

MATRIX ANALYSIS

MA353

Feb 7, 2019

① Complex Numbers

"Traditional"
way to
think

- writing " $a+ib$ " is equiv to writing a pair $(a, b) \in \mathbb{R}^2$
 $a, b \in \mathbb{R}$

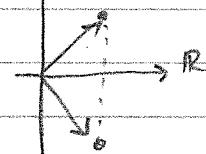
- Addition $(a, b) + (c, d) = (a+c, b+d)$

- Multiplication $(a, b) \cdot (c, d) = (ac - bd, ad + bc)$

↳ same as $(a+ib)(c+id) = (ac - bd) + i(ad + bc)$

Think of i as $\sqrt{-1}$, a multiplication helps
 "better" multiplication than element wise multiplication,
 because there are properties we're familiar with
 in real multiplication.

$"i" \sim (0, 1), "1" \sim (1, 0), "0" \sim (0, 0)$



- Conjugate of $a+ib$ is $a-ib \rightarrow$ reflecting about the real axis.

"New way to think abt complex no."

↳ $a+ib \sim \begin{bmatrix} a & -b \\ b & a \end{bmatrix} = a \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}}_{\text{real-part}} + b \underbrace{\begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}}_{\text{"i"-part}}$

So $(aI + b\overset{\circ}{I})(cI + d\overset{\circ}{I}) \stackrel{?}{=} (cI + d\overset{\circ}{I})(aI + b\overset{\circ}{I})$

$\rightarrow \begin{pmatrix} a & -b \\ b & a \end{pmatrix} \circ \begin{pmatrix} c & -d \\ d & c \end{pmatrix}$ commutes! same as
 complex no. multiplication

Addition works exactly the same way as before. \rightarrow adding matrices.

Taking the adjoint (conjugate) \equiv transpose them.

(2)

$$\begin{pmatrix} a & -b \\ b & a \end{pmatrix} \begin{pmatrix} a & b \\ -b & a \end{pmatrix} = MN^T = \begin{pmatrix} a^2+b^2 & 0 \\ 0 & a^2+b^2 \end{pmatrix} = \underbrace{\det(M)}_{= (a^2+b^2)} \cdot I$$

Similarly $\rightarrow (a+ib)(a-ib) = a^2+b^2$

Note $A \cdot [J] = (\det A) I$

Note, $\det A = 0 \Leftrightarrow a = b = 0$

$A \cdot [J] / \det A = I \Rightarrow [J = A^{-1}] = \text{"reciprocal" of complex no.}$

Note $A^T = A^{-1}$ here if $A = \begin{pmatrix} a & -b \\ b & a \end{pmatrix}$

$$\frac{1}{3+5i} = \frac{3-5i}{3-5i} \cdot \frac{1}{3+5i} = \frac{1}{3^2+5^2} (3-5i)$$

number $\frac{1}{\| \text{length} \|}$ new number..

We will show that there always exists a classical adjoint B
s.t. $AB = (\det A) I$

Geometric interpretation

$$\begin{pmatrix} a & -b \\ b & a \end{pmatrix} \begin{pmatrix} c & d \\ d & c \end{pmatrix} = \begin{pmatrix} ac-bd & ad+bc \\ bc+ad & ac-bd \end{pmatrix} \rightarrow \begin{matrix} \text{multiplication is some} \\ \text{linear transformation} \end{matrix}$$

$$a+ib \quad c+id = "(a+ib)(c+id)"$$

Think normalize

$$A = \frac{1}{\sqrt{a^2+b^2}} \begin{pmatrix} a & -b \\ b & a \end{pmatrix} \rightarrow \text{is a } \boxed{\text{Unitary matrix}}$$

\rightarrow preserves length, orthogonality, angles...

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Can write $A = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \rightarrow$ rotation by θ

So complex number multiplication \equiv rotate + scale by length.

C Euler notation $re^{i\theta}$

$$\text{Q} \quad re^{i\phi} pe^{i\theta} = rp e^{i(\theta+\phi)} \rightarrow \text{captures multiplication.}$$

Formulas

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

Why complex numbers?

Closed - complex fundamental theorem of algebra

C poly. w/ real coeffs. $P(x) = \sum a_i x_i^n$

Let $z_0 \in \mathbb{C}$ be a root $\sum a_i z_0^i = 0$

$$\text{So } (\sum a_i z_0^i)^* = 0 = \sum (a_i z_0^i)^* \quad (AB)^T = B^T A^T$$

$$= \sum (a_i)^* (z_0^i)^* = 0$$

$$= \sum (a_i)^* (z_0^*)^i = 0$$

$$= \sum (a_i) (z_0^*)^i = 0$$

S if z_0 is a root, then z_0^* is also a root.

$$\text{So } P(x) = (x-r_1)(x-r_2) \cdots (x-z_1)(x-z_1^*) \cdots (x-z_n)(x-z_n^*)$$

quadratic

C "real" fundamental theorem of algebra \rightarrow linear terms + quadratic terms
 any poly can be factored as ...

Feb 12, 2019

Review of Linear Spaces & Linear Functions

- Linear Space \equiv Vector Space $\rightarrow \mathbb{C}^n, \mathbb{P}_n$ (Poly. deg $\leq n$ complex coeffs)
- Coordinate system \equiv Basis $\rightarrow \mathbb{P} \rightarrow$ polynomials no restrictions on degrees closed under addition, multiplication scalar
- More examples. $C(X, \mathbb{C})$ continuous functions $[0, 1] \mapsto \mathbb{C}$
- Space of $n \times k$ matrices $M_{n \times k}(\mathbb{C})$

Linear Function $L: V \mapsto W$

$$\Leftrightarrow \begin{cases} L(\alpha \cdot v) = \alpha L(v) \\ L(v_1 + v_2) = L(v_1) + L(v_2) \end{cases} \quad \begin{matrix} \xrightarrow{\text{operations in } W} \\ \xrightarrow{\text{closed}} \end{matrix}$$

Ex. $\circ M_{2 \times 3}: \mathbb{C}^3 \mapsto \mathbb{C}^2$

$$A = [v_1 \ v_2 \ v_3]$$

$$A \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = \alpha v_1 + \beta v_2 + \gamma v_3$$

\Rightarrow "Matrix"

$$\circ \varphi: \mathbb{C}^3 \mapsto \mathbb{P}_2$$

\Rightarrow linear

We write

$$\varphi = [x^2 + ix \quad 3x - i \quad ix^2 + 2x - 7]$$

$$\hookrightarrow \varphi \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = \alpha(x^2 + ix) + \beta(3x - i) + \gamma(ix^2 + 2x - 7)$$

"Abit" $\varphi: \mathbb{C}^n \rightarrow V \rightarrow$ always linear.

Define $\varphi = [v_1 v_2 \dots v_n]$

$$\varphi(a) = \sum_{i=1}^n a_i v_i = a' v$$

Theorem

Every linear function $\varphi: \mathbb{C}^n \xrightarrow{\text{lin}} V$ is an atrix

key $\varphi(a) = a' \varphi(\vec{e}_i)$

So $\varphi = [\varphi(\vec{e}_i)]$ So entries are linear functions.

Why are atrix useful? $\rightarrow \left\{ \begin{array}{l} \text{linear independence} \\ \text{span} \\ \text{bases.} \end{array} \right.$

$\bullet \text{Im}(\varphi) = \text{span}\{\vec{v}_i\}$ IF $\varphi = [v_1 \dots v_m]$

Linear independence: $a' v_i = 0 \Leftrightarrow a_i = 0 \forall i$

$\bullet \varphi = [v_1 \dots v_m]$ injective when v_i 's linearly independent
 $\Leftrightarrow \ker(\varphi) = \{0\}$

\bullet "Basis" = "linearly independent" + "span"

$\bullet \varphi$ bijective $\Leftrightarrow \{\vec{v}_i\}$ is a basis of V

Elementary Column Operations

- { swap
- scale
- add one $\cdot \alpha$ to another.

$\Rightarrow \bullet [v_1 \dots v_n]: \mathbb{C}^n \mapsto V$

(6)

Note Let $\varphi = [v_1 \dots v_n] : \mathbb{C}^n \rightarrow V$

Consider $\varphi \xrightarrow[\text{lin}]{} M \xrightarrow[\text{lin}]{} \varphi \xrightarrow[\text{lin}]{} V \rightarrow W$

"Every matrix can be combined with a matrix"

$$\left\{ \begin{array}{l} \text{Ex } A = [v_1 \ v_2 \ v_3]_{V \times 3} \\ M = \left[\begin{array}{ccc} \vec{v}_1 & \vec{v}_2 & \vec{v}_3 \end{array} \right]_{3 \times 3} \end{array} \right\} A \circ M = \left[A\vec{v}_1 \ ? \ A\vec{v}_2 \right]_{V \times 2}$$

↳ Use the standard tuples

$$(A \circ M)(\vec{e}_i) = A(M(\vec{e}_i)) = A\vec{v}_i$$

$$\text{So } \boxed{A \circ M = \left[A(\vec{v}_1) \dots A(\vec{v}_n) \right]}_{[v_n]}$$

Ex $\boxed{A} = [v_1 \ v_2 \ v_3] \circ \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} = [v_3 \ v_1 \ v_2] = \textcircled{B}$

"Swap" is equiv to composing with bijection

→ "jectivity" is preserved.
 $\text{Im } A = \text{Im } B$

$[v_1 \ v_2 \ v_3] \circ \begin{pmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{pmatrix} = [v_1 \ 3v_2 \ v_3]$

$[v_1 \ v_2 \ v_3] \circ \begin{pmatrix} 1 & 0 & 0 \\ -4 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = [v_1 - 4v_2 \ v_2 \ v_3]$

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So, performing elementary column operations don't affect

$\left\{ \begin{array}{c} \text{injectivity} \\ \text{Image} \end{array} \right\}$ (composing w/ bijection)

$\blacksquare \quad \text{Im} [v_1 \dots v_n] = \text{Im} [v_1 \dots v_n \phi_v] \quad \Rightarrow \text{same thing}$

$\blacksquare \quad \text{Im} [v_1 \dots v_n] = \text{Im} [v_1 \dots v_{n-1}] \Leftrightarrow v_n \in \text{Ran} [v_1 \dots v_{n-1}]$

• (\Leftarrow) If $v_n \in \text{Ran} (v_1 \dots v_{n-1}) = \text{Im} (v_1 \dots v_{n-1})$

then $\text{Im} [v_1 \dots v_n]$

$$= \text{Im} [v_1 \dots v_{n-1} \sum_{i=1}^{n-1} v_i a_i] \quad \Rightarrow \text{E.C.O}$$

$$= \text{Im} [v_1 \dots v_{n-1} \phi_v]$$

$$= \text{Im} [v_1 \dots v_{n-1}]$$

• (\Rightarrow) $v_n \in \text{Ran} [v_1 \dots v_{n-1}] = \text{Ran} [v_1 \dots v_{n-1}]$

$$\text{Ran}^n \equiv {}^n \text{Im}^n$$

\blacksquare If not all $\{v_i\}$'s are null, by removing some of them, we can arrive at a linearly independent columns with the same span.

Consequently, If $\{v_i\}$ spans V then $\{v_i\}$ contains a

Note NOT THIS $\{\phi_v\} \rightarrow$ not a basis! Err!

Every basis of V has the same dimension

(8)

A

B

Suppose $\{v_1, \dots, v_{13}\}$ and $\{w_1, \dots, w_{17}\}$ span V

basis

$$\{v_1, \dots, v_{13}\} : \mathbb{F}^{13} \xrightarrow{\text{bij}} V \xleftarrow[\text{lin.}]{} \{w_1, \dots, w_{17}\} : \mathbb{F}^{17}$$

$$\hookrightarrow \{v_1, \dots, v_{13}\} : \mathbb{F}^{13} \xrightarrow[\text{lin.}]{} V \xrightarrow[\text{lin.}]{} \{w_1, \dots, w_{17}\} : \mathbb{F}^{17}$$

$$\hookrightarrow A : \mathbb{F}^{12} \xrightarrow{A} V \xrightarrow{B^{-1}} \mathbb{F}^{17}$$

Note $AB^{-1} : \mathbb{F}^{13} \xrightarrow{\text{inj.}} \mathbb{F}^{17}$ has to be a linear bijection

But $\leq AB^{-1}$ is a 17×13 matrix

But this can't happen \Rightarrow Contrad(+). true.

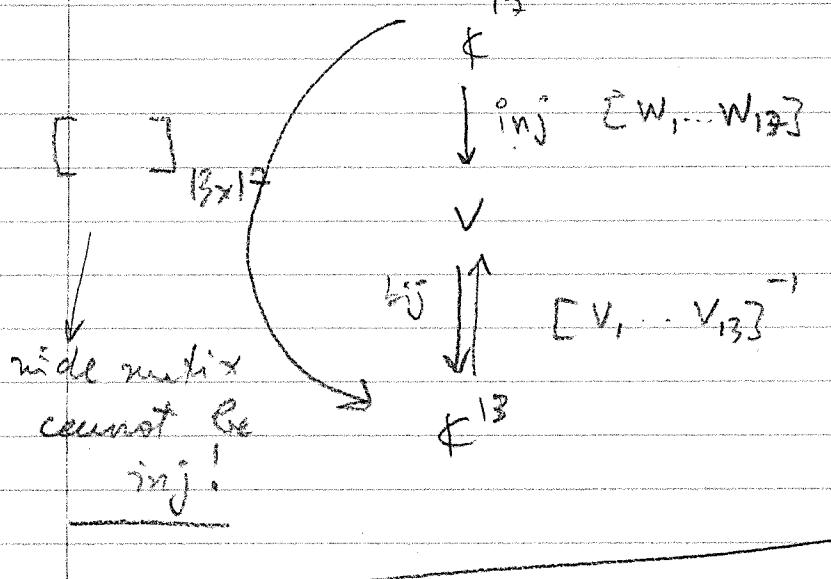
So

{ span sets have at least $\dim(V)$ elements
 { linearly independent sets have at most $\dim(V)$ elements.

Claim basis v_1, \dots, v_{13}
 lin. ind. w_1, \dots, w_{17}

b)

Every lin. ind. in finite dim can be enlarged to a basis



Start w. $\{v_1, \dots, v_n\}$ lin. ind.

If \rightarrow Span \rightarrow Stop

\rightarrow not Span $\rightarrow \exists v_{n+1}$

Set

$$v_{n+1} \notin \text{Im}[\{v_1, \dots, v_n\}]$$

\rightarrow Bigger matrix $A = [v_1, \dots, v_{n+1}]$

look at .tex

(9)

Claim $[v_1, \dots, v_n, v_{n+1}]$ lin. ind.

\hookrightarrow since $v_{n+1} \notin \text{Im}(w_1, \dots, w_n)$ $\Leftrightarrow \sum a_i v_i = 0 \Leftrightarrow a_i = 0$

$\rightarrow [v_1, \dots, v_n, v_{n+1}]$ lin. ind. (\Leftarrow)

\hookleftarrow keep repeating until $\{v_1 \neq \emptyset\}$
no v_i is a lin. ind. of the previous v_i 's

$$\# [J] = \dim(V)$$

Next time Isomorphisms, linear functions + matrices ...
 \rightarrow change of basis

then Rank-Nullity theorem

Then \rightarrow decomposition of vector space w/ a basis

Products of Vector Spaces

(recall $\mathbb{C}^2 \times \mathbb{C}^3$)

$$\mathbb{C}^5 \begin{pmatrix} x \\ y \\ z \\ w \\ v \end{pmatrix} \rightarrow \mathbb{C}^7$$

$\subseteq \mathbb{C}^2 \times \mathbb{C}^5 \cong \mathbb{C}^7$

Note $\mathbb{C}^5 \times \mathbb{C}^2 \neq \mathbb{C}^7$

... $\mathbb{C}^2 \times \mathbb{C}^5 \rightarrow \mathbb{C}^7$

$$\text{def. } \alpha \begin{pmatrix} 0 \\ \square \end{pmatrix} + \beta \begin{pmatrix} \hat{0} \\ \hat{\square} \end{pmatrix} = \begin{pmatrix} \alpha 0 + \beta \hat{0} \\ \alpha \square + \beta \hat{\square} \end{pmatrix}$$

So, what if we have 2 vector spaces V & W

$$V \times W = \left\{ \begin{pmatrix} v \\ w \end{pmatrix} \mid v \in V, w \in W \right\}$$

Define addition

$$\left\{ \begin{array}{l} \begin{pmatrix} v_1 \\ w_1 \end{pmatrix} + \begin{pmatrix} v_2 \\ w_2 \end{pmatrix} \triangleq \begin{pmatrix} v_1 + v_2 \\ w_1 + w_2 \end{pmatrix} \\ \alpha \begin{pmatrix} v \\ w \end{pmatrix} \triangleq \begin{pmatrix} \alpha v \\ \alpha w \end{pmatrix} \end{array} \right\} \text{This makes } V \times W \text{ a vector space.}$$

Neutral element $\rightarrow \begin{pmatrix} \emptyset_v \\ \emptyset_w \end{pmatrix}$

