cong \mathbb{C}^n$.

Matrix representation

Let $\{ |v_1\rangle, \dots, |v_n\rangle \}$ basis for V

$\{ |w_1\rangle, \dots, |w_m\rangle \}$ basis for W

1) A linear operator $A: V \rightarrow W$ can be represented as a $n \times m$ -matrix with entries A_{ij} :

$$A|v_j\rangle = \sum_i A_{ij} |w_i\rangle$$

Ex $V = \mathbb{C}^2$ with $\{|0\rangle, |1\rangle\}$

$$\cdot X: \mathbb{C}^2 \rightarrow \mathbb{C}^2 \quad \begin{array}{l} |0\rangle \mapsto |1\rangle \\ |1\rangle \mapsto |0\rangle \end{array}$$

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

$$\cdot Z: \mathbb{C}^2 \rightarrow \mathbb{C}^2 \quad \begin{array}{l} |0\rangle \mapsto |0\rangle \\ |1\rangle \mapsto -|1\rangle \end{array}$$

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

$$\cdot Y: \mathbb{C}^2 \rightarrow \mathbb{C}^2 \quad \begin{array}{l} |0\rangle \mapsto i|1\rangle \\ |1\rangle \mapsto -i|0\rangle \end{array}$$

$$Y = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix}$$

Note that $Y = iXZ$

The matrices $\{I, X, Y, Z\}$ are called the Pauli matrices.

Alternate notation:

$$\sigma_0 = I, \quad \sigma_1 = \sigma_x = X, \quad \sigma_2 = \sigma_y = Y, \quad \sigma_3 = \sigma_z = Z$$

2) Composites of linear operators correspond to matrix multiplication

$B: V \rightarrow U$ and $A: U \rightarrow V$ we have

$$(AB)_{ij} = \sum_{k=1}^{\dim U} A_{ik} B_{kj}$$

Notation When $V = W$ we write $L(V) = L(V, V)$

Inner products

1) A function $(\cdot, \cdot) : V \times V \rightarrow \mathbb{C}$ is called an inner product if

a) (\cdot, \cdot) is linear in the second argument

$$(|v\rangle, \sum \lambda_i |w_i\rangle) = \sum \lambda_i (|v\rangle, |w_i\rangle)$$

b) Conjugate symmetry

$$(|v\rangle, |w\rangle) = (|w\rangle, |v\rangle)^*$$

Notation For $z \in \mathbb{C}$ we write z^* for complex conjugate.

c) Positive definiteness

$$(|v\rangle, |v\rangle) \geq 0 \text{ with equality if and only if } |v\rangle = 0.$$

Bra notation

For $v \in \mathbb{C}^n$ with $v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$ we write

$$v^* = (v_1^* \dots v_n^*)$$

We will denote v^* as follows:

$$\langle v| = (v_1^* \dots v_n^*)$$

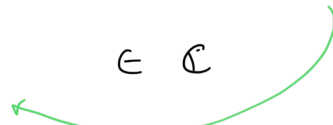
Ex

For $y, z \in \mathbb{C}^n$ define

$$(y, z) = \sum_i y_i^* z_i \in \mathbb{C}$$

This gives an inner product on \mathbb{C}^n .

$\} |e_i\rangle$



In bra-ket notation (y_i) is denoted by

$$\langle y | z \rangle = (y_1^* \dots y_n^*) \begin{pmatrix} z_1 \\ \vdots \\ z_n \end{pmatrix}$$

2) A vector space V together with an inner product is called an inner product space.
Also called finite-dimensional Hilbert space.

For $v, w \in V$ we write

$$\langle v | w \rangle \quad \text{for} \quad (v, w)$$

3) Norm of $|v\rangle$ is defined by

$$\|v\| = \| |v\rangle \| = \sqrt{\langle v | v \rangle}$$

$|v\rangle$ is called normalized, or unit vector, if

$$\|v\| = 1$$

We can normalize a vector

$$|v\rangle \mapsto \frac{1}{\|v\|} |v\rangle \quad (\text{normalized})$$

4) A set $\{|v_1\rangle, \dots, |v_k\rangle\}$ of vectors is called

a) orthogonal if

$$\langle v_i | v_j \rangle = 0 \quad \forall i, j \quad i \neq j$$

b) orthonormal if

$$\langle v_i | v_j \rangle = \delta_{ij} = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$$

$\forall i, j$

Gram-Schmidt procedure

Let $\{ |w_1\rangle, \dots, |w_n\rangle \}$ be a basis for an inner product space V .

The Gram-Schmidt process gives an orthonormal basis $\{ |v_1\rangle, \dots, |v_n\rangle \}$ as follows:

$$a) \quad |v_1\rangle = \frac{1}{\|w_1\|} |w_1\rangle$$

$$b) \quad \text{for } 1 \leq k \leq n-1$$

$$|v_{k+1}\rangle = \frac{|w_{k+1}\rangle - \sum_{i=1}^k \langle v_i | w_{k+1} \rangle |v_i\rangle}{\| |w_{k+1}\rangle - \sum_{i=1}^k \langle v_i | w_{k+1} \rangle |v_i\rangle \|}$$

Convenient notation

1) Instead of writing $\{ |v_1\rangle, \dots, |v_n\rangle \}$ for an orthonormal basis we will sometimes write $\{ |i\rangle, \dots, |n\rangle \}$

a) $\{ |v_i\rangle \}_{i=1}^n$ denotes an isomorphism

$$V \cong \mathbb{C}^n \quad \text{and} \quad \{ |i\rangle \}_{i=1}^n = \{ |e_i\rangle \}_{i=1}^n$$

b) $\{ |i\rangle \}_{i=1}^n$ the vectors, $|i\rangle$ stand for vectors in V , such as $|v_i\rangle$.