\square Suppose V is 3 dimensional
 W is 5 dimensional } finite-dim

claim $\dim(V \times W) = \dim(V) + \dim(W)$

Proof let $\{v_1, v_2, v_3\}$ be a basis of V

$\{w_1, w_2, w_3, w_4, w_5\}$ be a basis of W

Consider $\begin{pmatrix} v_1 \\ w_1 \end{pmatrix}, \begin{pmatrix} v_1 \\ w_2 \end{pmatrix}, \begin{pmatrix} v_1 \\ w_3 \end{pmatrix}, \begin{pmatrix} v_2 \\ w_1 \end{pmatrix}, \dots, \begin{pmatrix} v_2 \\ w_5 \end{pmatrix}, \begin{pmatrix} v_3 \\ w_1 \end{pmatrix}, \dots, \begin{pmatrix} v_3 \\ w_5 \end{pmatrix}$

claim these span $V \times W$

Given $\begin{pmatrix} v \\ w \end{pmatrix}$ $v = \sum_{i=1}^3 a_i v_i$, $w = \sum_{j=1}^5 b_j w_j$

$$\begin{pmatrix} v \\ w \end{pmatrix} = a_1 \begin{pmatrix} v_1 \\ w_1 \end{pmatrix} + a_2 \begin{pmatrix} v_1 \\ w_2 \end{pmatrix} + \dots + a_3 \begin{pmatrix} v_1 \\ w_3 \end{pmatrix} + b_1 \begin{pmatrix} v_2 \\ w_1 \end{pmatrix} + \dots + b_5 \begin{pmatrix} v_2 \\ w_5 \end{pmatrix} + \dots + c_3 \begin{pmatrix} v_3 \\ w_1 \end{pmatrix} + \dots + c_5 \begin{pmatrix} v_3 \\ w_5 \end{pmatrix} \Rightarrow \text{span}$$

(11)

$$\text{Lin-indep} \quad a_1 \begin{pmatrix} v \\ w \end{pmatrix} + \dots + a_8 \begin{pmatrix} \phi_v \\ w \end{pmatrix}$$

$$= \begin{pmatrix} \sum a_i v_i \\ \sum a_i w_i \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Leftrightarrow \begin{cases} a_1 = \dots = a_8 = 0 \\ \sum a_i v_i = 0 \\ \sum a_i w_i = 0 \end{cases}$$

So all coeffs are zero

$$\Rightarrow \text{so } \dim(V+W) = \dim(V) + \dim(W) = 8$$

Cartier Vector products are not commutative



$$W \times V \neq V \times W$$

$$W \times V \sim V \times W$$

But they're not the same

Not associative

$$V \times (W \times Z) \neq (V \times W) \times Z$$

$$(\text{natural}) \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \neq \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} (\text{natural})$$

Def

$$V \times \dots \times V_n = \left\{ \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \mid v_i \in V_i \right\}$$

Note

$$\begin{pmatrix} v \\ w \end{pmatrix} \rightsquigarrow \begin{pmatrix} v \\ b_w \end{pmatrix} + \begin{pmatrix} \phi_v \\ w \end{pmatrix} \quad \text{"ubspac"} \nearrow$$

Note

$$\left\{ \begin{pmatrix} v \\ b_w \end{pmatrix} \mid v \in V \right\} \leftarrow V \times W$$

$$\left\{ \begin{pmatrix} \phi_v \\ w \end{pmatrix} \mid w \in W \right\} \leftarrow V \times W$$

in a

Every object in $V \times W$ can be written unique way with $\begin{pmatrix} v \\ b_w \end{pmatrix}, \begin{pmatrix} \phi_v \\ w \end{pmatrix}$

→ this is something along the line of fair...

Q: If Z, U are subspaces of a vector space W
then $Z + U \triangleq \{z + u \mid z \in Z, u \in U\}$

\rightarrow idea → decomposing a vector space

Question is $Z + U$ a subspace? → not empty $\phi_2 + \phi_3$

$$\text{Yes. } \alpha(z+u) = \alpha z + \alpha u \in W$$

$\begin{matrix} \uparrow & \uparrow \\ \in Z & \in U \end{matrix} \quad \begin{matrix} \uparrow & \uparrow \\ \in Z & \in U \end{matrix} \quad \begin{matrix} \in Z & \in U \\ \in Z & \in U \end{matrix}$

$$(z_1 + u_1) + (z_2 + u_2) = (z_1 + z_2) + (u_1 + u_2)$$

So $Z + U$ subspace

Question is $Z + U$ commutative? Yes
 $U + Z$

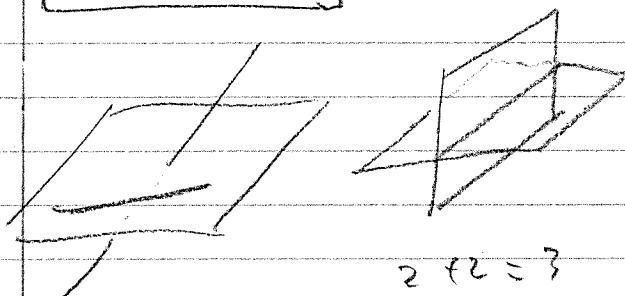
→ $Z + U = U + Z$

Question Associativity? Yes. $(Z + U) + W = Z + (U + W)$

Dimensions?

→ They don't add!

{ span
linear. and }



$$2 + 2 = 3$$

$$2 + 1 = 2$$

Suppose $\mathbb{Z}_1, \mathbb{Z}_2$ are subspaces of W

Consider $\mathbb{Z}_1 \times \mathbb{Z}_2 \xrightarrow{\oplus} \mathbb{Z}_1 + \mathbb{Z}_2 \subset W$

Defined by $\varphi\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right) \stackrel{\oplus}{=} z_1 + z_2$

(1) $\rightarrow \varphi$ is a linear function. $\rightarrow \varphi$ is linear

$$\varphi\left(\alpha\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right) + \beta\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right)\right) = \varphi\left(\begin{pmatrix} \alpha z_1 + \beta z_1 \\ \alpha z_2 + \beta z_2 \end{pmatrix}\right) = \alpha z_1 + \beta z_1 -$$

$$= \alpha \varphi\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right) + \beta \varphi\left(\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}\right) = \quad \checkmark$$

(2) If W finite dim, then $\mathbb{Z}_1, \mathbb{Z}_2$ finite dim

so

$\mathbb{Z}_1 \times \mathbb{Z}_2$ and $\mathbb{Z}_1 + \mathbb{Z}_2 \subset W$ finite dim

By Rank-Nullity Theorem (fin \rightarrow fin, linear)

$$\dim(\mathbb{Z}_1 \times \mathbb{Z}_2) = \underbrace{\dim(\mathbb{Z}_1 + \mathbb{Z}_2)}_{\dim(\ker(\varphi))} + \dim(\text{Im } \varphi)$$

$$\dim(\mathbb{Z}_1) + \dim(\mathbb{Z}_2) = \dim(\mathbb{Z}_1 + \mathbb{Z}_2) + \dim(\ker \varphi)$$

What is $\ker(\varphi)$?

?

$$\ker(\varphi) = \left\{ \begin{pmatrix} z \\ -z \end{pmatrix} \mid z \in \mathbb{Z}_1, z \in \mathbb{Z}_2 \right\} = \left\{ \begin{pmatrix} z \\ -z \end{pmatrix} \mid z \in \mathbb{Z}_1 \cap \mathbb{Z}_2 \right\}$$

From Assignment 1, $\boxed{\dim \left\{ \begin{pmatrix} u \\ -u \end{pmatrix} \mid u \in U \right\} = \dim(U)}$ also

start with \rightarrow

a subspace
of W

$$\text{So } \dim(\ker(\varphi)) = \dim(Z_1 \cap Z_2)$$

(A sieve formula)

$$\boxed{\dim(Z_1 + Z_2) = \dim(Z_1) + \dim(Z_2) - \dim(Z_1 \cap Z_2)}$$

when $Z_1 \cap Z_2$ is trivial, then $Z_1 + Z_2$ is direct.

$$\rightarrow \dim(\ker \varphi) = 0 \rightarrow \varphi \text{ is injective}$$

But φ is also surjective by def

$$\Rightarrow \varphi \text{ is a bijection.} \Rightarrow Z_1 \otimes Z_2 \xrightarrow{\sim} Z_1 \oplus Z_2$$

isomorphic

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Warm up $W_1 + W_2 = V$ direct when $W_1 \cap W_2 = \{\emptyset\}$

φ bijective $\Leftrightarrow a = \emptyset$

$b = \emptyset$

Contrapositive

If $\{\emptyset\} \neq W_1 \cap W_2$, then $\exists u \neq \emptyset$
st $u \in W_1 \cap W_2$

then $u + (-u) = \emptyset$ (contradiction)

Why direct sum?

\rightarrow generalize the idea of Basis.

(s decomposes a big vector space into understandable ones.)

Now let V be a f.dim vector space

$$L \in \mathcal{L}(V) \quad (L: V \xrightarrow{\text{fin}} V)$$

$$\begin{array}{c} \text{Im}(L^2) \subseteq \text{Im}(L) \subseteq V \\ \parallel \quad \parallel \quad \parallel \\ L^2[V] \subseteq L[V] \subseteq I[V] \\ \parallel \\ L^0[V] \end{array}$$

What about $L^2[V]$? $\rightarrow L[V] \supset L[L[V]]$

automatic

Claim if $\text{Im}(L^p) = \text{Im}(L^{p+1})$, then

$$\text{Im}(L^{p+1}) = \text{Im}(L^{p+2})$$

(i.e. if $\text{Im}(L^p) = \text{Im}(L^{p+1})$ then

$$(\text{Im } L^{p+2} \subset \text{Im } L^{p+1}) \text{ then } \text{Im}(L^{p+1}) \subset \text{Im}(L^{p+2})$$

To Show $L^{p+1}(x) \in \text{Im } L^{p+2} \forall x \in V$, suppose $\text{Im}(L^p) = \text{Im}(L^{p+1})$

$$\begin{aligned} L^{p+1}(x) &= L(L^p(x)) = L(L^{p+1}(v)) \text{ for some } v \in V \\ &= L^{p+2}(v) \in \text{Im}(L^{p+2}) \end{aligned}$$

Step 5

claim true

"If you set equality, then you'll set equality forever..."

Theorem If $\dim(V) = n$, and $L \in \mathcal{L}(V)$, then

$$\text{Im } L^n = \text{Im } L^{n+1} = \text{Im } L^{n+2} = \text{Im } L^{n+3} = \dots$$

$$\text{and } \ker L^n = \ker L^{n+1} = \ker L^{n+2} = \dots$$

Reason
dimension

Kernel grows... $\{\phi_v\} \subseteq \ker(L) \subseteq \ker(L^2) \subseteq \dots = \dots =$

Image shrinks... $= \dots = \dots \subseteq \text{Im}(L^2) \subseteq \text{Im}(L) \subseteq V$

What about

$\text{Im}(L^n)$ and $\text{ker}(L^n)$ and V ?

for any $n \in \mathbb{N} = \dim(V)$

Claim

Theorem

If V is finite dimensional and $L \in \mathcal{L}(V)$

$$\dim(V) = n$$

$$\text{Then } V = \text{Im}(L^n) \oplus \text{ker}(L^n)$$

Proof
use
rank-nullity
formula

By rank-nullity

$$\dim(V) = \dim(\text{Im}(L^n)) + \dim(\text{ker}(L^n))$$

If we can show that $\text{Ran } L^n \cap \text{ker } L^n = \{0_V\}$

then the sum is direct, and

$$\underbrace{\dim(\text{Ran } L^n \oplus \text{ker } L^n)}_{\text{dimension of } V} = \dim(\text{Ran } L^n) + \dim(\text{ker } L^n)$$

$$= n = \dim(V)$$

$$\text{So } \text{Ran}(L^n) \oplus \text{ker}(L^n) = V$$

Now if $x \in \text{Im } L^n \cap \text{ker } L^n$, then

$$x \in L^n(x), \quad x \in V \Rightarrow L^n(x) = 0_V$$

$$\text{So } L^n(L^n(x)) = 0_V = L^{2n}(x)$$

$$\text{So } v \in \text{ker}(L^{2n}) \Rightarrow v \in \text{ker}(L^n) \Rightarrow x = 0_V$$

(17)

Note only true if $n = \dim(V)$

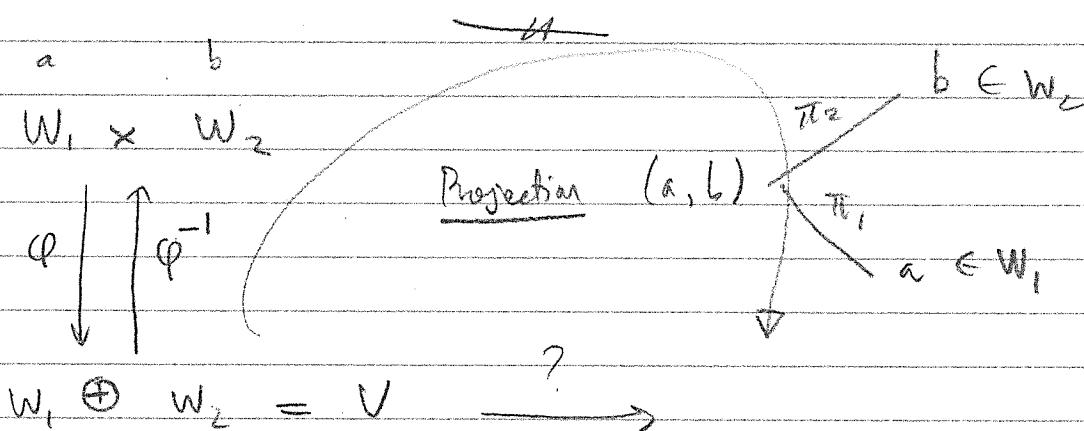
If $\dim V = 3$, then $\ker(L^2) \oplus \text{Ran}(L^2) \rightarrow$ may not give V .

■ "The loss in dimension" $V \rightarrow \text{Im}(L) \rightarrow \text{Im}(L^2) \rightarrow \dots$

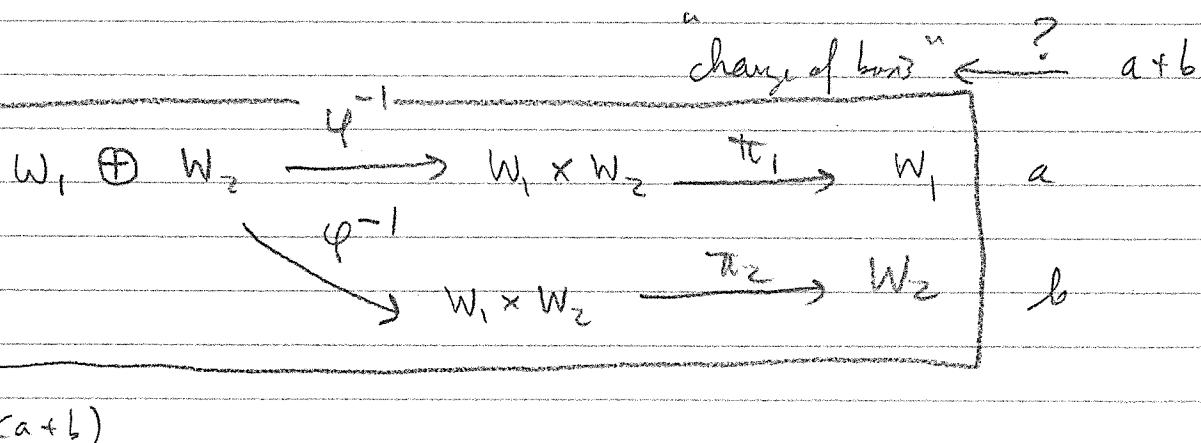
↳ Weyr number → can change each step.

What about gain in dimension kernel?

↳ By nullity, Weyr for loss of dim of range (image)
is the same of gain in dimension of kernel ...



What are projections on V ? $W_1 \oplus W_2 \rightarrow V$



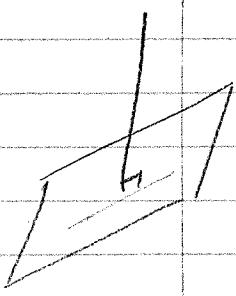
Q Let $E: W_1 \oplus W_2 \rightarrow W_1$, " $E(a+b) = a$ "

defined as $E(v) = a \in W_1$, s.t. $\exists b \in W_2$ s.t. $a+b=v$

So $\underbrace{E(E(v))}_a = E(v) = a$ $a = a+0$ unique since \oplus
 $\cap \cap \cap$
 $W_1 \quad W_1 \quad W_2$
 true for any $v \in V$

- $E^2 = E$ \rightarrow "idempotent" (squares to themselves)

\hookrightarrow Note $\left\{ \begin{array}{l} \text{Ran}(E) = \text{Im}(E) = W_1 \\ \ker(E) = W_2 \end{array} \right\}$



S

Decomposition of $W_1 \oplus W_2$ gives rise to idempotent
 whose $\text{Im} = W_1$
 $\ker = W_2$

\rightarrow Are there other idempotents like this? No

\hookrightarrow For idempotent (F) $\rightarrow \ker(F) \oplus \text{Im}(F) = V$

In fact Idempotent \leftrightarrow direct sum

\hookrightarrow coding space by functions $E_1 + \dots + E_k = I$

\hookrightarrow "resolution of identity"

Ex $(\begin{smallmatrix} 1 & 0 \\ 1 & 1 \end{smallmatrix}) = (\begin{smallmatrix} 1 & 0 \\ 0 & 0 \end{smallmatrix}) + (\begin{smallmatrix} 0 & 0 \\ 1 & 0 \end{smallmatrix}) + (\begin{smallmatrix} 0 & 0 \\ 0 & 1 \end{smallmatrix})$

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

$$\text{span}(\vec{e}_1) \quad \text{span}(\vec{e}_2) \quad \text{span}(\vec{e}_3)$$

Why functions code for space? \rightarrow So we can do algebra on space!