With this notation

$$|v\rangle = \sum_{i=1}^n v_i |i\rangle$$

$$|w\rangle = \sum_{i=1}^n w_i |i\rangle$$

" For \mathbb{C}^2 we write

$$\{ |1\rangle, |2\rangle \} \text{ for}$$

$$\{ e_1, e_2 \}."$$

$$\begin{aligned}
\langle w | v \rangle &= \left(\sum w_j |j\rangle, \sum v_i |i\rangle \right) \\
&= \sum_{i,j} w_j^* v_i \underbrace{\langle j | i \rangle}_{\delta_{ij}} \\
&= \sum_i w_i^* v_i \\
&= (w_1^* \dots w_n^*) \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}
\end{aligned}$$

2) We can think of $\langle v |$ as a linear operator in $L(V, \mathbb{C})$

$$\langle v | : V \rightarrow \mathbb{C} \quad |v'\rangle \mapsto \langle v | v' \rangle$$

More generally, for $v \in V$, $w \in W$

$|w\rangle \langle v |$ can be regarded as a linear operator

in $L(V, W)$:

$$|w\rangle \langle v | : V \rightarrow W \quad |v'\rangle \mapsto \langle v | v' \rangle |w\rangle$$

This is called the outer product of $|v\rangle$ and $|w\rangle$.

We can take linear combinations of

$$\left\{ |w_j\rangle \langle v_i | \mid 1 \leq j \leq m, 1 \leq i \leq n \right\}$$

to define $V \rightarrow W$

$$\left(\sum_{i,j} a_{ij} |w_j\rangle \langle v_i| \right) |v'\rangle = \sum_{i,j} a_{ij} \langle v_i | v' \rangle |w_j\rangle$$

Ex 1) Consider \mathbb{C}^n with $\{|e_i\rangle\}$

$$\sum_i |e_i\rangle \langle e_i| : \mathbb{C}^n \rightarrow \mathbb{C}^n$$

$$\left(\sum_i |e_i\rangle \langle e_i| \right) |v\rangle = \sum_i \underbrace{\langle e_i | v \rangle}_{v_i} |e_i\rangle = |v\rangle$$

$$\text{Therefore } \sum_i |e_i\rangle \langle e_i| = I_{\mathbb{C}^n}$$

This is called the completeness relation.

2) Let $\{|v_i\rangle\}$ and $\{|w_j\rangle\}$ be orthonormal bases for V and W , and $A: V \rightarrow W$.

$$A = I_W A I_V$$

$$= \sum_j |w_j\rangle \langle w_j| A \sum_i |v_i\rangle \langle v_i| \quad (\dots \dots) \begin{pmatrix} \vdots \\ A_{ij} \\ \vdots \end{pmatrix} \begin{pmatrix} \vdots \\ v_i \\ \vdots \end{pmatrix}$$

$$= \sum_{i,j} \underbrace{\langle w_j | A | v_i \rangle}_{A_{ij}} |w_j\rangle \langle v_i| \quad \parallel \quad (|w_j\rangle, A |v_i\rangle)$$

This is a typical application of completeness relation.

Cauchy - Schwarz inequality

Let $v, w \in V$. Then

$$\|v\| = \sqrt{\langle v|v \rangle}$$

$$\|w\| = \sqrt{\langle w|w \rangle}$$

$$|\langle v|w \rangle| \leq \|v\| \cdot \|w\|$$

with equality if and only if one of v, w is a scalar multiple of the other.

Proof Using Gram - Schmidt procedure we can construct an orthonormal basis

$\{ |v_i\rangle \}_{i=1}^n$ such that

$$|v_i\rangle = \frac{1}{\|w\|} |w\rangle$$

Then

$$\langle v|v \rangle \langle w|w \rangle = \sum_i \langle v|v_i\rangle \langle v_i|v \rangle \langle w|w \rangle$$

$$\geq \langle v|v_i\rangle \langle v_i|v \rangle \langle w|w \rangle$$

$$= \langle v|w \rangle \langle w|v \rangle \frac{\langle w|w \rangle}{\|w\|^2}$$

$$= |\langle v|w \rangle|^2$$



Eigenvectors and eigenvalues

1) An eigenvector of a linear operator $A \in L(V)$ is a non-zero vector $|v\rangle$ such that

$$A |v\rangle = \lambda |v\rangle \quad \text{for some } \lambda \in \mathbb{C}.$$

λ is called the eigenvalue corresponding to $|v\rangle$

We will write V_λ for the eigenspace corresponding to λ :

$$V_\lambda = \{ v \in V \mid A|v\rangle = \lambda|v\rangle \}$$

λ is said to be degenerate if $\dim V_\lambda > 1$.

Fact Every linear operator has at least one eigenvalue and a corresponding eigenvector since the characteristic polynomial

$$c(\lambda) = \det(A - \lambda I_V)$$

has at least one root.

2) A is diagonalizable if

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

where $|i\rangle$ form an orthonormal set of eigenvectors with corresponding eigenvalue λ_i .

Ex 1) $z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = |0\rangle\langle 0| - |1\rangle\langle 1|$

2) $x = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ $x: \begin{matrix} |0\rangle \mapsto |1\rangle \\ |1\rangle \mapsto |0\rangle \end{matrix}$

$$|+\rangle = \frac{|0\rangle + |1\rangle}{\sqrt{2}}, \quad |-\rangle = \frac{|0\rangle - |1\rangle}{\sqrt{2}}$$

$$x|+\rangle = |+\rangle, \quad x|-\rangle = -|-\rangle$$

$$x = |+\rangle\langle +| - |-\rangle\langle -|$$


Adjoint

1) The adjoint (Hermitian conjugate) of $A \in L(V)$ is the linear operator

$A^\dagger \in L(V)$ defined by the equation

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

for all $|v\rangle, |w\rangle \in V$.