Idempotents \rightarrow linear operator.

Feb 21
2019

Operator: linear function from vector space onto itself.

$$E: V \mapsto V$$

Def

Idempotents are operators with the property $E^2 = E$, i.e. $E \circ E = E$.

Last time

If $V = W \oplus Z$, then there exists an idempotent $E \in L(V)$ such that

$$\begin{aligned} W &= \text{Im}(E) \\ Z &= \text{ker}(E) \end{aligned}$$

"projection"

In fact, $V = W \oplus Z$ then there are at least 2 idempotents E , $W = \text{Im}(E)$, $Z = \text{ker}(E)$ and F , $W = \text{ker}(F)$, $Z = \text{Im}(F)$

Now (?) \rightarrow Is every idempotent $E \in L(V)$ generated this way?

(I)

The answer is Yes \rightarrow Two parts to answer { existence } uniqueness }

II/

Theorem:

(II)

\hookrightarrow If $E \circ E = E^2 = E$, then $V = \text{Im}(E) \oplus \text{ker}(E)$
 $V = \text{Im}(E) \oplus \text{ker}(E)$

Proof 1) Show $\text{Im}(E) + \text{ker}(E) = V$

2) Show $\text{Im}(E) \oplus \text{ker}(E) = V$

↗
↙

(1) First, $v = E(v) + (v - E(v))$ - Now $E(v - E(v)) = E(v - E^2 v) = E(v - E^2 v) = E(v - E^2 v) = 0$

$$\begin{array}{ccc} A & & A \\ \text{Im}(E) & & \text{ker}(E) \end{array} \leftarrow$$

Now (2) directness. Show $\text{Im}(E) \oplus \ker(E) = V$

$$\text{i.e. } \text{Im}(E) \cap \ker(E) = \{0_V\}$$

Proof let $x \in \text{Im}(E) \rightarrow E(\underbrace{\text{Im}(E)}_{\text{generic element in range image}}) = \{0_V\}$

$$\therefore E(x) = 0_V$$

$$\therefore E(x) \in \ker(E)$$

By Product

$$\hookrightarrow \boxed{\begin{array}{l} E(E(x)) = E(x) \text{ for all } x \in V \\ \Leftrightarrow E = E^2 \end{array}}$$

i.e. E is an idempotent matrix when it acts as an identity for its own image.

Theorem

$$\text{II/} \quad \left. \begin{array}{l} \text{if } E^2 = E \text{ and } G^2 = G \text{ and} \\ \text{in } \mathcal{L}(V) \end{array} \right\} \begin{array}{l} \text{Im}(E) = \text{Im}(G) \\ \ker(E) = \ker(G) \end{array}$$

$$\text{then } E = G$$

$$\text{not if } W = \text{Im}(E) = \text{Im}(G) \text{ Note } W \oplus Z = V$$

$$Z = \ker(E) = \ker(G)$$

$$\text{Then } E(w) = E(w+z) = E(w) + \underbrace{E(z)}_{0_V} = \underbrace{E(w)}_{=w}$$

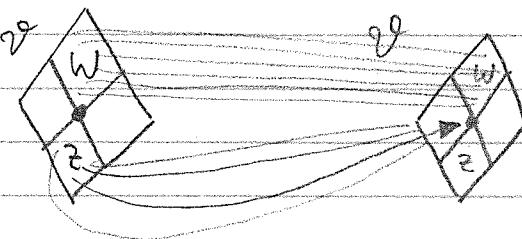
where $w = w+z$ is unique

Can do the same thing with $G \rightarrow G(u) = w = E(u) + z$

$$\rightarrow G = E$$

Picture of idempotent

→ think "projections"



$$W = \text{Im}(E)$$

$$Z = \ker(E)$$

So

For each decomposition $V = W \oplus Z$, there is a unique idempotent whose Im is W and kernel is Z

Now, notice:

$$W + Z \stackrel{?}{=} V = W + Z \xrightarrow{\quad E \quad} W \quad \xrightarrow{\quad F \quad} Z$$

$$\text{So } E + F = I$$

Now, E, F are idempotents s.t. $\text{Ran}(E) = \text{Im}(F)$

$\ker(E) = \text{Im}(F)$, then

$$E + F = I \sim \text{identity}$$

So

(1) If $E^2 = E \in \mathcal{L}(V)$, then $(I - E)^2 = (I - E)$

(2) $\text{Im}(E) = \ker(I - E)$

$\ker(E) = \text{Im}(I - E)$

So idempotents come in pairs

Summary If $V = W \oplus Z$, then there is a pair of idempotents ... more concretely ...

\Rightarrow If $V = W \oplus Z$, then there exist idempotents $E_2 F$ such that $E + F = I$ and

$$\begin{cases} W = \text{Im}(E) = \text{Ker}(F) \\ Z = \text{Ker}(E) = \text{Im}(F) \end{cases}$$

\Leftarrow If \exists exist idempotents $E_2 F$ -- such th. -- then $V = W \oplus Z$

\blacksquare All great, but what about $W \oplus Z \oplus V = V$?

well $W \oplus (Z \oplus V) = V \Rightarrow$

By \nearrow \exists unique $E_W^2 = E_W$ such that $\text{Im}(E_W) = W$
 $\text{Ker}(E_W) = Z \oplus V$

(*) Similarly $E_V^2 = E_V$ s.t. $\text{Im}(E_V) = V$
 $\text{Ker}(E_V) = Z \oplus W$
 $E_Z^2 = E_Z$ s.t. $\text{Im}(E_Z) = Z$
 $\text{Ker}(E_Z) = V \oplus W$

So there exists \nearrow

Observation ① $E_W + E_V + E_Z = ?$

$$(E_Z + E_V + E_W)(x) = (E_W + E_V + E_Z)(u + w + z)$$

$$= E_W(u + u + z) + E_V(w + u + z) + E_Z(u + w + z)$$

$$= w + u + z$$

$\underline{\text{So}}$ $E_W + E_V + E_Z = I$

Obs (2) $E_w(E_z(v)) = E_w(z) = 0,$

So $E_w \circ E_z$ is null.

In fact compositions of different E_i 's get Φ_v .

compositions of the same E_i ? Let E_i :

Now other direction

↳ If we know $E_w^2 = E_w$, $E_u^2 = E_u$, $E_z^2 = E_z$ such that
... (page 22), then

$$V = W \oplus U \oplus Z$$

↳ Result (Suppose E_w, E_u, E_z satisfy $(*)$ for some
 $W, U, Z \subset V$)

Suppose E_1, E_2, E_3 are idempotents such that

$$E_1 + E_2 + E_3 = I \text{ and } E_i E_j = \delta_{ij} E_i, \text{ then}$$

$$\text{Im}(E_1) \oplus \text{Im}(E_2) \oplus \text{Im}(E_3) = V \text{ and}$$

$$\text{Im}(E_1) = \text{Im}(E_2) \oplus \text{Im}(E_3), \text{ etc}$$

Once we established $(*)$ and $(**)$, we have the following assertions are equivalent for $w, z, u \in V$

① $W \oplus Z \oplus U = V$ \Rightarrow unique

② There are 3 idempotents $E_1, E_2, E_3 \in \mathcal{L}(V)$
such that

$$\text{Im}(E_1) = W, \text{Im}(E_2) = Z, \text{Im}(E_3) = U$$

$$\text{and } E_1 + E_2 + E_3 = I, E_i E_j = \delta_{ij} E_i$$

$\textcircled{1} \rightarrow \textcircled{2}$ is already shown.

$\textcircled{2} \rightarrow \textcircled{1}$ We will show now.

(1) We have $\text{Im}(E_1) + \text{Im}(E_2) + \text{Im}(E_3) = V$

$$\hookrightarrow v = id(v) = (E_1 + E_2 + E_3)(v) = \underset{\textcircled{1}}{E_1(v)} + \underset{\textcircled{2}}{E_2(v)} + \underset{\textcircled{3}}{E_3(v)}$$

$$\text{Im}(I) = V \quad \underset{\textcircled{1}}{\text{Im}(E_1)} + \underset{\textcircled{2}}{\text{Im}(E_2)} + \underset{\textcircled{3}}{\text{Im}(E_3)}$$

$$\underline{\text{So } V = \text{Ran}(E_1) + \text{Ran}(E_2) + \text{Ran}(E_3)}$$

(2) Show directness. Suppose $x_1 + x_2 + x_3 \in D_V$, $x_i \in \text{Im}(E_i)$

To show: $x_i = 0_V \forall i$

$$\text{Then } E_1(x_1) + E_2(x_2) + E_3(x_3) = 0_V$$

$$\text{So } E_1(E_1(x_1) + E_2(x_2) + E_3(x_3)) = E_1(0_V) = 0_V$$

$$\hookrightarrow E_1(E_1(x_1)) = 0_V$$

$$\underline{\text{So } E(x_1) = 0_V}$$

$$x_1 = 0_V$$

Same thing, set $x_2 = x_3 = x_1 = 0_V$

So got directness. $\text{Im}(E_1) \oplus \text{Im}(E_2) \oplus \text{Im}(E_3) = V$

\curvearrowright (there's more ...)

Note. for finite vector space $\sum_{i=1}^n E_i = I \Rightarrow E_i E_j = \delta_{ij}^1 E_i$

• Resolution of identity, code direct sum as algebra ...

Next, kernel

$$\text{Note } E_1(E_2(v)) = \emptyset,$$

$$\text{So } \text{Im}(E_2) \subset \ker(E_1)$$

$$\text{Similarly, } \text{Im}(E_3) \subset \ker(E_2)$$

$$\text{So } \text{Im}(E_3) + \text{Im}(E_2) \subset \ker E_1$$

But this is also a direct sum, so

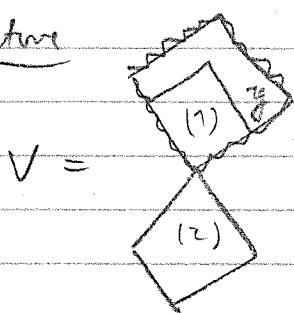
$$\text{Im}(E_3) \oplus \text{Im}(E_2) \subset \ker E_1 \quad \begin{matrix} \nearrow \text{cannot be} \\ \text{proper} \rightarrow \text{must} \\ \text{be equality} \end{matrix}$$

$$\text{Now } \underline{\text{Im}(E_1)} = \text{Im}(E_1)$$

$$V = V$$

$$\text{So } \text{Im}(E_2) \oplus \text{Im}(E_3) = \ker(E_1) \quad \square$$

Picture

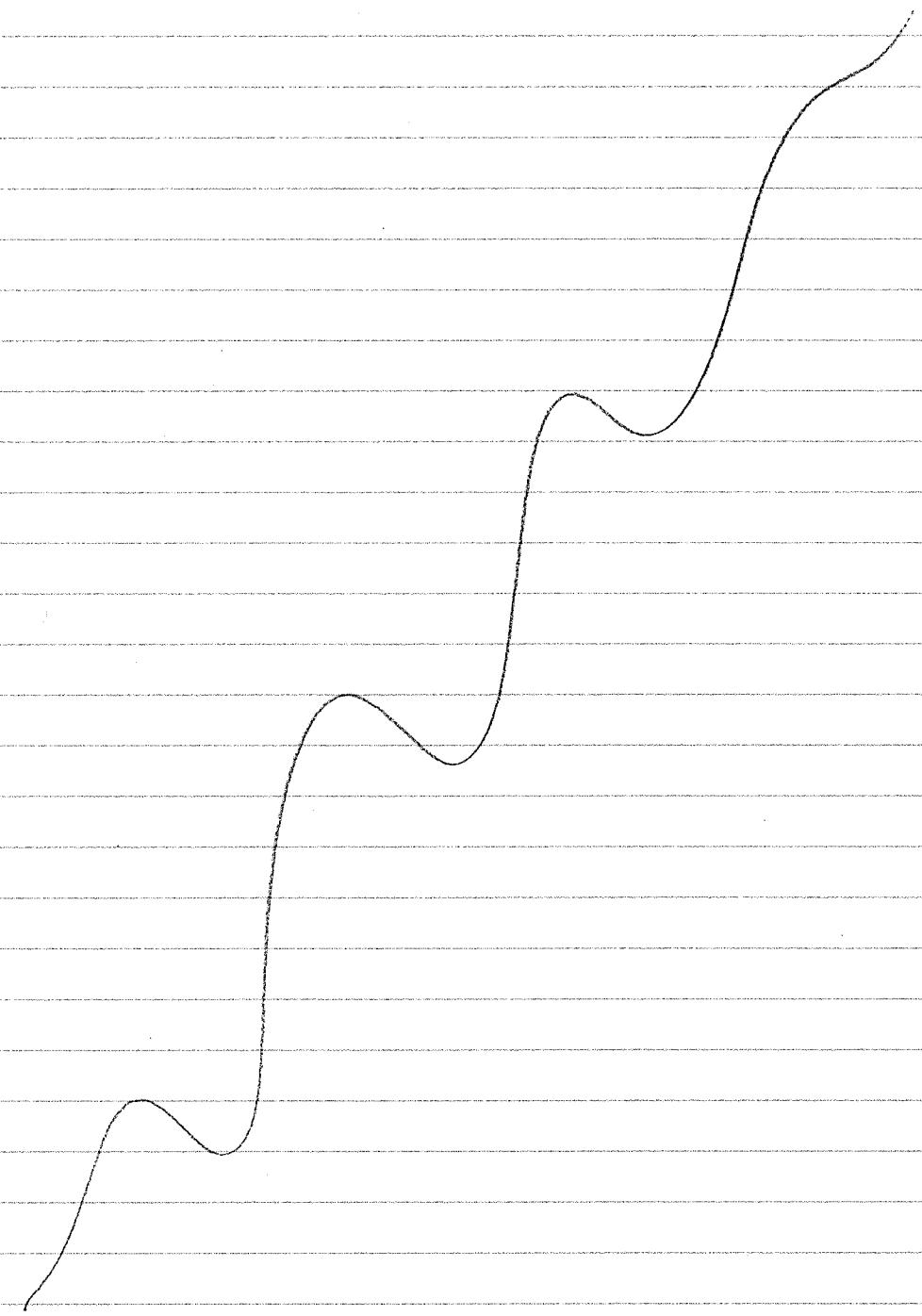


$$y = \text{shape} + \diamond = y + \emptyset_v$$

$$y = \diamond + \diamond = \text{shape} + \diamond = y + \emptyset_v$$

This is

So y is \diamond ... but
this isn't work



Review of eigenvectors.

Feb 26, 2019

TFAE for $A \in L(V)$

- ① $A^2 = A$
- ② $Ax = x$ for $x \in \text{Im}(A)$
- ③ $(I-A)^2 = (I-A)$
- ④ $\text{Im}(A) = \ker(I-A)$
- ⑤ $\text{Im}(I-A) = \ker(A)$

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Today

Representing linear functions as matrices

Consider $V \xrightarrow{\text{f.l.}} V$ and coordinate system, i.e. ordered basis.

$\beta = (v_1, \dots, v_m)$ ordered basis of V (coord. sys. of V)

Then if $z \in V$, $z = \sum_{i=1}^m a_i v_i$ unique. But by notation
 \rightarrow new notation

$$[z]_\beta = (a_1 \dots a_m)^T$$

So consider matrix $[v_1 \dots v_m]$, then $[v_1 \dots v_m][z]_\beta = z$

i.e. $[v_1 \dots v_m] \begin{pmatrix} a_1 \\ \vdots \\ a_m \end{pmatrix} = a_1 v_1 + a_2 v_2 + \dots + a_m v_m = z$

$$\boxed{A_\beta [z]_\beta = z}$$

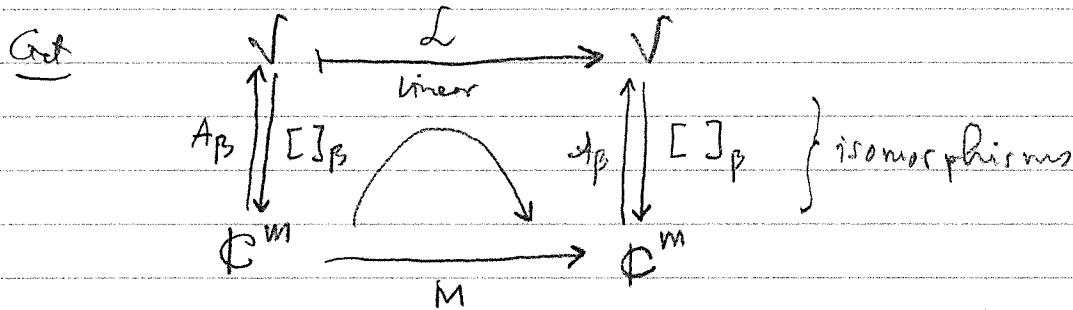
\Rightarrow gives the coords of z
gives z from $[z]_\beta$

$\xrightarrow{\text{Image}}$

$$\begin{matrix} & \checkmark & \xrightarrow{\text{f.l.}} & V \\ [z]_\beta & \downarrow A_\beta & \uparrow A_\beta^{-1} = [z]_\beta & \\ & \checkmark & \mathbb{C}^m & \end{matrix}$$

Note $[v_1 \dots v_m] = A_\beta$ is a bijection, obviously

More fully, consider $L: V \xrightarrow{\text{lin}} V$



\square M is a composition of linear fns mapping $C^m \mapsto C^m$, so M is a matrix. Note $A_B^{-1} = [J_B]$ are isomorphisms

\square We note $M \equiv [L]_{\beta \leftarrow \beta}$

$$\square \text{ Def } [L]_{\beta \leftarrow \beta} := [J_\beta \circ L \circ A_\beta] = A_\beta^{-1} \circ L \circ A_\beta$$

Now $z \xrightarrow{} L(z)$

$$\begin{array}{ccc} & \downarrow & \downarrow \\ [z]_\beta & \xrightarrow{} & [L(z)]_\beta \end{array}$$

$$\square \text{ Claim } \underset{\parallel}{[L(z)]_\beta} = [L]_{\beta \leftarrow \beta} \underset{\parallel}{[z]_\beta}$$

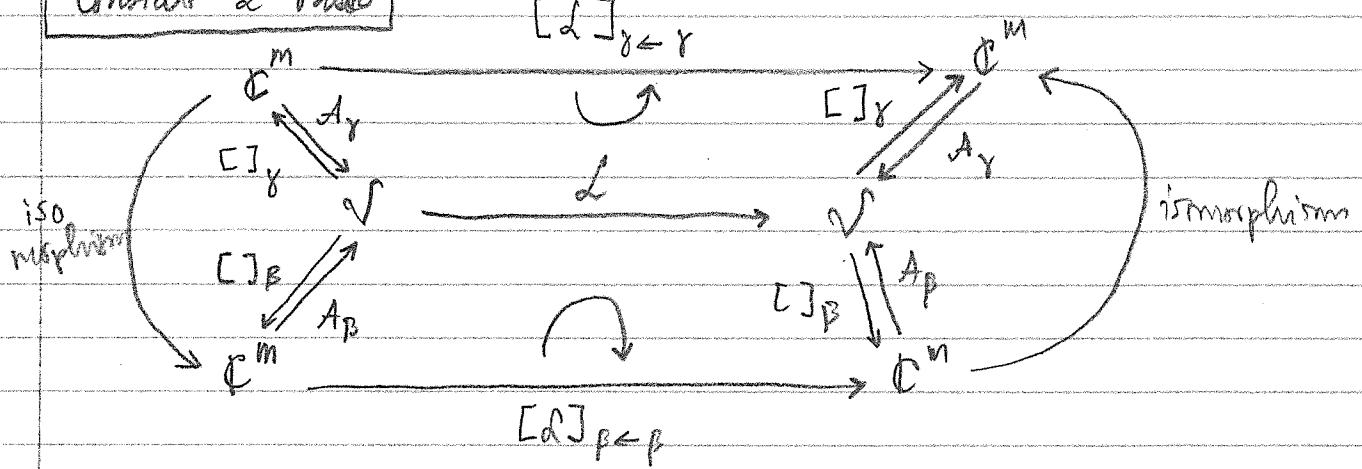
$$\hookrightarrow [J_\beta \circ L(z)] \text{ or } ([J_\beta \circ L \circ A_\beta] \circ \underset{\parallel}{[z]_\beta})$$

$$\therefore ([J_\beta \circ L(z)] = ([J_\beta \circ L](z)) \stackrel{I}{\longrightarrow} \text{true.}$$

So, observe $(z) \xrightarrow{L} L(z)$

$$\begin{array}{ccc} & \downarrow & \\ [z]_\beta & \xrightarrow{M} & [L(z)]_\beta \end{array}$$

Consider 2 basis



Are $\mathcal{L}_{\gamma \leftarrow \gamma}$ the same as $\mathcal{L}_{\beta \leftarrow \beta}$

$$\begin{aligned}\mathcal{L}_{\gamma \leftarrow \gamma} &= [J_\gamma \circ \mathcal{L} \circ A_\gamma] \\ &= [J_\gamma \circ (A_\beta \circ \mathcal{L}_{\beta \leftarrow \beta} \circ J_\beta) \circ A_\gamma] \\ &= ([J_\gamma \circ A_\beta] \circ [\mathcal{L}_{\beta \leftarrow \beta}] \circ ([J_\beta \circ A_\gamma])\end{aligned}$$

$$\begin{aligned}\text{Note } [J_\gamma \circ A_\beta] &= ([J_\beta \circ A_\gamma])^{-1} \\ &= A_\gamma^{-1} \circ [J_\beta^{-1}] \\ &= [J_\gamma \circ A_\beta]\end{aligned}\quad \boxed{\text{nice}}$$

But note $[J_\gamma \circ A_\beta]$ and $[J_\beta \circ A_\gamma]$ are invertible matrices $\mathbb{C}^m \xrightarrow{\sim} \mathbb{C}^m$.

if we chose different bases.

invertible matrix

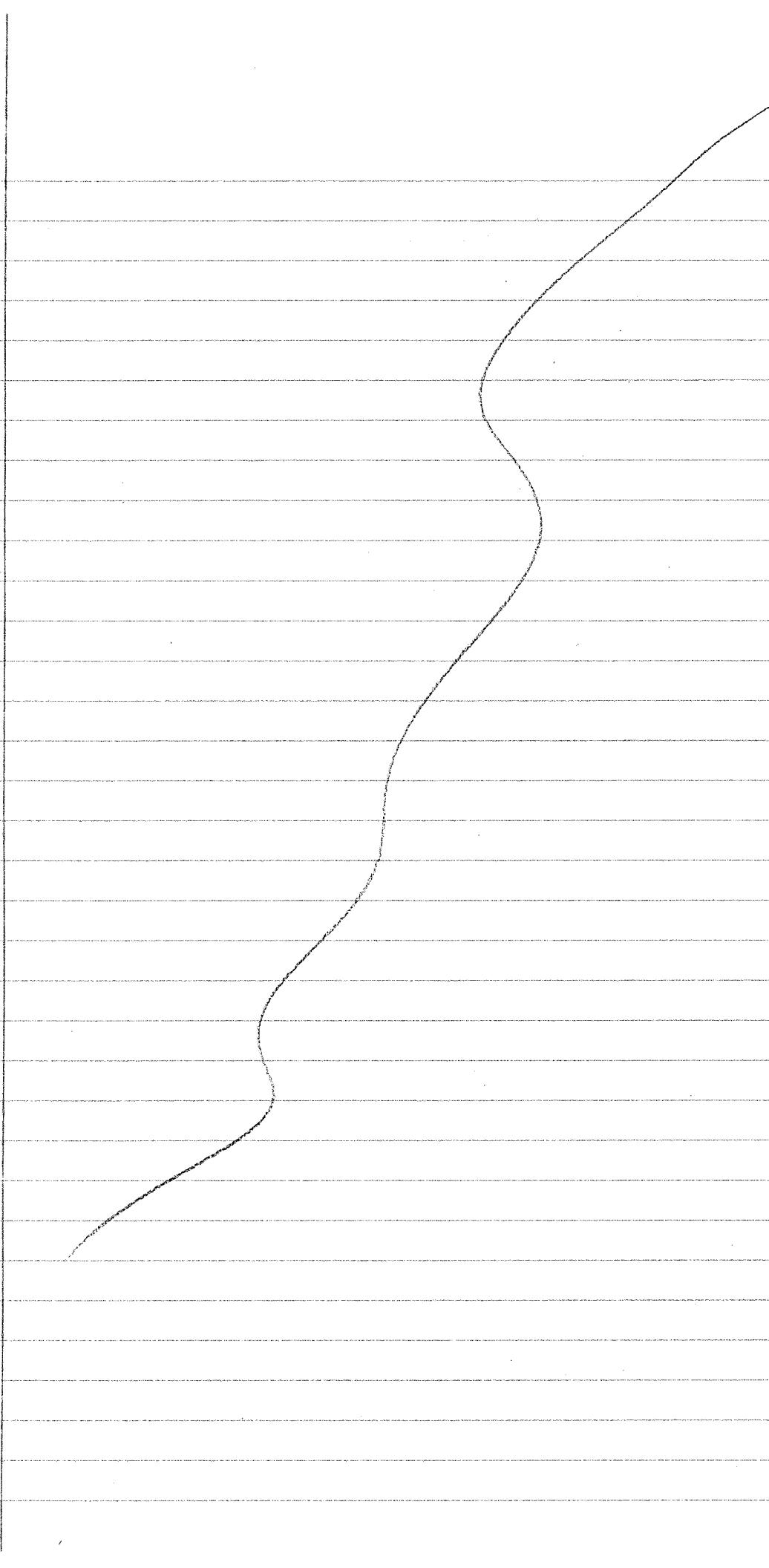
$$\boxed{[\mathcal{L}]_{\gamma \leftarrow \gamma} = M^{-1} \circ [\mathcal{L}]_{\beta \leftarrow \beta} \circ M}$$

Is $[\mathcal{L}]_{\gamma \leftarrow \gamma}$ similar to $[\mathcal{L}]_{\beta \leftarrow \beta}$?

?

Question If $B \sim [\mathcal{L}]_{\beta \leftarrow \beta}$, does there exist basis Y such that $B = [\mathcal{L}]_{\gamma \leftarrow \gamma}$

~~28~~



The answer is YES

If $B = k^{-1} [L]_{B \leftarrow k}$, we will show that there is a basis γ of V such that

$k = []_{B \circ A_\gamma}$, so that $B = [L]_{\gamma \circ \gamma}$.

i.e. we will show that there is a basis γ of V st

$$A_B \circ k = A_\gamma$$

Since $A_B \circ k$. So $A_B \circ k$ must be a bijective

$$\begin{array}{ccc} V & \xleftarrow{\text{isom}} & C^m \\ & & \xleftarrow{\text{isom}} \\ & & C^m \xleftarrow{\text{isom}} C^m \end{array}$$

$$A_\gamma = [\quad \quad \quad]$$

some basis δ of V

So similar matrices represent the same linear function. But which one? What is this linear function?

Since the answer is unsatisfying... $I = A_\varepsilon \xrightarrow{\text{isom}} [I]_\varepsilon = [I]_\varepsilon \xrightarrow{\text{isom}} A_\varepsilon = I \varepsilon \rightarrow \text{std basis}$

$$\text{Let } A \in M_{153}$$

$$\hookrightarrow \text{itself } A = [A]_{\varepsilon \leftarrow \varepsilon}$$

$$\begin{array}{ccc} \mathbb{C}^{153} & \xrightarrow{\text{isom}} & \mathbb{C}^{153} \\ A & \longmapsto & A \end{array}$$

So if $A \sim B$ then B represents A with respect to some different basis (not the standard basis)

Big Idea

Similarly is all about change of basis.

Similar matrices give represent the same linear fn.

The goal is to find better representation...

Ex

$$P_3 \xrightarrow{L} P_3$$

defined as $L(p) = p' + 2p$

Find $[L]_{\beta \leftarrow \beta}$ where $\beta = \{1, x+1, x^2, x^3\}$

$$A_\beta = \begin{bmatrix} 1 & x+1 & x^2 & x^3 \end{bmatrix}$$

First column of $[L]_{\beta \leftarrow \beta} = [I_p \circ L \circ A_\beta e_1]$

$$= [L(e_1)]_\beta$$

So i^{th} column of $[L]_{\beta \leftarrow \beta} = [L(e_i)]_\beta$

So

$$1^{st} \text{ col of } [L]_{\beta \leftarrow \beta} = (2 \ 0 \ 0 \ 0)^T$$

$$2^{nd} \text{ col of } [L]_{\beta \leftarrow \beta} = (1 \ 2 \ 0 \ 0)^T$$

$$3^{rd} \text{ col of } [L]_{\beta \leftarrow \beta} = (-2 \ 2 \ 2 \ 0)^T$$

$$4^{th} \text{ col of } [L]_{\beta \leftarrow \beta} = (0 \ 0 \ 3 \ 2)^T$$

$$\text{So } [L]_{\beta \leftarrow \beta} = \begin{pmatrix} 2 & 1 & -2 & 0 \\ 0 & 2 & 2 & 0 \\ 0 & 0 & 2 & 3 \\ 0 & 0 & 0 & 2 \end{pmatrix}$$

Side note

$$\alpha [L]_{\beta \leftarrow \beta} + [M]_{\beta \leftarrow \beta} = [\alpha L + M]_{\beta \leftarrow \beta}$$

$$[L]_{\beta \leftarrow \beta} \circ [M]_{\beta \leftarrow \beta} = [LM]_{\beta \leftarrow \beta}$$

Notice

~~$L(V) \rightleftarrows [L]_{\beta \leftarrow \beta}$~~ $[L]_{\beta \leftarrow \beta}$ is bijective

Note

$$L \in \mathcal{L}(V) \longrightarrow M_n(\mathbb{C}) \ni [L]_{\beta \leftarrow \beta}$$

This map is isomorphic. But there's also multiplication.

\Rightarrow called an algebra ... $\mathcal{L}(V) \sim M_n$ are algebras...

BLOCK MATRICES

Feb 28, 2019

Start with Cartesian product of 2 vector spaces

$$\mathbb{V} \times \mathbb{W} \xrightarrow{\text{Lin}} \mathbb{V} \times \mathbb{W}$$

$$\begin{pmatrix} v \\ w \end{pmatrix} \xrightarrow{\downarrow} \begin{pmatrix} v \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ w \end{pmatrix}$$

So formal...

$$\left[\begin{array}{|c|c|} \hline F_{11} & F_{12} \\ \hline F_{21} & F_{22} \\ \hline \end{array} \right] \begin{pmatrix} v \\ w \end{pmatrix} \triangleq \begin{pmatrix} F_{11}v + F_{12}w \\ F_{21}v + F_{22}w \end{pmatrix}$$

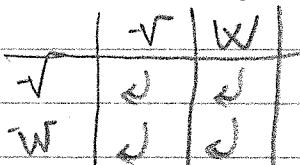
so

$$\left. \begin{array}{l} F_{11} : \mathbb{V} \mapsto \mathbb{V} \\ F_{12} : \mathbb{W} \mapsto \mathbb{V} \end{array} \right\} \text{ since } F_{11}(v) + F_{12}(w) \in \mathbb{V}$$

and

$$\left. \begin{array}{l} F_{21} : \mathbb{V} \mapsto \mathbb{W} \\ F_{22} : \mathbb{W} \mapsto \mathbb{W} \end{array} \right\} \text{ since } F_{21}(v) + F_{22}(w) \in \mathbb{W}$$

In diagram



$$\begin{matrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{matrix}$$

If F_{ij} are all linear, then $\begin{matrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{matrix}$ is a linear fn as well

$$\begin{matrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{matrix} : \mathbb{V} \times \mathbb{W} \rightarrow \mathbb{V} \times \mathbb{W}$$

[Q] → can every linear fn be represented this way?

Note $\begin{array}{|c|c|} \hline F_{11} & F_{12} \\ \hline F_{21} & F_{22} \\ \hline \end{array} \begin{pmatrix} v \\ o \end{pmatrix} = \begin{pmatrix} F_{11}(v) \\ F_{21}(v) \end{pmatrix}$

In particular $\pi_1 \left(\bigoplus_L \begin{pmatrix} v \\ o \end{pmatrix} \right) = F_{11}(v)$ (π : projection)

So

Given $L: V \times W \xrightarrow{\text{lin.}} V \times W$

Let $F_{11}(v) := \pi_1 \circ L \begin{pmatrix} v \\ o \end{pmatrix}$

$F_{12}(w) := \pi_1 \circ L \begin{pmatrix} v \\ w \end{pmatrix}$

$F_{21}(v) := \pi_2 \circ L \begin{pmatrix} v \\ o \end{pmatrix}$

$F_{22}(w) := \pi_2 \circ L \begin{pmatrix} v \\ w \end{pmatrix}$

Do we have

$$L = \begin{array}{|c|c|} \hline F_{11} & F_{12} \\ \hline F_{21} & F_{22} \\ \hline \end{array} ?$$

by linearity

Well

$$\left\{ \begin{array}{l} F_{11}(v) + F_{12}(w) = \pi_1 \circ L \begin{pmatrix} v \\ o \end{pmatrix} + \pi_1 \circ L \begin{pmatrix} v \\ w \end{pmatrix} = \pi_1 \circ L \begin{pmatrix} v \\ w \end{pmatrix} \end{array} \right.$$

$$\left. \begin{array}{l} F_{21}(v) + F_{22}(w) = \pi_2 \circ L \begin{pmatrix} v \\ o \end{pmatrix} + \pi_2 \circ L \begin{pmatrix} v \\ w \end{pmatrix} = \pi_2 \circ L \begin{pmatrix} v \\ w \end{pmatrix} \end{array} \right\}$$

So $\begin{pmatrix} F_{12}(w) + F_{21}(v) \\ F_{11}(v) + F_{22}(w) \end{pmatrix} = \begin{pmatrix} \pi_1 \circ L \begin{pmatrix} v \\ w \end{pmatrix} \\ \pi_2 \circ L \begin{pmatrix} v \\ w \end{pmatrix} \end{pmatrix} = L \begin{pmatrix} v \\ w \end{pmatrix}$ obviously...