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

In matrix representation

$$A^\dagger = (A^*)^T$$

i.e. transpose of the conjugate.

By convention $|v\rangle^\dagger = \langle v|$

Important classes of operators

2) Normal operators: $A \in L(V)$ is normal if

$$A A^\dagger = A^\dagger A$$

We will see that normal operators are diagonalizable

2) Hermitian operators: $A \in L(V)$ is Hermitian if $A = A^\dagger$. Note that Hermitian operators are normal.

Notation: $\text{Herm}(V)$

3) Positive operators (positive semidefinite):

$A \in L(V)$ is positive if

$$\langle v | A | v \rangle \in \mathbb{R}_{\geq 0} \quad (\text{real and non-negative})$$

for all $v \in V$

Notation $\text{Pos}(V)$

4) Positive definite operators

$A \in L(V)$ is positive definite if

$$\langle v | A | v \rangle \in \mathbb{R}_{> 0} \quad (\text{real and positive})$$

for all non-zero $v \in V$.

Notation $\text{Pd}(V)$

Rem (3) & (4) are Hermitian.

$$\text{Pos}(V) = \{ B^+ B \mid B \in L(V) \}$$

$$\text{Pd}(V) = \{ A \in \text{Pos}(V) \mid \det A \neq 0 \}$$

5) Projection operators: $P \in \text{Pos}(V)$ is a projection operator if $P^2 = P$.

Notation $\text{Pos}(V)$

The image of P will be denoted by V_P .

More: a) Given $W \subset V$ subspace we can construct a projector P with $\text{im}(P) = W$.

Let $\{ |i\rangle \}_{i=1}^{\dim(W)}$ orthonormal basis for W .

Extend this to orthonormal basis for V

$$\{ |i\rangle \}_{i=1}^{\dim(V)}$$

Define

$$P = \sum_{i=1}^{\dim W} |i\rangle \langle i|$$

$$\begin{aligned} P|v\rangle &= \left(\sum_{i=1}^{\dim W} |i\rangle \langle i| \right) \left(\sum_{j=1}^{\dim V} v_j |j\rangle \right) \\ &= \sum_{i=1}^{\dim W} v_i |i\rangle \in W \end{aligned}$$

Therefore $\text{im}(P) = W$. ($W = V_P$)

b) Conversely, given P we can diagonalize

$$P = \sum_{i=1}^{\dim V} \delta_i |i\rangle\langle i|$$

Since $P^2 = P$ and $\{|i\rangle\}$ orthonormal
we have $\delta_i = 0$ or 1 .

$$\left(P^L = \sum_i \delta_i^2 |i\rangle\langle i| \right)$$

$\delta_i \geq 0$

Then $P = \sum_i |i\rangle\langle i|$

This gives a bijection

$$\begin{array}{ccc} \text{Proj}(V) & \longrightarrow & \{W \subset V \mid W \text{ subspace}\} \\ P & \longmapsto & V_P \end{array}$$

The orthogonal complement of P is the projector

$$Q = \mathbb{I}_V - P = \sum_{i=\dim(V_P)+1}^{\dim V} |i\rangle\langle i|$$

V_Q is called the orthogonal complement of V_P .

6) Unitary operators: $U \in L(V)$ is unitary if $U^*U = I_V$.

Notation $U(V)$

Rem Note that

$U^*U = I_V$ implies that $\det U \neq 0$ and therefore $U^{-1} = U^*$. This implies that $UU^* = I_V$ as well.

We have

$$\begin{aligned} \langle Uv, Uw \rangle &= \langle v | U^*U | w \rangle \\ &= \langle v | w \rangle \quad \forall v, w \in V. \end{aligned}$$

Even $U \in U(V)$ is diagonalizable and its eigenvalues λ satisfy $|\lambda| = 1$.