So If we start with any L , we can represent L as a matrix of linear functions ...

But we don't often work with Cartesian products. So we want to use this idea to break a vector space into direct sum

What about direct sums?

Bad notation Given $W \oplus Z = V$. Let us write $\begin{pmatrix} w \\ z \end{pmatrix}$ instead of $w+z$

$\hookrightarrow \begin{pmatrix} w \\ z \end{pmatrix}_+ \sim w+z$ so that we can mimick our previous idea

Result

	w	z
w	\downarrow	\downarrow
z	\downarrow	\downarrow

Given $F_{11} : W \mapsto W$

$F_{12} : Z \mapsto W$

$F_{21} : W \mapsto Z$

$F_{22} : Z \mapsto Z$

same as

we can define linear fun

\int before...

$$\begin{bmatrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{bmatrix} : V \mapsto V \text{ leg}$$

$$\begin{bmatrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{bmatrix} \begin{pmatrix} w \\ z \end{pmatrix}_+ = \begin{pmatrix} \dots \\ \dots \end{pmatrix}_+$$

Now, we want to find something that is similar to Π_i (projection)

→ the answer is idempotent (analogous to projections)

Think

$$W \oplus Z = V \quad \text{where } E_i \text{ are idempotents.}$$

$$\| \quad \| \quad \text{Ran}(E_1) \quad \text{Ran}(E_2) \quad \text{where } E_2 = I - E_1$$

If $L = \begin{bmatrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{bmatrix}$, then

$$L \begin{pmatrix} w \\ z \end{pmatrix}_+ = \begin{pmatrix} F_{11}(w) \\ F_{21}(w) \end{pmatrix}_+ \in W \quad \stackrel{\text{def}}{=} E \circ L(w) = F_{11}(w) \in W$$

$$(I-E) \circ L(w) = F_{21}(w) \in Z$$

Same thing $L \begin{pmatrix} 0 \\ z \end{pmatrix}_+ = \begin{pmatrix} F_{12}(z) \\ F_{22}(z) \end{pmatrix}_+$

So $E(L(z)) = F_{12}(z) \in W$

$(I - E)L(z) = F_{22}(z) \in \mathbb{C}^2$

So we can again check that

$$L = \begin{array}{|c|c|c|c|} \hline & \overset{W}{\text{W}} & & \overset{Z}{\text{Z}} \\ \hline & E \circ L & |_W & E \circ L & |_Z \\ \hline & (I - E) \circ L & |_W & (I - E) \circ L & |_Z \\ \hline \end{array}$$

restricted maps...

◻ if " $L : V \rightarrow V$ and wrt to the composition $V = W \oplus Z$
 L has the block form $\begin{array}{|c|c|} \hline A & B \\ \hline C & D \\ \hline \end{array}$ "

then it means

$$A : E \circ L \Big|_W : W \rightarrow W \text{ etc}$$

where E is an idempotent with $\text{Im}(E) = W$
 $\text{ker}(E) = Z$

◻ Caution direct sum is commutative
But here order matters!

[Order of the direct sum matters in our exploration here]

Example ◻ ex $G : \text{idem } V \rightarrow V$, $V = \text{Ran } G \oplus \text{ker } G$

So $G = \begin{array}{|c|c|} \hline \text{Im } G & \text{ker } G \\ \hline \text{Im } G & I_{\text{Im } G} \quad 0 \\ \hline \text{ker } G & 0 \quad 0 \\ \hline \end{array}$

Properties of Block-representation ...

Algebra of Block-matrices

→ All the block-matrices go from $\mathbb{V} \rightarrow \mathbb{V}$ and $\mathbb{V} = W \oplus Z$

$$\textcircled{1} \quad \begin{array}{|c|c|} \hline A & B \\ \hline C & D \\ \hline \end{array} + \begin{array}{|c|c|} \hline P & Q \\ \hline R & T \\ \hline \end{array} = \begin{array}{|c|c|} \hline w & z \\ \hline ?_{11} & ?_{12} \\ \hline ?_{21} & ?_{22} \\ \hline \end{array} \quad W = \text{Im}(E), E^2 = E \\ L \qquad \qquad \qquad S \qquad \qquad \qquad (L+S) \quad Z = \text{Ker}(E)$$

$$?_{11}(w) = E[(L+S)(w)] = EL(w) + ES(w) = A(w) + P(w)$$

So... The algebra is very simple :-)

$$\begin{array}{|c|c|} \hline A & B \\ \hline C & D \\ \hline \end{array} + \begin{array}{|c|c|} \hline P & Q \\ \hline R & T \\ \hline \end{array} = \begin{array}{|c|c|} \hline A+P & B+Q \\ \hline C+R & D+T \\ \hline \end{array}$$

$$\textcircled{2} \quad \underline{\text{Scaling also works...}} \quad \alpha \begin{array}{|c|c|} \hline A & B \\ \hline C & D \\ \hline \end{array} = \begin{array}{|c|c|} \hline \alpha A & \alpha B \\ \hline \alpha C & \alpha D \\ \hline \end{array}$$

$$\textcircled{3} \quad \underline{\text{Composition}} \quad \begin{array}{|c|c|} \hline A & B \\ \hline C & D \\ \hline \end{array}_L \circ \begin{array}{|c|c|} \hline P & Q \\ \hline R & T \\ \hline \end{array}_S = \begin{array}{|c|c|} \hline w & z \\ \hline ?_{11} & ?_{12} \\ \hline ?_{21} & ?_{22} \\ \hline \end{array}_{L \circ S}$$

$$?_{11}(w) = E[L \circ S(w)] = ELS(w)$$

$$= EL[E + (\text{id} - E)]S(w)$$

$$= ELES(w) + EL(\text{id} - E)S(w)$$

~~$$= ELE(ES(w)) + EL(\text{id} - E)(\text{id} - E)S(w)$$~~

$$= \underbrace{ELP(w)}_{\uparrow} + \underbrace{ELR(w)}_{\downarrow}$$

$$= AP(w) \stackrel{w}{=} B(R(w)) \stackrel{z}{=}$$

$$\text{for } ?_{11}(w) = [A \circ P + B \circ R](w) \rightarrow \text{multiply like matrix.}$$

2

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} \circ \begin{bmatrix} P & Q \\ R & T \end{bmatrix} = \begin{bmatrix} A \circ P + B \circ R & A \circ Q + B \circ T \\ C \circ P + D \circ R & C \circ Q + D \circ T \end{bmatrix}$$

~~H~~

(4) Suppose I have $\begin{bmatrix} A & B \\ C & D \end{bmatrix} \in L$. If I pick a basis β for W and γ for Z

then

$$[L]_{\beta \mid Y \leftarrow P \mid X} = \begin{bmatrix} c_1 & \dots & \end{bmatrix}$$

concatenate

$$c_i = [L(b_i)]_{\beta \mid Y} \quad b_i \rightarrow \text{first vector of } P_\beta$$

$$\begin{aligned} &= \left[\begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{pmatrix} b_1 \\ 0 \end{pmatrix} \right]_{\beta \mid Y} = \left[\begin{pmatrix} A(b_1) \\ C(b_1) \end{pmatrix} \right]_{\beta \mid Y}^{\in W} \quad A(b_i) = \text{lin comb of } b_i, \\ &\qquad\qquad\qquad \text{C}(b_i) = b_i \text{ or } f_i g_i \\ &= \left[\begin{bmatrix} A(b_1) \\ C(b_1) \end{bmatrix}_\beta \right]_{\gamma}^Z \quad \text{same thing} \end{aligned}$$

If we keep doing this ... $\rightarrow \underline{\text{Note}} \quad W \sim P, Z \sim Y$

$$[L]_{\beta \mid Y \leftarrow P \mid Y} = \begin{bmatrix} [A]_{\beta \leftarrow P} & [B]_{\beta \leftarrow Y} \\ [C]_{Y \leftarrow P} & [D]_{Y \leftarrow Y} \end{bmatrix} \quad \underline{\text{Note}} \quad [C]_{Y \leftarrow P} \text{ is jumbled}$$

- (1) Take 2nd in beta
- (2) $C(b)$

$$\rightarrow [C]_{Y \leftarrow P} = \left[[C(b_1)]_Y, [C(b_2)]_Y, \dots, [C(b_n)]_Y \right]^T \quad \text{(3) coord in } Y$$

Representing linear functions as matrix-like stuff...

Mar 5, 2019

$$d: V_1 \oplus V_2 \oplus V_3 \xrightarrow{\quad} W_1 \oplus W_2$$

	v_1	v_2	v_3
w_1	d_{11}	d_{12}	d_{13}
w_2	d_{21}	d_{22}	d_{23}

$$d_{13} = E_1 \circ d \Big|_{V_2} = E_1 \circ d \circ F_2 \Big|_{V_2}$$

Back to idempotents

Consider idempotent matrices. $E = E^2 \in M_n$

Ex $\begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$ is such a matrix.

↳ we know that $\text{Im}(E) \oplus \text{ker}(E) = \mathbb{C}^n$

$$\text{so } E: \text{Im}(E) \oplus \text{ker}(E) \xrightarrow{\quad} \text{Im}(E) \oplus \text{ker}(E)$$

So we can represent E as

$$\begin{array}{cc} E & I-E \\ \hline \text{Im}E & \text{ker}E \\ \hline (E) \text{ Im}E & \begin{matrix} E_{11} & E_{12} \\ E_{21} & E_{22} \end{matrix} & E_{12} = E \circ E \Big|_{\text{ker}E} = 0 \\ (I-E) \text{ ker}E & \end{array}$$

$$E_{22} = (I-E) \circ E \Big|_{\text{ker}E} = 0$$

$$E_{11} = E \circ E \Big|_{\text{Im}E} = \text{Id}$$

$$E_{21} = (I-E) \circ E \Big|_{\text{Im}E} = 0$$

$$\text{So } E = \begin{bmatrix} \text{Id} & 0 \\ 0 & 0 \end{bmatrix}$$

Flea
Can pick a basis for $\text{Im}(E)$:
 $\hookrightarrow \text{Im}(E)$

→ then concatenate to get basis for whole space

$$[E]_{P_1 \parallel P_2 \leftarrow P_1 \parallel P_2} = \begin{bmatrix} [\text{Id}]_{P_1 \times P_1}, [0]_{P_2 \times P_2} \\ [0]_{n-r \times r}, [0]_{P_2 \times P_2} \end{bmatrix}$$

Now $[Id]_{P_i \leftarrow P_j} = I$ $\underline{\text{So}} \quad E = \boxed{\begin{array}{|c|c|} \hline I & O \\ \hline O & O \\ \hline \end{array}}$

 $[O]_{P_i \leftarrow P_j} = O$

(1) So I can write $[E]_{P_0 \leftarrow P_0} = E = E^2 \in M_n$

\uparrow standard basis

(2) But we also know:

$[E]_{P_1 \sqcup P_2 \leftarrow P_1 \sqcup P_2} = \boxed{\begin{array}{|c|c|} \hline I & O \\ \hline O & O \\ \hline \end{array}}$

From (1) + (2), we have a theorem Every idempotent $E \in M_n$ is similar to a matrix of the form

$$\begin{pmatrix} 1 & & & & 0 \\ & 1 & & & \\ & & \ddots & & \\ & & & 0 & \\ k & & & & 0 & 0 & 0 \end{pmatrix} \quad \text{where } k = \text{Rank}(E) = \text{Tr}(E)$$

So $\text{Rank}(E) = \text{Trace}(E)$

Case 2 } $\{ \Rightarrow$ Since $\text{Rank}(E) = \text{Trace}(E)$, if $\text{Trace}(M)$ is not an integer or negative... then M is NOT idempotent

(PRE-) **Corollary** Suppose $E_1 + E_2 + \dots + E_m = I_{\#n}$

m idempotents

Then

$\text{Trace}(E_1 + E_2 + \dots + E_m) = \text{Trace}(I_{\#n}) = n$

$\Rightarrow \text{Tr}(E_1) + \text{Tr}(E_2) + \dots + \text{Tr}(E_m) = n$

∴

$\text{Rank}(E_1) + \text{Rank}(E_2) + \dots + \text{Rank}(E_m) = n$

Now, consider $\text{Im}(E_1 + \dots + E_m)$

and

$$\text{Im } E_1 + \text{Im } E_2 + \dots + \text{Im } E_m$$

Consider $(E_1 + \dots + E_m)(x) = E_1 x + E_2 x + \dots + E_m x \in \sum \text{Im } E_i$

$$\text{So } \mathbb{C}^n = \text{Im}(I_{\mathbb{C}^n}) = \text{Im}\left(\sum_{i=1}^m c_i\right) \subseteq \sum_{i=1}^m \text{Im } c_i \subseteq \mathbb{C}^n$$

$$\therefore \sum_{i=1}^m \text{Im } E_i = \mathbb{C}^n \quad (\text{i})$$

$$\begin{aligned} \text{But we also know } & \sum_{i=1}^m \dim(\text{Im}(E_i)) \\ &= \sum_{i=1}^m \text{rank}(E_i) = n \quad (\text{ii}) \end{aligned}$$

From (i) & (ii), then

$$\boxed{\bigoplus_{i=1}^m \text{Im}(E_i) = \mathbb{C}^n} \leftarrow \text{direct sum.}$$

In particular,

$$\dim\left(\bigoplus_{i=2}^m \text{Im}(E_i)\right) = n - \dim(\text{Im } E_1)$$

$$= \text{nullity}(E_1) = \dim(\ker E_1)$$

Now, $\text{Im}\left(\sum_{i=2}^m E_i\right) = \text{Im}(I - E_1)$

Now $\bigoplus_{i=2}^m \text{Im}(E_i) = \ker(E_1)$

Since $\ker(E_1) \subseteq \bigoplus_{i=2}^m \text{Im}(E_i)$
 But $\dim(E_1) = \dim\left(\bigoplus_{i=2}^m \text{Im}(E_i)\right)$

$$\boxed{\ker(E_1) = \bigoplus_{i=2}^m \text{Im}(E_i)}$$

◻ This implies $\text{Im } E_2 \subseteq \ker E_1$

$$\hookrightarrow E_1(E_2(\square)) = O_n + \square$$

$$\text{Hence } E_1 \circ E_2 = O$$

◻ So, in general

$$E_i \circ E_j = O_{n \times n} \text{ if } i \neq j$$

So {Corollary}

If E_1, \dots, E_m are idempotents and

Note

$$\bigoplus_{i=2}^m \text{Im } E_i = \ker(E_1)$$

$$\sum_{i=1}^m E_i = I, \text{ then } \left\{ \begin{array}{l} E_i E_j = O \text{ for } i \neq j \\ \bigoplus_{i=1}^m \text{Im}(E_i) = \mathbb{C}^n \end{array} \right.$$

Note $\text{Trace}(AB) = \text{Trace}(BA)$

$$\text{So } \left\{ \begin{array}{l} \text{Trace}(S^{-1}AS) = \text{Trace}(ASS^{-1}) \\ \text{Trace}(B) = \text{Trace}(A) \end{array} \right\} \left\{ \begin{array}{l} \text{Trace}(AB) = \text{Trace } b_A \cdot b_B = \text{Trace}(B^T A) \\ (\Rightarrow \text{use indices.}) \end{array} \right.$$

So similar matrices have the same trace & rank.

◻ Now, let's revisit $E \rightarrow$ what we've done is change of basis to get a better matrix (singular...)

◻ Now, suppose $d: V \xrightarrow{\text{lin}} V \rightarrow$ finite dim

$$\text{Suppose } \{O\} \neq W \in \text{Lat}(d)$$

\oplus

V

$$V = W \oplus \boxed{?}$$

Let $w_1, \dots, w_m \in W$ basis of W , then $\text{span}\{w_1, \dots, w_m\} \rightarrow$ basis of

◻

(41)

Now think $\mathcal{S} v_1 \dots v_{37} | v_{10} \dots v_{15}$

$$\text{Then } \text{Span}(v_1 \dots v_{37}) + \text{Span}(v_{38} \dots v_{15}) = V$$

$$\text{base } V = \sum_{i=1}^{37} a_i v_i + \sum_{i=10}^{15} a_i v_i$$

$$w_1 \quad w_2 \quad (\checkmark)$$

$$\cap \quad \cap$$

$$\text{So } \text{Span}(v_1 \dots v_{37}) \oplus \text{Span}(v_{38} \dots v_{15}) = V \quad \left. \begin{array}{l} \\ \end{array} \right\}$$

$$\text{Since } \dim W_1 \times \dim W_2 = V$$

$$E \xrightarrow{\quad \quad \quad}$$

$$\text{Now } V = W \oplus Z \quad (W \in \text{Lat}(E))$$

$$\text{Then } W \quad \begin{array}{|c|c|} \hline & & \\ \hline & & \\ \hline \end{array} \quad ? = (I-E) \circ \delta \Big|_W$$

$$(I-E) \circ \delta \quad \begin{array}{|c|c|} \hline & 0 \\ \hline 0 & \\ \hline \end{array} \quad = 0$$

$$\text{Now } \delta(w) \in W = \text{Im } E = \text{ker}(I-E)$$

$$\rightarrow (I-E) \circ \delta \Big|_W = 0$$

In fact, converse also true

$$(I-E) \circ \delta \Big|_W = 0 \Rightarrow \text{Im}(\delta|_W) \subseteq \text{ker}(I-E) = \text{Ran}(E) = W$$

$$\text{So } W \in \text{Lat}(\delta)$$

So

Now, consider

So $\begin{matrix} w \\ z \end{matrix} \in \boxed{\begin{array}{|c|c|} \hline & w \\ \hline w & z \\ \hline \end{array}}$ $\Leftrightarrow w, z \in \text{Lat}(L)$

Observe $\begin{matrix} w \\ z \end{matrix} \in \boxed{\begin{array}{|c|c|} \hline & w \\ \hline w & z \\ \hline \end{array}} \Leftrightarrow z \in \text{Lat}(L)$

So if $w, z \in \text{Lat}(L)$, $w \otimes z = v$

then $\begin{matrix} w \\ z \end{matrix} \in \boxed{\begin{array}{|c|c|} \hline & w \\ \hline A & 0 \\ \hline 0 & B \\ \hline \end{array}}$ \rightarrow in essence, there's no mixing
 $w = z$

Q: How do we know there's an invariant subspace to start?

Given $L: V \rightarrow V \rightarrow \text{Span}(v_1, \dots, v_m)$
 \hookrightarrow How do I find a good invariant subspace.

We can construct one... for $v_0 \neq 0$, we want $v_0 \in W \in \text{Lat}(L)$

Want $v_0 \in W$

$L(v_0) \in W$ If $m = \dim(V)$

$L^1(v_0) \in W$ Then there are at most m

$L^2(v_0) \in W$ things...

$L^n(v_0) \in W$

Let $L^k(v_0)$ be the k^{th} one that is a lin. comb. of previous ones, i.e.

$$L^n(v_0) = \sum_{i=0}^{n-1} a_i L^i(v_0)$$

$$\text{Then } L^{n+1}(v_0) = L\left(\sum_{i=0}^n a_i L^i(v_0)\right)$$

$$= \sum_{i=0}^n a_i L^{i+1}(v_0) \quad a_i \in \text{Span}(L^i(v_0))$$

$i=1 \dots n-1$

In other way

Consider $(\lambda^3 - \lambda_1 \lambda^2 - \lambda_1 \lambda' - \lambda_2 \lambda'^2) v_0 = 0 \rightarrow$ complex r

$$\text{So } (\lambda - r_1 I)(\lambda - r_2 I)(\lambda - r_3 I)(v_0) = 0 \quad \left. \begin{array}{l} \\ \\ \text{but since } v_0 \neq 0 \end{array} \right\}$$

At least one of $\lambda - r_1 I$, $\lambda - r_2 I$, $\lambda - r_3 I$ is NOT injective.

If $\lambda - r_1 I$ not injective

$$\Rightarrow L(\vec{v}) = r_1 \vec{v} \quad \text{for some } \vec{v} \neq 0$$

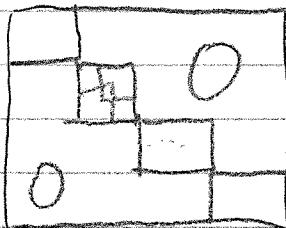
\uparrow eigenvector
eigenvalue

on V

We have just proved, every linear form in $\dim V$ space has an eigenvector!

\hookrightarrow \exists 1 dim invariant subspace.

So we can always have $\{0\} \neq W \neq V$



~~the~~

Mar 7, 2019

Idea

$$P \xrightarrow{\quad} "P(L)"$$

$$c_0 + c_1 z + \dots + c_n z^n \quad \xrightarrow{\quad} c_0 I + c_1 \lambda + \dots + c_n \lambda^n$$

$(\lambda: V \xrightarrow{c_n} V)$

$$P \xrightarrow{P_L} L(V)$$

What sort of properties does this map have?

Properties of $R \xrightarrow{L} L(V)$

① Linear

② Multiplicative (product of linear polynomial \rightarrow products of lin. fn.)

Consider something from previous proof

$$\rightarrow (-\lambda_0 I - \lambda_1 d - \dots - \lambda_n d^n)(v_0) = 0$$

$$\hookrightarrow (P(L))(v_0) = 0,$$

By Fundamental theorem of algebra,

unique ... $\rightarrow P(z) = (z - \lambda_1)^{m_1} \cdots (z - \lambda_k)^{m_k}$, λ_k are the roots
 m_k multiplicity of root λ_k

So since P_L multiplicative

$$\text{Hence } (d - \lambda_1 I)^{m_1} \cdots (d - \lambda_k I)^{m_k}$$

$$\text{So } (d - \lambda_1 I)^{m_1} \cdots (d - \lambda_k I)^{m_k}(v_0) = 0$$

i.e. $(d - \lambda_1 I) \underbrace{\cdots}_{m_1} (d - \lambda_k I) \underbrace{\cdots}_{m_k} (d - \lambda_{k+1} I) \cdots (d - \lambda_n I)(v_0) = 0$

So $(d - \lambda_1 I) \cdots (d - \lambda_k I) \cdots (d - \lambda_{k+1} I) \cdots (d - \lambda_n I)$
not injective for at least one i ,

i.e. there exists some $v_i \neq 0$ such that $(d - \lambda_i I)(v_i) = 0$,

$$\text{or } L(v_i) = \lambda_i v_i$$

So one of λ_i 's is an eigenvalue of L .

finite dim

(*) Every $L: V \xrightarrow{\text{fin.dim}} V$ has an eigenvalue

Every $L: V \xrightarrow{\text{fin.dim}} V$ has a 1-dimensional invariant subspace

Consider $V = W_1 \oplus \mathbb{Z}_2$ V 2-dimensional

(1-dim) $\text{Lat}(L)$

Then

$$L = \begin{matrix} W_1 & \xrightarrow{\text{1-dim}} \\ \mathbb{Z}_2 & \end{matrix} = \begin{pmatrix} "1" & B \\ 0 & D \end{pmatrix}$$

What about D ? $D: \mathbb{Z} \rightarrow \mathbb{Z}$ or ~~$\mathbb{Z} \rightarrow \mathbb{Z}$~~

If W_1 not one-dimensional, then $D: \mathbb{Z} \rightarrow \mathbb{Z}$, then

there is an $\mathbb{Z} = W_2 \oplus U$

1-dim invariant, $\in \text{Lat}(D)$

Then

$$L = \begin{matrix} W_1 & \mathbb{Z} \\ \mathbb{Z} & D \end{matrix} = \begin{matrix} W_1 & W_2 & U \\ 0 & "1" & B_1 \\ 0 & 0 & "1" \\ \vdots & \vdots & \vdots \\ 0 & 0 & D \end{matrix} = \dots$$

UPPER TRIANGULAR MATRIX

Schur's theorem

Every $n \times n$ matrix is similar to an upper triangular matrix

Proof by induction Base case $n=1 \rightarrow$ trivial

Suppose n_0 is smallest size for which there is a counterexample $A_{(n_0 \times n_0)}$

- Now, $A: \mathbb{R}^{n_0} \rightarrow A$ has an eigenvector by (*), called w_0 and $\text{span}(w_0)$ is invariant under A . Then

$$\rho^n = \text{Span}(\omega) \oplus z$$

w	z
w	z

w.r.t to this decomposition, A has the form $\begin{bmatrix} P_2 & P_1 \\ w_0 & 0 \end{bmatrix}$

Pick a basis of $z : w_0, \dots, w_{n_0-1}, \overbrace{w_0}^{P_1} = T$

A

w_0

\vdots

w_{n_0-1}

(1)

$$\text{recall } [A]_P = \begin{bmatrix} [A(w_0)]_P & \dots & [A(w_{n_0-1})]_P \end{bmatrix}$$

So

P_1

P_2

$$= \begin{bmatrix} A \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad B$$

$$[A]_P =$$

$$(1) P_1 \begin{array}{|c|c|} \hline 0 & \text{[T]} \\ \hline \end{array} \quad P_2 \rightarrow P_2$$

A is the smallest
quintic exp

$$\text{So } A \sim [A]_P$$

Since B smaller than A

$$B = S \circ T \circ S^{-1} \text{ for some}$$

But A is a counterexample \rightarrow

replace S, T

So

$$A \sim \begin{bmatrix} 0 & M \\ 0 & STS^{-1} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & S^{-1} \end{bmatrix}$$

$$\begin{bmatrix} 0 & MS^{-1} \\ T & 0 \end{bmatrix}$$

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

↑ ↓
increases

So

$$\begin{bmatrix} 0 & MS^{-1} \\ T & 0 \end{bmatrix}$$

\rightarrow upper triangular \rightarrow contradiction

Now, back to polynomials...

① For each $v_0 \in V$ there is some $P \neq Q_0 \in P$ such that

$$\overset{\#}{\underset{v_0}{\text{O}}} \quad (P(L))(v_0) = 0$$

Consider

$$P + \xrightarrow{P_L} P(L)$$

$$\left\{ P \in P \mid (P(L))(v_0) = 0 \right\} \subseteq P$$

Closed under scaling { ↑ closed under scaling

$$\text{if } (P(L))(v_0) = 0 \text{ then } (B \cdot P(L))(v_0) = 0$$

in particular

$$(g(L) \circ P(L))(v_0) = 0, \text{ for any } g \in P$$

i.e.

$$(g \cdot P)(L)(v_0) = 0 \text{ for any } g \in P$$

Closed under addition { ↑ closed under addition

↳ **Ideals** ← subspace that is closed under multiplication

There's the smallest unique non-zero polynomial

unique $P(L)$ smallest that kills (v_0)

Now, each v_0 , can associate a polynomial to it.

Is there a polynomial that kills them all?

Consider

$$V \xrightarrow{d} \mathbb{F}$$

basis + multiplication

$$\mathbb{C}^n \xrightarrow{[L]_{P \in P}} \mathbb{C}^n$$

$$[P(S)]_{P \in P} = P \left(\sum [L]_{P \in P} \right)$$

$$\text{Result } [L^2]_{P \in P} = [L^2]_{P \in P}$$

$$= [L]_{P \in P} \circ [L]_{P \in P} = [L]^2_{P \in P}$$

$$\text{Case where add } [d^2 + d]_{P \in P} = [dJ]_{P \in P} + [dJ]_{P \in P}$$

Relate to this

$$[P(A)]_{P \in P} = P([d]_{P \in P})$$

To will discuss for \rightarrow will matrix $\in M_{n \times n}$

But a matrix is just a $v_0 \in V$

$$\text{So stronger or later } A^k - gA^{k-1} - \dots = \theta_{n \times n}$$

$$\text{Look at } \{P \in P \mid (P(A)) = \theta_{n \times n}\}$$

\hookrightarrow an ideal again ... (

\hookrightarrow \exists single minimalmonic element that kills $A \dots$

But if $P(A) = \theta_{n \times n}$ then $(P(A))(v_0) = \theta_v$

$$\hookrightarrow P(A) \in \{P \in P \mid (P(A))(v_0) = \theta_v\}$$

So all $P(A)(v_0)$ are divisor of $P(A)$ after $P(A) = \theta_{n \times n}$

Mar 12, 2019

Review Let $M \in M_n = n^2$ -dimensional vector space (h)

Look at I, M, M^2, M^3, \dots eventually come to the first power of M such that ~~that~~ \rightarrow smallest $k_0 \in \mathbb{N}$ such that M^{k_0} is a lin. combination of $I, M, M^2, \dots, M^{k_0-1}$

$$\text{Note } k_0 \leq n^2 \quad | \quad M^{k_0} = \sum_{i=1}^{k_0} a_i M^i \quad \hookrightarrow$$

So $M^{k_0} + (-a_{k_0-1})M^{k_0-1} + (-a_{k_0-2})M^{k_0-2} + \dots + (-a_0)I = \Theta = P(M)$

→ This polynomial annihilate matrix M . In fact, it is the smallest polynomial to do so.

degree (monic)

Now

for any matrix we can do this → now look at the set of all polys that annihilate M .

$$W = \{ p \in P \mid P(M) = \Theta \}$$

This is a vector space. Recall: $P \xrightarrow[\text{linear multiplicative}]{} M_n$

$$\text{subspace description property } P \xrightarrow[\text{core}]{} P(M)$$

So W is also an ideal. Why? → singly generated.

→ We say that it is generated by a single monic polynomial.

Called: $\boxed{\mu_M \rightarrow \text{the minimal polynomial of } M}$

Recall that

$$Z = \{ p \in P \mid (P(M))(v_0) = \Theta \} \text{ is also an ideal of } P$$

[Here $M \in M_n$, $\Theta \neq v_0 \in \mathbb{C}^n$]

μ_M is a multiple
of μ_{M,v_0} .

local → and we have the minimal poly μ_{M,v_0} as the generator of this ideal.

Observe that if $(\mu_M(M))(v_0) = \Theta(v_0) = \Theta$

{
So, μ_{M,v_0} divides μ_M
- $Z \subseteq W$ }

So the global μ_M is in Z , generated by μ_{M,v_0} so $- Z \subseteq W$

Also note that a polynomial has at most finitely many monic divisors

$$\text{Suppose } \mu_M(z) = (z - r_1)^{\gamma_1} \cdots (z - r_k)^{\gamma_k}$$

$$\text{Consider } \mu_{M, V_i}(z), \dots, \mu_{M, V_{135}}(z)$$

$$\text{Consider } \left\{ v \mid (\mu_{M, V_i}(M))(v) = 0 \right\} = \ker(\mu_{M, V_i}(M))$$

= subspace

The union of these 135 subspaces is \rightarrow the whole space. So, by PSET 2, one of these subspaces equals the whole union, i.e. equals the whole space.

$$\text{So } \mu_{M, V_i}(M) = 0 \text{ for at least one } i$$

i.e. μ_{M, V_i} annihilates M

But μ_{M, V_i} divides $\mu_M = q \cdot \mu_{M, V_i}$

$$\text{So } \boxed{\mu_{M, V_i} = \mu_M \text{ for at least one } i}$$

Why do we want to look at these polynomials?

$$\text{Suppose } \mu_A(z) = (z - \alpha_1)^{\gamma_1} \cdots (z - \alpha_k)^{\gamma_k}$$

$$\text{Then } (A - \alpha_1 I)^{\gamma_1} \cdots (A - \alpha_k I)^{\gamma_k} = 0$$

But none of these factors can be invertible (proof by contradiction)

$$\hookrightarrow \text{So } \cancel{A - \alpha_1 I \neq 0, \dots, A - \alpha_k I \neq 0} \rightarrow (A - \alpha I)v = 0 \Rightarrow A(v) = \lambda v \rightarrow v \neq 0$$

$$\text{So } \boxed{\text{All roots of } \mu_A \text{ are eigenvalues of } A}$$

Theorem 12

(51)



Are there other eigenvalues that are Not roots of μ_A ?

Consider $(z-\lambda)$

→ eigen-val of A and $\lambda \neq \alpha$;

(No)

So $\mu_A(z)$ and $(z-\lambda)$ relatively prime. $\rightarrow \exists p, q$ such that

$$p(z)(z-\lambda) + q(z)\mu_A(z) = 1$$

$$\text{So } p(A)(A - \lambda I) + q(A)(\mu_A(A)) = I$$

$$\underbrace{\quad}_{\alpha} \quad \underbrace{\quad}_{\text{ker}(A - \lambda I)} = \{0\}$$

$$\text{So } p(A)(A - \lambda I) = I$$

$$\text{ker}(A - \lambda I) = \{0\}$$

So $(A - \lambda I)$ invertible $\rightarrow \lambda$ not eigenvalue of A
 $(f^{-1}(A) = (A - \lambda I)^{-1})$

So All roots of μ_A are exactly eigenvalues of A

Much easier than finding eigenvalues using characteristic polynomials
 Since the degrees are often much smaller.

Recap

$$\mu_A(z)$$

$$I, A, A^2, \dots$$

$$\mu_{A,0}(z)$$

$$v, Av, A^2v, \dots$$

at most n^2

at most n

characteristic polynomial

But $\mu_{A,0}(z)$ divides $\mu_A(z)$

So Then: [For $A \in M_n$, $\deg(\mu_A) \leq n$]

recall $T = \begin{pmatrix} A & 0 \\ 0 & D \end{pmatrix} \quad T^2 = \begin{pmatrix} A^2 & 0 \\ 0 & D^2 \end{pmatrix} \dots \quad T^0 = \begin{pmatrix} I & 0 \\ 0 & I \end{pmatrix}$

so $P(T) = \begin{pmatrix} P(A) & 0 \\ 0 & P(D) \end{pmatrix}$

What is μ_T ?