Rem.. The adjoint can also be defined for $A \in L(V, W)$ similarly by the eqn

$$(w, Av) = (A^*w, v)$$

$$\forall v \in V, w \in W.$$

An operator $A \in L(V, V)$ is a normal if

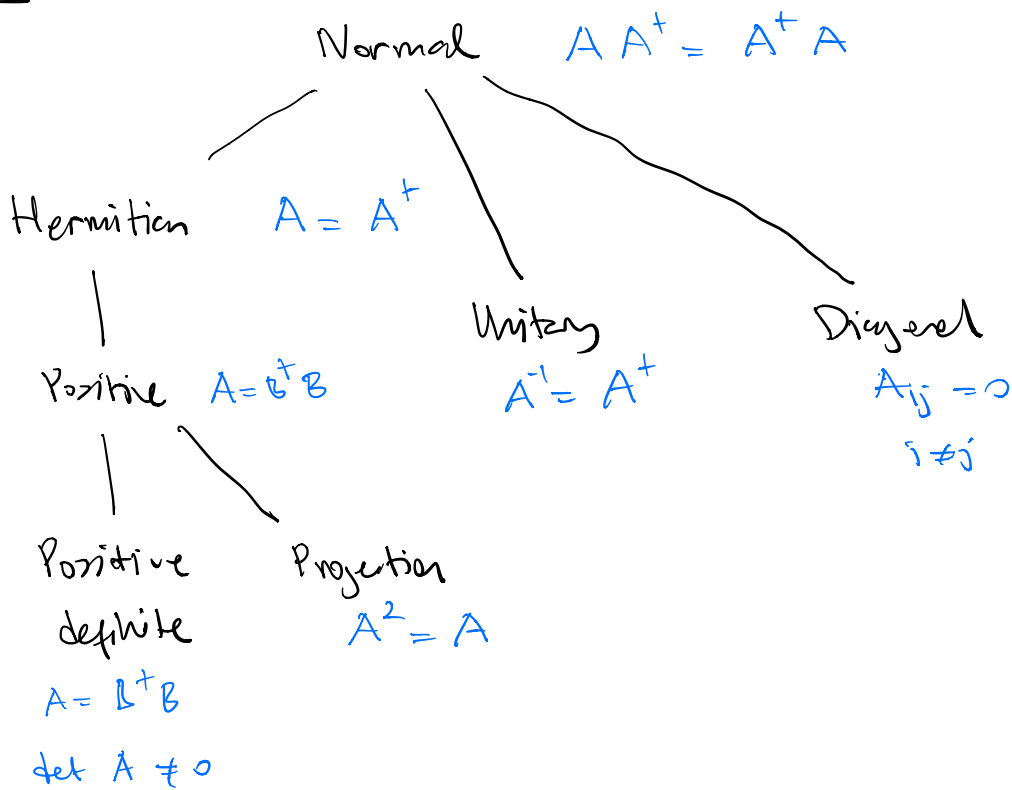
$$A^*A = AA^*.$$

Notation $U(V, W)$. Note that $U(V, V) = U(V)$.

7) Diagonal operators: $A \in L(\mathbb{C}^n)$ is diagonal if $A_{ij} = 0$ for all $1 \leq i, j \leq n$ and $i \neq j$.

Rem $A \in L(\mathbb{C}^n)$ is diagonalizable if and only if there exists $U \in U(\mathbb{C}^n)$ such that $U A U^+$ is diagonal.

Picture



Theorem (Spectral decomposition)

$A \in L(V)$ is diagonalizable if and only if A is normal.

Proof (Easy direction \Rightarrow)

If A is diagonalizable then

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

where $\{|v_i\rangle\}$ is an orthonormal basis consisting of eigenvectors.

Then

$$A^\dagger A = \left(\sum_j \lambda_j^* |v_j\rangle\langle v_j| \right) \left(\sum_i \lambda_i |v_i\rangle\langle v_i| \right)$$

$$= \sum_{j,i} \lambda_j^* \lambda_i \underbrace{\langle v_j | v_i \rangle}_{\delta_{ji}} |v_j\rangle\langle v_i|$$

$$= \sum_i \lambda_i^* \lambda_i |v_i\rangle\langle v_i|$$

$$= \sum_i \lambda_i \lambda_i^* |v_i\rangle\langle v_i|$$

$$= \sum_{j,i} \lambda_i \lambda_j^* \underbrace{\langle v_i | v_j \rangle}_{\delta_{ij}} |v_i\rangle\langle v_j|$$

$$= \left(\sum_i \lambda_i |v_i\rangle \langle v_i| \right) \left(\sum_j \lambda_j^* |v_j\rangle \langle v_j| \right)$$

$$= AA^\dagger$$

(\Leftarrow)

We will do induction on $d = \dim V$.

For $d=1$ result is clear: Every $A \in L(V)$

\square diagonalizable and normal.

Let λ be an eigenvalue of A .

Let P be the projector onto the eigenspace $V_\lambda (= V_P)$.

Let $Q = I_V - P$. $\curvearrowright I_V = P + Q$

Then

$$A = I_V A I_V$$

$$= PAP + \underbrace{QAP + PAQ}_{\text{Goal: } 0} + QAQ$$

1) $QAP = 0$

$$AP|v\rangle = \lambda P|v\rangle \in V_P = V_\lambda$$

$$QAP|v\rangle = \lambda QP|v\rangle = 0$$

$$2) \quad PAQ = 0$$

$$\text{let } w \in V_p = V_\lambda$$

$$AA^+|w\rangle = A^+A|w\rangle = \lambda A^+|w\rangle$$

(normality)

$$\text{Therefore } A^+|w\rangle \in V_\lambda = V_p$$

Similar to (1) we have

$$QA^+P = 0 \quad (2*)$$

$$QA^+P|v\rangle = 0$$

$\underbrace{\hspace{10em}}_{V_p = V_\lambda}$
 $\underbrace{\hspace{10em}}_{V_p}$

Take the adjoint of (2*)

$$(QA^+P)^+ = 0 \quad \rightsquigarrow \quad P^+(A^+)^+Q^+ = 0$$

$$PAQ = 0$$

Combining (1) & (2)

$$A = PAP + QAQ$$

3) $Q A Q$ is normal.

By (1) & (2*)

$$Q A = Q A (P + Q) \stackrel{(1)}{=} Q A Q \quad (3.1)$$

$$Q A^\dagger = Q A^\dagger (P + Q) = Q A^\dagger Q \quad (3.2)$$

$$\begin{aligned} (Q A Q) (Q A^\dagger Q) &\stackrel{3.1}{=} Q A Q A^\dagger Q \\ &\stackrel{3.2}{=} Q A A^\dagger Q \\ &= Q A^\dagger A Q \\ &\stackrel{3.2}{=} Q A^\dagger Q A Q \\ &= (Q A^\dagger Q) (Q A Q) \\ &\quad (Q = Q^*) \end{aligned}$$

Here $Q A Q \in L(V_Q)$ where $\dim V_Q < d$

By induction $Q A Q$ is diagonalizable, say

with respect to $\left\{ |w_i\rangle \right\}_{i=1}^{\dim V_Q}$

Let $\left\{ |v_i\rangle \right\}_{i=1}^{\dim V_P}$ an orthonormal basis for V_P .