P annihilates $T \Leftrightarrow P$ annihilates A and D

$\Leftrightarrow \mu_A | P$ and $\mu_D | P$

$\Leftrightarrow P$ is a common multiple of μ_A, μ_D

So $\boxed{\mu_T = \text{LCM}(\mu_A, \mu_D)}$

For a diagonal matrix, $T = \begin{pmatrix} \alpha_1 & & \\ & \ddots & \\ & & \alpha_k \end{pmatrix}$

So $\mu_T = \text{LCM}(z - \alpha_1, z - \alpha_2, \dots, z - \alpha_k)$

So $\boxed{\mu_T = \text{LCM}() = (z - \lambda_1)^{e_1} \dots (z - \lambda_k)^{e_k}}$

where $\lambda_i \neq \lambda_j \neq \dots \neq \lambda_s$

some α_i can be repeated
power 1



Diagonalizability

Recall $\sqrt[n]{\lambda} \rightarrow \sqrt[p]{\lambda} \Rightarrow \boxed{P(C)}_{P \in P} = P\left(\left[\lambda\right]_{P \in P}\right)$

(we shown) $\begin{array}{ccc} \mathbb{C}^n & \xrightarrow{\lambda} & \mathbb{C}^n \\ \downarrow U & \downarrow \text{id} & \downarrow U \\ \mathbb{C}^n & \xrightarrow{\text{id}} & \mathbb{C}^n \end{array}$

i.e. Similar matrices have the same minimal poly.

Q "Polynomials belong to linear fn, not the basis/matrix"

So, for $P\left(\left[\lambda\right]_{P \in P}\right) = 0$, $P(\lambda) = 0$

If $A \in M_n$ and $\alpha_0 I + \alpha_1 A + \dots + \alpha_k A^k = 0$

then by transposition $\rightarrow P(A) = 0$

$$\text{So } \sum_k \alpha_k (A^k)^T = \sum_k \alpha_k (AT)^k = 0$$

So if $P(A) = 0$, then $P(AT) = 0$

But if $P(AT) = 0$, then $P(A) = 0$

So $\boxed{\mu_{AT} = \mu_A}$

What about $T = \begin{pmatrix} A & B \\ 0 & D \end{pmatrix} \Rightarrow T^2 = \begin{pmatrix} A^2 & AB \\ 0 & D^2 \end{pmatrix}$

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

So $P(T) = \begin{pmatrix} P(A) & PB \\ 0 & P(D) \end{pmatrix}$

not least

only \rightarrow So $\{P(T)\} \text{ annihilates } T \Rightarrow P(A) = P(D) = 0$
1 direction $\Rightarrow P = \underline{\text{common multiple of } \mu_A, \mu_D}$

So $(\mu_A \cdot \mu_D)(T) = \mu_A(T) \cdot \mu_D(T)$

$$\begin{pmatrix} 0 & \infty \\ 0 & \mu_A(D) \end{pmatrix} \cdot \begin{pmatrix} \mu_D(A) & \infty \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

So $\mu_A \cdot \mu_D$ annihilates T \Rightarrow $\mu_A | T | \mu_D$

Therefore, $\boxed{\text{LCM}(\mu_A, \mu_D) | T | \mu_A \mu_D}$

$P(A), \mu_A$ *prime, irreducible*
LCM \Rightarrow $LCM(\mu_A, \mu_D) | T | \mu_A \mu_D$

Mar 14, 2019

Review

- (1) Minimal Poly.
- (2) Eigenvectors are exactly the roots of the minimal polynomial
- (3) Local v Global min poly
 - (a) Local divides Global
 - (b) One of ideals is global
- (4) Similar matrices have the same minimal poly.

$$(\mu_L = \mu_{T \circ J_{P \in F}})$$

permutation T of pair A, B

$$(5) \quad M_A = \mu_A T \Rightarrow \sigma_F(T) = \sigma_C(T^T)$$

$$(6) \quad \text{if } T = \begin{bmatrix} A & C \\ 0 & B \end{bmatrix} \text{ then } \mu_T = \text{LCM}(\mu_A, \mu_B) \rightarrow \text{gap multiplicity}$$

$$(7) \quad \text{if } T = \begin{bmatrix} A & D \\ 0 & B \end{bmatrix} \text{ then } [\text{LCM}(\mu_A, \mu_B) | \mu_T | \mu_A, \mu_B]$$

D

$$\sigma_F(T) = \sigma_C(A) \cup \sigma_C(B)$$

both of T are exactly
those that appear in
 μ_A or μ_B (or both)

$$(8) \quad \text{if } \mu_A, \mu_B \text{ relatively prime} \Rightarrow \text{LCM}(\mu_A, \mu_B) = \mu_A \mu_B = \mu_T$$

$$(6a) \quad \mu_T \text{ of } T = \begin{pmatrix} * & \dots & 0 \\ 0 & \ddots & * \\ \vdots & & \vdots \end{pmatrix} \text{ is a product of } (x-1), (x-d_i) \text{ for } i=1, \dots, n$$

distinct

$$\text{Ex: } T = \begin{pmatrix} 3 & 4 & & \\ & 1 & & \\ & & 3 & \\ & & & 4 \end{pmatrix} \rightarrow \mu_T = (z-3)(z-4)(z-1)$$

$$\therefore \sigma_F(T) = \{1, 3, 4\}$$

$$\begin{pmatrix} A & D \\ 0 & B \end{pmatrix} = T = \begin{pmatrix} 3 & 1 & 2 & 8 \\ 4 & 3 & 1 & \\ 2 & 2 & & \\ \hline & 3 & & \end{pmatrix} \Rightarrow \sigma_F(T) = \sigma_C\left(\begin{pmatrix} 3 & 1 \\ 0 & 4 \end{pmatrix}\right) \cup \sigma_C\left(\begin{pmatrix} 2 & 7 \\ 0 & 3 \end{pmatrix}\right)$$

$$= \sigma_C(\{3\}) \cup \sigma_C(\{4\}) \cup \sigma_C(\{2\}) \cup \sigma_C(\{3\})$$

$$= \{3, 4, 2\} \rightarrow \text{just the diagonal!}$$

Q: what is $\mu_T(z)$?

$$\mu_T(z) = (z-2)(z-3)(z-4)$$

$$\text{we have } \text{LCM}(\mu_A, \mu_B) | \mu_T | \mu_A, \mu_B \quad \{ \rightarrow$$

Observe If $T = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ and $p(z) = (z-a)(z-d) - bc$ then p annihilates T

Part $p(T) = (T-aI)(T-dI) - bcI$ is the minimal poly if....

$$= \begin{pmatrix} 0 & b \\ c & d-a \end{pmatrix} \begin{pmatrix} a-d & b \\ c & 0 \end{pmatrix} = \begin{pmatrix} bc & 0 \\ 0 & bc \end{pmatrix} - \begin{pmatrix} bc & 0 \\ 0 & bc \end{pmatrix} = [0]$$

$-bcI$

Thm If $T = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ is not a multiple of I_2 , then $\mu_T =$

$$\mu_T = (z-a)(z-d) - bc$$

Back to example

$$\mu_A = \mu_{\begin{bmatrix} 3 & 1 \\ 0 & 4 \end{bmatrix}}(z) = (z-3)(z-4)$$

$$\text{LCM} = (z-2)(z-3)(z-4)$$

$$\mu_B = \mu_{\begin{bmatrix} 2 & 2 \\ 0 & 3 \end{bmatrix}}(z) = (z-2)(z-3)$$

so $(z-2)(z-3)(z-4) \mid \mu_T \mid (z-2)^2(z-3)(z-4)$

$\therefore \mu_T(z) = (z-2)(z-3)^{1 \text{ or } 2}(z-4)$

Question

$$\begin{pmatrix} 3 & 1 \\ 0 & 4 \end{pmatrix} \xrightarrow{?} \begin{pmatrix} 3 & 0 \\ 0 & 4 \end{pmatrix} \text{ or } \begin{pmatrix} 4 & 0 \\ 0 & 3 \end{pmatrix}$$

well $\begin{pmatrix} 3 & 0 \\ 0 & 4 \end{pmatrix} \sim \begin{pmatrix} 4 & 0 \\ 0 & 3 \end{pmatrix} \leq \begin{pmatrix} 3 & 1 \\ 0 & 4 \end{pmatrix} \xrightarrow{?} \begin{pmatrix} 3 & 0 \\ 0 & 4 \end{pmatrix}$ Yes.

Why? Same minimal polynomial, $(z-3)(z-4)$

\square Suppose $p_1, p_2 \rightarrow$ relatively prime.

$$\text{Then } \ker(P_1(L)) \cap \ker(P_2(L)) = \{\emptyset\}$$

"Proof" $\rightarrow \exists q_1, q_2 \text{ s.t. } q_1(z) p_1(z) + q_2(z) p_2(z) = 1$

$$\therefore q_1(L) p_1(L) + q_2(L) p_2(L) = I$$

$$\therefore (q_1(L) p_1(L) + q_2(L) p_2(L))(v) = v$$

$$\therefore q_1(L)[(p_1(L))(v)] + q_2(L)[(p_2(L))(v)] = v$$

if $v \in \ker(P_1(C)) \cap \ker(P_2(C))$ then

$$q_1(L)\overset{\leftrightarrow}{\phi} + q_2(L)\overset{\leftrightarrow}{\phi} = v \Rightarrow v = \overset{\leftrightarrow}{\phi}$$

$\therefore \{\emptyset\} = \ker(P_1(L)) \cap \ker(P_2(L))$ if p_1, p_2 relatively prime

Observe

$$\text{If } p_A(z) = \underbrace{(z - \lambda_1)}_{P_1}^{\ell_1} \cdots \underbrace{(z - \lambda_m)}_{P_2}^{\ell_m}$$

Since p_1, p_2 relatively prime $\Rightarrow \ker(p_1(A)) \cap \ker(p_2(A)) = \{\emptyset\}$

$$\text{Also } P_1(A) P_2(A) = \mu_A(A) = \emptyset = P_2(A) P_1(A)$$

$$\therefore \begin{cases} \text{Im}(P_1(A)) \subseteq \ker(P_2(A)) \\ \text{Im}(P_2(A)) \subseteq \ker(P_1(A)) \end{cases}$$

$$\text{Yet also } q_1(z) p_1(z) + q_2(z) p_2(z) = 1 \text{ for some } q_1, q_2$$

$$\therefore p_1(z) \cdot q_1(z) + p_2(z) \cdot q_2(z) = 1$$

$$\therefore P_1(A) q_1(A) + P_2(A) q_2(A) = I$$

hence, $\underbrace{P_1(A)(q_1(A)(y))}_{\text{Im}(P_1)} + \underbrace{P_2(A)(q_2(A)(y))}_{\text{Im}(P_2)} = y \quad \forall y \in V$

$\Leftrightarrow \text{Im}(P_1) \subset \ker(P_2(A)) \quad \text{Im}(P_2) \subset \ker(P_1(A))$

So

$$\ker(P_1(A)) + \ker(P_2(A)) = V$$

But since $\ker(P_1(A)) \cap \ker(P_2(A)) = \{0\}$

So $V = \ker(P_1(A)) \oplus \ker(P_2(A))$ $\quad (\#)$

Observe that $P_i(A) \hookrightarrow A$, ($P_i(A)$ commutes with A)

So $\ker(P_1(A)) \in \text{Lat}(A)$

Similarly, $\ker(P_2(A)) \in \text{Lat}(A)$

So, with respect to the decomposition, A can be expressed as

$$A = \begin{bmatrix} C & O \\ O & D \end{bmatrix} = \begin{matrix} \ker(P_1(A)) & \ker(P_2(A)) \\ \ker(P_2(A)) & \ker(P_1(A)) \end{matrix}$$

Observe what are μ_C & μ_D ? $\mu_C = P_1 \quad \{ \rightarrow$ to use help
 $\mu_D = P_2 \quad \text{going}$

[Primary decomposition]

WED 19, 2019

Summarize. If q_1, q_2 relatively prime and q_1, q_2 annihilates $A: V \mapsto V$
 Then $V = \ker(q_1(A)) \oplus \ker(q_2(A))$ and

and this decompn A has the form

$$A = \begin{pmatrix} B & 0 \\ 0 & D \end{pmatrix} \quad \left\{ \begin{array}{l} q_1 \text{ annihilates } B \\ q_2 \text{ annihilates } D \end{array} \right.$$

$\cap \text{Im}(q_2(A)) \subset \ker q_1$

"Proof sketch" $\Rightarrow (1) \cdot q_1(A) q_2(A) = 0 = q_2(A) q_1(A) \Rightarrow \left\{ \begin{array}{l} \text{Im}(q_1(A)) \subset \ker q_2 \\ \text{Im}(q_2(A)) \subset \ker q_1 \end{array} \right.$

$$(2) \quad f \cdot f \cdot q_1 + q_1 \cdot q_2 = \mathbb{1} \Rightarrow q_1(A) f(A) + q_2(A) g(A) = \mathbb{I}$$

$$\text{So } \underbrace{q_1(A) f(A)(x)}_{\begin{array}{l} \cap \\ \text{Im } q_1(A) \end{array}} + \underbrace{q_2(A) g(A)(x)}_{\begin{array}{l} \cap \\ \text{Im } q_2(A) \\ \cap \\ \text{ker } (q_2(A)) \end{array}} = x$$

$$\text{So } \text{ker } (q_1(A)) + \text{ker } (q_2(A)) = V$$

$$(3) \quad \text{for any } y \neq 0, \quad \underbrace{f(A) q_1(A)(y)}_{\text{cannot both be zero unless } y=0} + \underbrace{g(A) q_2(A)(y)}_{\text{in } \text{ker } (q_1(A)) \cap \text{ker } (q_2(A))} = 0$$

$$\rightarrow \text{ker } (q_1(A)) \cap \text{ker } (q_2(A)) = \{0\}$$

$$\text{So } \text{ker } (q_1(A)) \oplus \text{ker } (q_2(A)) = V$$

$$(4) \quad \begin{array}{l} \text{ker } (q_1(A)) \in \text{Lat}(A) \\ \text{ker } (q_2(A)) \in \text{Lat}(A) \end{array} \quad \Rightarrow$$

Please note $V = \text{ker } (q_1(A)) \oplus \text{ker } (q_2(A))$,

$$A = \begin{pmatrix} B & 0 \\ 0 & C \end{pmatrix} \Rightarrow q_1(A) = \begin{pmatrix} q_1(B) & 0 \\ 0 & q_1(C) \end{pmatrix}$$

$$\text{So, } \quad \text{ker } (q_1(A)) \quad \text{ker } (q_2(A))$$

$$A \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} Bx \\ Cy \end{pmatrix} = Bx + Cy$$

$\bullet q_1(B) : \text{ker } (p_1(A)) \mapsto \text{ker } (p_1(A))$

Now

$$q_1(A) \begin{pmatrix} x \\ y \end{pmatrix} = q_1(B)(x) + q_1(C)(y)$$

$$\text{But } x \in \text{ker } (q_1(A)) \Rightarrow q_1(B)(x) + q_1(C)(y) = q_1(A)(y)$$

$$\text{where } y=0 \Rightarrow \boxed{q_1(B)(x) = 0} + x \in \text{ker } (q_1(A))$$

S $q_1(B) = \emptyset : \ker(q_1(A)) \rightarrow \ker(q_1(A))$

B q_1 annihilates B

Same argument $\Rightarrow q_2$ annihilates C

Corollary

Under the hypothesis of the previous lemma

If q_1, q_2 are monic and $\mu_A = q_1 \cdot q_2$ then

$$q_1 = \mu_B, q_2 = \mu_C$$

"Proof" $\mu_A = \mu = \text{LCM}(\mu_B, \mu_C)$

$$\begin{array}{|c|c|} \hline & 6 \\ \hline 8 & 12 \\ \hline \end{array}$$

We know that μ_B divides q_1 since q_1 annihilates B

μ_C divides q_2 since q_2 annihilates C

But since q_1, q_2 relatively prime $\Rightarrow \mu_B, \mu_C$ relatively prime

$$\Rightarrow \text{LCM}(\mu_B, \mu_C) = \mu_B \cdot \mu_C \Leftrightarrow \mu = \mu_B \cdot \mu_C$$

But we also know $\mu_A = q_1 \cdot q_2 = \mu_B \cdot \mu_C$ and $\mu_B, \mu_C \mid q_2$ and
that q_1, q_2 are monic

D $q_1 = \mu_B$ and $q_2 = \mu_C$

H

Primary Decomposition Theorem

(distinct)

Suppose $A : \mathbb{V} \rightarrow \mathbb{V}$ and $\mu_A(z) = (z - z_1)^{p_1} \cdots (z - z_k)^{p_k}$, then

$$\mathbb{V} = \ker(A - z_1 I)^{p_1} \oplus \cdots \oplus \ker(A - z_k I)^{p_k}$$

and not this decomposition, A has the form —
where

$$\mu_{A_i} = (z - z_i)^{p_i}$$

$$\begin{bmatrix} A_1 \\ A_2 \\ \vdots \\ A_k \end{bmatrix}$$

"Proof" Induction k.