Then A is diagonal in

$$\left\{ |v_i\rangle, |w_j\rangle \right\}_{i,j}$$



Tensor product

1) Given vector spaces V and W the tensor product, denoted by $V \otimes W$, is the vector space with basis given by the symbols,

$|v\rangle \otimes |w\rangle \quad v \in V \quad \text{and} \quad w \in W$
modulo the relations

$$\begin{aligned} \text{a) } \alpha(|v\rangle \otimes |w\rangle) &= (\alpha|v\rangle) \otimes |w\rangle \\ &= |v\rangle \otimes (\alpha|w\rangle) \end{aligned}$$

$$\begin{aligned} \text{b) } (|v_1\rangle + |v_2\rangle) \otimes |w\rangle \\ = |v_1\rangle \otimes |w\rangle + |v_2\rangle \otimes |w\rangle \end{aligned}$$

$$\begin{aligned} \text{c) } |v\rangle \otimes (|w_1\rangle + |w_2\rangle) \\ = |v\rangle \otimes |w_1\rangle + |v\rangle \otimes |w_2\rangle \end{aligned}$$

i.e. $V \otimes W$ is the quotient vector space obtained this way.

2) Let $\{|v_i\rangle\}_i$ and $\{|w_j\rangle\}_j$
bases for V and W .

Then $V \otimes W$ has basis

$$\{|v_i\rangle \otimes |w_j\rangle\}_{i,j}$$

since

$$|v\rangle \otimes |w\rangle = \left(\sum_i a_i |v_i\rangle \right) \otimes \left(\sum_j b_j |w_j\rangle \right)$$

$$\stackrel{(a_i, b_j)}{=} \sum_{i,j} a_i b_j |v_i\rangle \otimes |w_j\rangle$$

Notation

We will also write $|v\rangle |w\rangle$,
 $|v, w\rangle$, $|v, w\rangle$ for $|v\rangle \otimes |w\rangle$.

Ex a) $V = W = \mathbb{C}^2$ with $\{|0\rangle, |1\rangle\}$

$V \otimes W$ has basis $\{|00\rangle, |01\rangle, |10\rangle, |11\rangle\}$

Rem: $\dim V = n$ and $\dim W = m$

Then $\dim(V \otimes W) = n \cdot m$

b) $V \otimes V \otimes V$ with $V = \mathbb{C}^2$

has basis $\{ |000\rangle, |001\rangle, |011\rangle, \dots \}$

3) Given $A \in L(V, W)$ and $B \in L(V', W')$
we can define

$$(A \otimes B) \left(\underbrace{|v\rangle}_{\in V} \otimes \underbrace{|v'\rangle}_{\in V'} \right) = \underbrace{A|v\rangle}_{\in W} \otimes \underbrace{B|v'\rangle}_{\in W'}$$

and obtain $A \otimes B \in L(V \otimes V', W \otimes W')$

by extending linearly.

This defines a linear operator

$$L(V, W) \otimes L(V', W') \longrightarrow L(V \otimes V', W \otimes W')$$

which turns out to be an isomorphism.

That is any linear operator

$C \in L(V \otimes V', W \otimes W')$ can be
expressed as

$$C = \sum_i c_i A_i \otimes B_i$$

for some $A_i \in L(V, W)$, $B_i \in L(V', W')$

$$\underline{\text{Ex}} \quad a) \quad X: \mathbb{C}^2 \rightarrow \mathbb{C}^2 \quad \begin{array}{l} |0\rangle \mapsto |1\rangle \\ |1\rangle \mapsto |0\rangle \end{array}$$

$$X \otimes X: \mathbb{C}^2 \otimes \mathbb{C}^2 \rightarrow \mathbb{C}^2 \otimes \mathbb{C}^2$$

$$\begin{aligned} X \otimes X (|0\rangle \otimes |0\rangle) &= X|0\rangle \otimes X|0\rangle \\ &= |1\rangle \otimes |1\rangle \end{aligned}$$

$$\begin{array}{ccc} \begin{array}{l} \downarrow \downarrow \\ |00\rangle \\ |01\rangle \\ |10\rangle \\ |11\rangle \end{array} & \xrightarrow{X \otimes X} & \begin{array}{l} |11\rangle \\ |10\rangle \\ |01\rangle \\ |00\rangle \end{array} \end{array}$$

$$b) \quad Z: \mathbb{C}^2 \rightarrow \mathbb{C}^2 \quad \begin{array}{l} |0\rangle \mapsto |0\rangle \\ |1\rangle \mapsto -|1\rangle \end{array}$$

$$Z \otimes X: \mathbb{C}^2 \otimes \mathbb{C}^2 \rightarrow \mathbb{C}^2 \otimes \mathbb{C}^2$$

$$\begin{array}{ccc} |00\rangle & \mapsto & |01\rangle \\ |01\rangle & \mapsto & |00\rangle \\ |10\rangle & \mapsto & -|11\rangle \\ |11\rangle & \mapsto & -|10\rangle \end{array}$$

" unit complex numbers are also called phase "

Remark i) $Z \otimes X$ is not the same as ZX .

$$Z \otimes X: \mathbb{C}^2 \otimes \mathbb{C}^2 \rightarrow \mathbb{C}^2 \otimes \mathbb{C}^2$$

$$ZX: \mathbb{C}^2 \rightarrow \mathbb{C}^2$$

$$ii) \quad Z \otimes X \neq X \otimes Z.$$

4) Let V and W be inner product spaces.

We can define an inner product on $V \otimes W$:

$$\left(\sum_i a_i |v_i w_i\rangle, \sum_j b_j |v'_j w'_j\rangle \right) = \sum_{ij} a_i^* b_j \langle v_i | v'_j \rangle \langle w_i | w'_j \rangle$$

$$v_i, v'_j \in V \quad \text{and} \quad w_i, w'_j \in W$$

Ex $V = W = \mathbb{C}^2$ with $\{|0\rangle, |1\rangle\}$

$$x, y \in V \otimes W$$

$$\langle x | y \rangle = \left(x_{00}^* \langle 00| + x_{01}^* \langle 01| + x_{10}^* \langle 10| + x_{11}^* \langle 11| \right) \left(y_{00} |00\rangle + y_{01} |01\rangle + y_{10} |10\rangle + y_{11} |11\rangle \right)$$

$$= x_{00}^* y_{00} + x_{01}^* y_{01} + x_{10}^* y_{10} + x_{11}^* y_{11}.$$

5) Matrix representation of $A \otimes B$:

$$\begin{pmatrix} A_{11} & \dots & A_{1n} \\ \vdots & & \vdots \\ A_{m1} & \dots & A_{mn} \end{pmatrix} \otimes \begin{pmatrix} B_{11} & \dots & B_{1q} \\ \vdots & & \vdots \\ B_{p1} & \dots & B_{pq} \end{pmatrix}$$

$$= \begin{pmatrix} \boxed{A_{11} B} & \dots & \boxed{A_{1n} B} \\ \vdots & & \vdots \\ \boxed{A_{m1} B} & \dots & \boxed{A_{mn} B} \end{pmatrix}$$

$$= \begin{pmatrix} A_{11} B_{11} & \dots & A_{11} B_{1q} \\ \vdots & & \vdots \\ A_{m1} B_{p1} & \dots & A_{m1} B_{pq} \end{pmatrix}$$

This is called the Kronecker product.

Ex

$$X \otimes Z = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

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

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

similarly for the other basis vectors.

Operator functions

- 1) Let A be a normal operator and
 $f: \mathbb{C} \rightarrow \mathbb{C}$ be a function.
By $f(A)$ we mean

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

where $\{|v_i\rangle\}$ is an orthonormal basis
of eigenvectors.

Ex a) $\exp(\alpha X) = \exp(\alpha (|+\rangle\langle +| - |-\rangle\langle -|))$

$$X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = e^{\alpha} |+\rangle\langle +| + e^{-\alpha} |-\rangle\langle -|$$

$$|+\rangle = (|0\rangle + |1\rangle) / \sqrt{2}$$

$$|-\rangle = (|0\rangle - |1\rangle) / \sqrt{2}$$

$$\alpha \in \mathbb{C}.$$

- b) Let $v = (v_1, v_2, v_3) \in \mathbb{R}^3$ unit
vector.

$$\text{Let } \sigma = (\sigma_1, \sigma_2, \sigma_3) \quad \begin{array}{l} \sigma_1 = X \\ \sigma_2 = Y \end{array}$$

We write

$$\sigma_3 = Z$$

$$v \cdot \sigma = \sum_{i=1}^3 v_i \sigma_i$$

$$\exp(i\theta v \cdot \sigma) = \cos \theta \mathbb{I} + i \sin \theta v \cdot \sigma$$

$$\theta \in \mathbb{R}$$

Idea : 1) Since v unit vector $v \cdot v = 1$
has eigenvalues ± 1 .

2) Euler's formula

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

No need to find eigenvectors.

2) Trace of a matrix A is defined by

$$\text{Tr } A = \sum_i A_{ii}$$

2.1) It satisfies the cyclicity property

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

Using this we can define the trace of
a linear operator $A \in L(V)$ by

$$\text{Tr}(A) = \sum_i A_{ii}$$

(A_{ij}) matrix representation
for A .

Note that any two matrix representations
 A and A' are related by $A' = B A B^{-1}$
so that

$$\begin{aligned} \text{Tr}(A') &= \text{Tr}(B A B^{-1}) = \text{Tr}(B^{-1} B A) \\ &= \text{Tr}(A). \end{aligned}$$

2.2) Trace is the unique linear operator

$$\text{Tr}: L(V) \rightarrow \mathbb{F} \quad \text{such that}$$

$$\text{Tr}(|v\rangle\langle u|) = \langle u|v\rangle.$$

$$|v\rangle\langle u| : V \rightarrow V$$

$$|v'\rangle \mapsto \langle u|v'\rangle |v\rangle$$

2.3) We can define an inner product on $L(V)$

$$(A, B) = \text{Tr}(A^\dagger B)$$

called the trace inner product.

Commutator and anti-commutator

The commutator of A and B is defined by

$$[A, B] = AB - BA$$

i.e. $[A, B] = 0$ if and only if $AB = BA$.

In this case we say A commutes with B .

The anti-commutator of A and B :

$$\{A, B\} = AB + BA$$

Ex: $[G_i, G_i] = 0$

$$\{G_i, G_j\} = 0 \quad i \neq j$$

Theorem (Simultaneous diagonalization) $[A, B] = AB - BA$

Let A and B be Hermitian operators.

Then $[A, B] = 0$ if and only if A and B are simultaneously diagonalizable i.e.

there exists an orthonormal basis in which A and B are both diagonal.

Proof (\Leftarrow)

$$\begin{aligned}
 [A, B] &= \left[\sum_i \lambda_i^A |v_i\rangle\langle v_i|, \sum_j \lambda_j^B |v_j\rangle\langle v_j| \right] \\
 &= \sum_i \lambda_i^A \lambda_i^B |v_i\rangle\langle v_i| - \sum_i \lambda_i^B \lambda_i^A |v_i\rangle\langle v_i| \\
 &= 0.
 \end{aligned}$$

$\begin{pmatrix} \lambda_1^A & & 0 \\ & \ddots & \\ 0 & & \lambda_n^A \end{pmatrix}$
 $\begin{pmatrix} \lambda_1^B & & 0 \\ & \ddots & \\ 0 & & \lambda_n^B \end{pmatrix}$

(\Rightarrow) a) Key idea: if λ is an eigenvalue of A and V_λ eigenspace then

$$B|v\rangle \in V_\lambda \quad \forall |v\rangle \in V_\lambda$$

More explicitly, let $\left\{ |\lambda, j\rangle \right\}_{j=1}^{\dim V_\lambda}$ be an orthonormal basis for V_λ . Then

$$\underbrace{A(B|\lambda, j\rangle)} = \underbrace{B(A|\lambda, j\rangle)}_{\lambda|\lambda, j\rangle} = \lambda \underbrace{(B|\lambda, j\rangle)}_{\uparrow V_\lambda}$$

b) Let P_λ be the projector onto V_λ .

Let $B_\lambda = P_\lambda B P_\lambda$.

We have

$$(B_\lambda)^\dagger = P_\lambda B^\dagger P_\lambda = P_\lambda B P_\lambda = B_\lambda$$

Therefore B_λ is diagonalizable. For an eigenvalue p of B_λ let $V_{\lambda,p}$ be the corresponding eigenspace in V_λ .

Let $\{| \lambda, p, k \rangle\}_k$ be an orthonormal basis of $V_{\lambda,p}$.

c) Note that

$$\begin{aligned} B_\lambda | \lambda, p, k \rangle &= P_\lambda B P_\lambda | \lambda, p, k \rangle \\ \underbrace{P_\lambda B P_\lambda}_{P} | \lambda, p, k \rangle &= P_\lambda B | \lambda, p, k \rangle \\ &= p | \lambda, p, k \rangle \end{aligned}$$

thus

$$B | \lambda, p, k \rangle = p | \lambda, p, k \rangle.$$

Therefore $\{| \lambda, p, k \rangle\}_{\lambda, p, k}$ is an orthonormal set of eigenvectors of both A and B . \square

$$\begin{array}{cccc} V & V_\lambda & V_{\lambda,p} & \{| \lambda, p, k \rangle\} \\ & A & B & \end{array}$$

Ex $[X \otimes X, Z \otimes Z] = 0$

$Z \otimes Z$ is diagonal in

$$\begin{array}{c}
 X \\
 \begin{array}{c}
 100 \rangle \\
 101 \rangle \\
 110 \rangle \\
 111 \rangle
 \end{array}
 \end{array}
 \begin{array}{c}
 | \\
 - \\
 - \\
 - \\
 |
 \end{array}$$

V_+ $\{ |00\rangle, |11\rangle \}$ $\left\{ \frac{|00\rangle + |11\rangle}{\sqrt{2}}, \frac{|00\rangle - |11\rangle}{\sqrt{2}} \right\}$

V_- $\{ |01\rangle, |10\rangle \}$ $\left\{ \frac{|01\rangle + |10\rangle}{\sqrt{2}}, \frac{|01\rangle - |10\rangle}{\sqrt{2}} \right\}$

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

diagonal both $Z \otimes Z$ and $X \otimes X$.

V_X $V_{X,IP}$ $|X, IP\rangle$

V_+ $V_{+,1}$ \leftarrow 1-dim

$|1, 1\rangle$

Recall $X : \mathbb{C}^2 \rightarrow \mathbb{C}^2$

$$\begin{array}{c}
 10 \rangle \mapsto 11 \rangle \\
 11 \rangle \mapsto 10 \rangle
 \end{array}$$

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

$X \otimes X \left(\frac{|00\rangle + |11\rangle}{\sqrt{2}} \right) = X \otimes X \left(\frac{|0\rangle \otimes |0\rangle + |1\rangle \otimes |1\rangle}{\sqrt{2}} \right)$

$$\begin{aligned}
&= \frac{(X|0\rangle) \otimes (X|0\rangle) + (X|1\rangle) \otimes (X|1\rangle)}{r_2} \\
&= \frac{|11\rangle \otimes |11\rangle + |00\rangle \otimes |00\rangle}{r_1} \\
&= \frac{|111\rangle + |000\rangle}{r_1}
\end{aligned}$$

Do this with matrices

$$V_1 = \text{span} \{ |00\rangle, |11\rangle \}$$

matrix of $X \otimes X$ is $\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$

$$(X \otimes X: V_1 \rightarrow V_1)$$

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \frac{|1\rangle\langle 1| - |0\rangle\langle 0|}{2}$$

$$|1\rangle = \frac{|00\rangle + |11\rangle}{r_2}$$

$$|0\rangle = \frac{|00\rangle - |11\rangle}{r_1}$$

$$V_1 \xrightarrow{\mathcal{U}} \mathbb{C}^2$$

$$|00\rangle \mapsto |0\rangle$$

$$|11\rangle \mapsto |1\rangle$$

$$X \otimes X \mapsto X$$

$$\begin{array}{ccc}
V_1 & \xrightarrow{\mathcal{U}} & \mathbb{C}^2 \\
X \otimes X \downarrow & \simeq & \downarrow X \\
V_1 & \xrightarrow{\mathcal{U}} & \mathbb{C}^2
\end{array}$$

Theorem (Polar decomposition)

Let $A \in L(V)$.

Then there exist, $U \in U(V)$ and $J, K \in P_+(V)$ such that

$$A = \underbrace{UJ}_{\text{left polar decomposition}} = \underbrace{KU}_{\text{right polar decomposition}}$$

"Consider $V = \mathbb{C}$
 $L(V) = \mathbb{C}$ "

where $J = \sqrt{A^+A}$ and $K = \sqrt{AA^+}$.

If A is invertible then U is unique.

Rem Here $A^+A \in P_+(V)$. $J = \sqrt{A^+A}$ is the unique positive operator such that

$$J^2 = A^+A. \quad (\text{similarly for } K)$$

Proof a) $J = \sqrt{A^+A}$ is positive so we can write

$$J = \sum_i \lambda_i |l_i\rangle\langle l_i| \quad \lambda_i \in \mathbb{R}_{\geq 0}$$

For $\lambda_i \neq 0$ define

$$|v_i\rangle = \frac{1}{\lambda_i} A |l_i\rangle$$

$|l_i\rangle$: eigenvector of J corresponds to λ_i .

$\{|v_i\rangle\}$ is an orthonormal set:

$$\begin{aligned}
\langle v_j | v_i \rangle &= \langle j | A^\dagger \frac{1}{\lambda_j} \frac{1}{\lambda_i} A | i \rangle \\
&= \frac{1}{\lambda_j \lambda_i} \underbrace{\langle j | A^\dagger A | i \rangle}_{\lambda_i^2} \\
&= \frac{1}{\lambda_j \lambda_i} \underbrace{\langle j | \delta | i \rangle}_{\lambda_i \langle j | i \rangle} = \delta_{ji}
\end{aligned}$$

We can extend this set to an orthonormal basis $\{|v_i\rangle\}$ of V .

$$\text{let } U := \sum_i |v_i\rangle \langle i|.$$

b) b.1) Note that for $\lambda_i \neq 0$

$$\begin{aligned}
U | i \rangle &= \left(\sum_k |v_k\rangle \langle k| \right) \left(\sum_j \delta_{kj} \lambda_j | j \rangle \langle j| \right) | i \rangle \\
&= \lambda_i |v_i\rangle = A | i \rangle
\end{aligned}$$

b.2) For $\lambda_i = 0$

$$U | i \rangle = U \left(\sum_j \delta_{ij} | j \rangle \langle j| \right) | i \rangle = 0$$

Note $A|i\rangle = 0$ ← since

$$\|A|i\rangle\| = \langle i|A^\dagger A|i\rangle = \lambda_i^2 = 0.$$

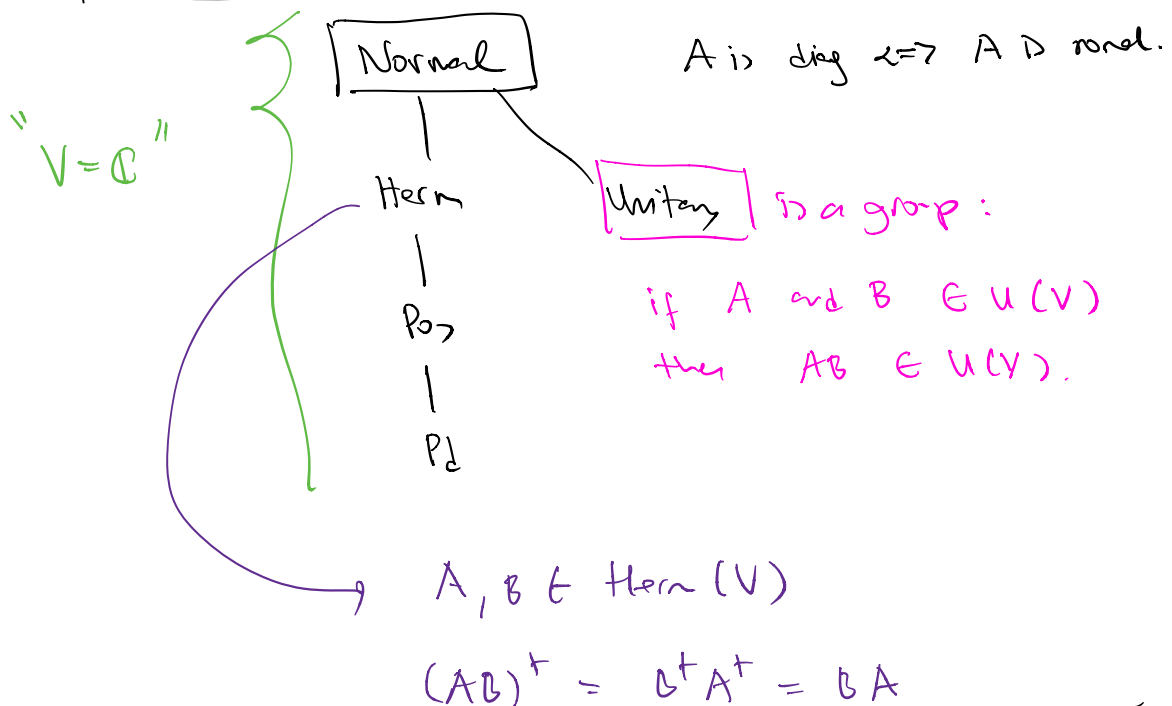
Therefore $A = UJ$.

c) If A is invertible then

$$\begin{aligned} 0 \neq \det A &= \det(UJ) \\ &= \det U \cdot \det J \neq 0 \end{aligned}$$

J is invertible:

$$A = UJ \Rightarrow U = AJ^{-1} \quad \square$$



Corollary (Singular value decomposition)

Let A square matrix.

Then there exists unitary matrices U and V , a diagonal D with non-negative entries such that

$$A = U D V.$$

Proof By the polar decomposition

$$A = W J$$

where W : unitary, $J = \sqrt{A^+ A}$.

Spectral decomposition:

$$J = T D T^+$$

where T : unitary.

Then we have

$$A = W (T D T^+)$$

and we take

$$U = W T \quad \text{and} \quad V = T^+ \quad \square$$

Remark: The diagonal entries in D are called the singular values of A .