Base case " $k=1$ " holds true trivially

Inductive hypothesis

\hookrightarrow Hypothesis " $k=1, 2, \dots, m$ " holds

Show Case " $k=m+1$ " holds.

$$\text{Thus } \mu_A(z) = \underbrace{(z - \lambda_1)^{q_1} \cdots (z - \lambda_m)^{q_m}}_{q_1, q_2, \dots, q_m} \underbrace{(z - \lambda_{m+1})^{p_{m+1}}}_{w_1}$$

By the lemma \Rightarrow we have $A = \begin{pmatrix} B & C \\ 0 & C \end{pmatrix}$ wrt $V = \ker(q_1(A)) \oplus \ker(q_2(A))$

and $M_B = q_1, M_C = q_2$

w_1

By inductive hyp., $\ker(q_1(A)) = \ker(B - \lambda_1 I)^{p_1} \oplus \dots \oplus \ker(B - \lambda_m I)^{p_m}$

"new V"

new $V = \ker(q_1(A)) = W_1$

and

$$B = \begin{pmatrix} B_1 & & 0 \\ & \ddots & 0 \\ 0 & & B_m \end{pmatrix} \quad \text{wrt this decomposition, with } \mu_{B_i}(z) = (z - \lambda_i)^{p_i}$$

By assignment (?), A can be represented by

$$A = \begin{bmatrix} B_1 & & 0 \\ & \ddots & 0 \\ 0 & & B_m \\ \hline 0 & & C \end{bmatrix}$$

wrt the decomposition

$$V = \underbrace{\ker(B - \lambda_1 I)^{p_1}}_{\ker(q_1(A))} \oplus \dots \oplus \underbrace{\ker(B - \lambda_m I)^{p_m}}_{\ker(q_m(A))} \oplus \ker(A - \lambda_{m+1} I)^{p_{m+1}}$$

So, the next step is to show $\ker(B - \lambda_i I)^{p_i} = \ker(A - \lambda_i I)^{p_i}$

idea

Now, $\ker(A - \lambda_i I)^{p_i} \subset \ker((A - \lambda_1 I)^{p_1} \cdots (A - \lambda_m I)^{p_m})$ since

commute ...

such that I can write $(A - \lambda_i I)^{p_i} = (A - \lambda_m I)^{p_m} (A - \lambda_1 I)^{p_1}$

more

(1) $\Leftrightarrow \ker (A - \lambda_1 I)^{p_1} \subset \ker (\phi_1(A)) = W_1$

(2) similarly, $\ker (B - \lambda_j I)^{p_j} \subset \ker (\phi_j(A)) = W_j$

Recall

$$A = \begin{pmatrix} w_1 & w_2 \\ w_1 & B & 0 \\ w_2 & 0 & C \end{pmatrix}$$

$$\text{So } (A - \lambda_i I)^{p_i} = \begin{pmatrix} w_1 & & w_2 \\ & (B - \lambda_i I)^{p_i} & 0 \\ w_2 & 0 & -\text{Stuff-} \end{pmatrix}^{p_i} (A - \lambda_i I)^{p_i}$$

Now $x \in W_j$ is in $\ker (A - \lambda_j I)^{p_j} \Leftrightarrow \phi_j(x) = 0 \Leftrightarrow \boxed{\quad}(x) = 0$
 $\Leftrightarrow (B - \lambda_j I)^{p_j}(x) = 0 \Leftrightarrow x \in \ker ((B - \lambda_j I)^{p_j})$

$\Leftrightarrow x \in \ker (A - \lambda_j I)^{p_j} \Leftrightarrow x \in \ker ((B - \lambda_j I)^{p_j})$

$$\text{So } \boxed{\ker ((A - \lambda_i I)^{p_i}) = \ker ((B - \lambda_i I)^{p_i}) \text{ for } i=1 \dots m} //$$

∴

-H-

↳ what does this theorem say?

think if $(A - \lambda I)^p = 0$ and $N = A - \lambda I$, N is nilpotent

$\Leftrightarrow A = \lambda I + N$, N nilpotent

{Corollary}

→ Every $L: V \xrightarrow{\text{lin}} V$, finding can be expressed as

$$L = L_1 \oplus L_2 \oplus L_3 \oplus \dots \oplus L_m, \text{ where } L_i = \lambda_i I + W_i$$

distinct

nilpotent

NILPOTENTS

label

Structure of Nilpotents:

such that $N^8 = 0 \Rightarrow \mu_N = z^8$

Given a nilpotent $N: V \rightarrow V$ and $v_0 \in V$, we write

$$\langle v_0 \rangle = \text{span} \{v_0, N(v_0), N^2(v_0), \dots, N^7(v_0)\}$$

$$= \{ P(N)(v_0) \mid \deg(P) \leq 7 \} \subset \text{Lat}(N)$$

Theorem

For any N as above, there exist some $v_1, \dots, v_k \in V$
such that

$$V = \langle v_0 \rangle \oplus \langle v_1 \rangle \oplus \dots \oplus \langle v_k \rangle$$

→ we will look at proof later, but wrt this decomposition, N has the form

$$N = \begin{matrix} \langle v_0 \rangle & \langle v_1 \rangle & \dots & \langle v_k \rangle \\ \left(\begin{matrix} * & 0 & \dots & 0 \\ 0 & * & & \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & * \end{matrix} \right) \end{matrix} \rightarrow N \text{ is block-diagonal}$$

Suppose $v_0, N(v_0), \dots, N^5(v_0), N^6(v_0)$

lin. ind

$$\underline{\text{S}} \quad N^6(v_0) = \sum_0^5 a_i N^i(v_0)$$

$$\underline{\text{S}} \quad \underbrace{(N^6 - a^5 N^5 - \dots - a_0 I)}_{\text{S}}(v_0) = 0$$

$\underline{\text{S}} \quad g(N)(v_0) = 0 \quad \underline{\text{S}} \quad g(N) = N^{(v_0)}$ but deg at most
But local is some z^k , $k < 8$ $\quad \underline{\text{S}} \quad$ but deg at least 6

$$\underline{\text{S}} \quad N^6 = 0$$

So $(v_0, N(v_0), \dots, N^s(v_0))$ = basis of $\langle v_0 \rangle$

$$\begin{array}{c} \xrightarrow{\text{So}} \\ \begin{array}{ccccc} & v_0 & N(v_0) & \cdots & N^s(v_0) \\ \begin{matrix} v_0 \\ N(v_0) \\ \vdots \\ N^s(v_0) \end{matrix} & \left(\begin{array}{ccccc} 0 & 0 & & & 0 \\ 1 & 0 & & & 0 \\ 0 & 1 & & & 0 \\ 0 & 0 & & & 0 \\ 0 & 0 & & & 0 \end{array} \right) \end{array} \end{array}$$

Mar 21 2019

Recall Primary Decomposition Theorem

If $\alpha: V \rightarrow V$ has minimal polynomial $\mu_\alpha = (z - \lambda_1)^{p_1} (z - \lambda_2)^{p_2} \cdots (z - \lambda_k)^{p_k}$
 Then $V = \ker(\alpha - \lambda_1 I)^{p_1} \oplus \cdots \oplus \ker(\alpha - \lambda_k I)^{p_k}$ and wrt this
 decomposition α has the form $\alpha_1 \oplus \alpha_2 \oplus \cdots \oplus \alpha_k$ with
 $\alpha_i(z) = (z - \lambda_i)^{p_i}$ $\downarrow \alpha = \begin{pmatrix} \alpha_1 & 0 \\ 0 & \ddots \\ 0 & \alpha_k \end{pmatrix}$

① Next, observe that

$\mu_A(z) = (z - \alpha)^p$ then $A = \alpha I - N$ where $\mu_N(z) = z^p$
 we say that " N is a nilpotent of order p "

② if N is nilpotent of order p , $N: W \rightarrow W$, and $0 \neq w_0 \in W$
 then $\mu_{N, w_0}(z) = z^m$, for some $m \leq p$.

Consider $w_0, N(w_0), N^2(w_0), \dots, N^{m-1}(w_0), 0, 0, 0, \dots$

lin. ind (Since otherwise we get another μ_N)

Recall $\langle w_0 \rangle_N = \{ q(N)(w_0) \mid q \in P_m \} \leftarrow$ cyclic invariant subspace
 $\subseteq \text{Lat}(N)$

So $= \{ q(N)(w_0) \mid q \in P_{m-1} \}$ & this invariant ss of N has
 basis $w_0, \dots, N^{m-1}(w_0)$

Let Γ = basis set $w_0, N(w_0), \dots, N^{m-1}(w_0)$.

$$\text{Note: } N \begin{bmatrix} & \\ \langle w_0 \rangle & \end{bmatrix} \in \langle w_0 \rangle \mapsto \langle w_0 \rangle \text{ & } [N]_{\langle w_0 \rangle} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & & \\ 0 & 1 & \ddots & \\ 0 & & \ddots & 1 \\ \vdots & & & 0 \end{bmatrix}$$

Let $\eta: N^{m-1}(w_0) \dots w_0 \xrightarrow{\Gamma} w_0 \rightarrow$ write basis in reverse

$$\text{then } [N]_{\langle w_0 \rangle} \eta \in \Gamma = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & & & \\ \vdots & \vdots & 0 & & \\ 0 & & & \ddots & 1 \\ w_0 & 0 & 0 & \ddots & 0 \end{bmatrix}$$

Theorem

Cyclic Decomposition for Nilpotents

If $N: V \rightarrow V$ nilpotent then $V = \bigoplus_{i=1}^k \langle v_i \rangle_N$

f.d. f.d.

for some non-zero v_i

In fact, we can choose any v_i to start with.

(and wrt to this decomposition, N has the form $N_1 \oplus N_2 \oplus \dots \oplus N_k$)

Corollary

If $N: V \rightarrow V$ nilpotent then there is a basis of V

f.d. f.d.

of V such that $[N]_{\psi \in \Psi} =$

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

Ψ is made by
conjugation

Now \rightarrow "improve", look at

$$\begin{aligned} [\alpha I + N]_{\psi \in \Psi} &= \begin{bmatrix} \alpha & & & & \\ & \alpha & & & \\ & & \alpha & & \\ & & & \alpha & \\ & & & & \alpha \end{bmatrix} \quad \text{where each } \begin{bmatrix} \alpha & & & \\ & \alpha & & \\ & & \alpha & \\ & & & \alpha \end{bmatrix} \\ \alpha I \parallel & + [N]_{\psi \in \Psi} \end{aligned}$$

is called a "Jordan block"

J_α , size n \leftarrow denoted

$$\begin{bmatrix} \alpha & 0 & & \\ 0 & \alpha & & \\ \vdots & \vdots & \ddots & \\ 0 & 0 & \dots & \alpha \end{bmatrix} \rightarrow J_{\alpha, 1}$$

may or
may not
be there

Jordan canonical form theorem

unique up to changing order of appearance

Every matrix is similar to a direct sum of Jordan blocks

Conclude: $\{0\} \subset \ker(N) \subseteq \ker(N^2) \subseteq \ker(N^3) \subseteq \dots \subseteq \ker(N^8) \subseteq \ker(N)$

$$N = \begin{pmatrix} J_{0,0} & & & & & \\ & \vdots & & & & \\ & & J_{0,1} & & & \\ & & & \nearrow & & \\ & & & 8 & 7 & 7 & 7 & 6 \end{pmatrix}$$

↑
of blocks

$$\text{Grau } \{0\} \subseteq \ker(N) \subseteq \dots \subseteq V$$

if | ↑ | ↑ | ↑ | ↑ | ↑ | ↑

Diagram illustrating the mapping of indices to blocks:

- Index 10 maps to 10 blocks
- Index 10 maps to no $J_{0,1}$ blocks
- Index 8 maps to 2 $J_{0,2}$ blocks
- Index 6 maps to 2 $J_{0,3}$ blocks

∴ we can reconstruct $N = \begin{pmatrix} J & & \\ & \ddots & \\ & & J \end{pmatrix}$ by Weyr characteristic...

↳ So, no matter how decomposition is different, the final Jordan composition is determined by just N.

P Similar N, M nilpotents then they have the same Weyr characteristic
 \rightarrow they have the same Jordan form

But we also show that if N, M have the same Jordan-Bloch decomposition they are similar.

Weyr characteristic of N completely determines the Jordan block sizes and their multiplicities, i.e. completely determines the J-form of N (up to order of

\Rightarrow Similar matrices have the same
Weyr characteristics; so similar nilpotent have the same J-form
Conversely -

Conversely, if 2 nilpotents have the same J-form, then they are similar to it, and so to each other.

So

The J-form is a "complete invariant" for similarity of nilpotent.

Q Now, what abt $A = \alpha I + N$?

{ IF $A = \alpha I + N$ then the Weyr characteristic of $A - \alpha I$ }
 { determines the Jordan form of A. }

Q What if we don't know α ? What if all we know is
 $A = \text{something} - T + \text{nilpotent}$

\hookrightarrow use Trace! $\hookrightarrow \frac{\text{Tr}(A)}{\text{size of } A} \rightsquigarrow$ the trace doesn't change
 in different representations...

("Trace" is similarity-invariant)

Step 2

~~Similar (nilpotent)~~ share the same Jordan forms.

\rightarrow matrices \rightarrow (not just restricted to nilpotent)

\rightarrow

April 2, 2019

Diagonalizability

① Recall Every $n \times n$ matrix is triangulizable, i.e. is similar to an upper or lower triangular matrix.

D-ty

Equivently / Conversely $L \in \mathcal{L}(V)$ can be represented by S^T matrix
 for some basis

② Which matrices are diagonalizable?

Reverse engineer: If $A \sim D$, D diagonal, then, $\mu_A = \mu_D$.

and if $\mu_D = (z - d_1) \dots (z - d_k)$, μ_D has no repeated roots
 ↑ ↗
 distinct i.e. all roots have
 d_i multiplicity 1

► **Observe**

$$\text{Ex } J_{\lambda, 3} = \begin{vmatrix} \lambda & 10 \\ & 21 \\ & \lambda \end{vmatrix} \quad (z - \lambda) \mid \mu_J(z) \mid (z - \lambda)^3$$

KANN

$$(J - \lambda I)^3 = 0$$

$$J_{\lambda, 3} \Rightarrow \mu_J(z) = (z - \lambda)^3$$

$$\text{So } J_{\lambda, 3} \Rightarrow \mu_J(z) = (z - \lambda)^3$$

Now, if $A \sim J_1 \oplus J_2 \oplus \dots \oplus J_k$ then $\mu_A = \text{LCM}(\mu_{J_1}, \mu_{J_2}, \dots, \mu_{J_k})$
 So in particular, $\mu_{J_i} \mid \mu_A$ for each i .

So if μ_A has no repeated roots, then no μ_{J_i} has repeated roots

i.e. all J_i is 1×1

similar to

So $A \sim J_1 \oplus \dots \oplus J_k$ is a diagonal matrix

Theorem

TFAE

- ① A is diagonalizable
- ② μ_A has no repeated roots
- ③ There is a basis of \mathbb{C}^n made up entirely of eigenvectors of A
 (an A -eigenbasis of \mathbb{C}^n)

How to check if μ_A has no repeated roots? \rightarrow look at μ_A, μ'_A
 if μ_A has repeated roots, $\mu_A(\text{root}) = \mu'_A(\text{root}) = 0$

How to check if μ_A and μ_B have no common roots? \Rightarrow division algorithm
 \rightarrow find $\gcd(\dots)$ (or Euclidean algorithm)
 If $\gcd = 1$ then no common roots...

matrix $A \rightarrow$

$$\sigma_{\mathbb{C}}(p(A)) = p[\sigma_{\mathbb{C}}(A)]$$

Spectral Mapping Theorem

$$\hookrightarrow \text{Sup} \left\{ \sigma_{\mathbb{C}}(\alpha(A)) = \alpha[\sigma_{\mathbb{C}}(A)] \right\}$$

$$\sigma_{\mathbb{C}}(A + \beta I) = \sigma_{\mathbb{C}}(A) + \beta$$

Suppose A invertible. what is $\sigma_{\mathbb{C}}(A^{-1})$?

$$\hookrightarrow A^{-1}(v) = \gamma v, v \neq 0$$

$$\hookrightarrow \frac{1}{\gamma} v = Av \quad (\gamma \neq 0 \text{ since...})$$

Theorem

$$r \in \sigma_{\mathbb{C}}(A) \Leftrightarrow \frac{1}{r} \in \sigma_{\mathbb{C}}(A^{-1})$$

$$\text{SMT} \quad \sigma_{\mathbb{C}}(A^{-1}) = \frac{1}{\sigma_{\mathbb{C}}(A)} = \frac{1}{z} [\sigma_{\mathbb{C}}(A)]$$

holds for
rational fn...

What about simultaneous...? Commute...?

→

Simultaneous Diag-ability, Diag-ability

April 4, 2019

① Sim. Diag-ability

$\sqrt{\lambda, M} \rightarrow \sqrt{\quad}$ Suppose $A \xrightarrow{S^{-1}} \text{diagonal}$
 $B \xrightarrow{S^{-1}} \text{diagonal}$

$F^n \rightarrow C^n$ i.e.
 $S^{-1}AS \text{ diag}$
 $S^{-1}BS \text{ diag}$.

Then

$$S^{-1}ASS^{-1}BS = S^{-1}ABS \Leftrightarrow [AB = BA]$$

" "

$$S^{-1}BS * S^{-1}AS = S^{-1}BAS$$

Necessary conditions for Simultaneous diagonalizability

- (1) Individual diagonalizability
- (2) Commutativity

In fact, there are also sufficient.

(+) **Theorem** TFAE for a collection of $F \in M_n$

① F is a commuting collection and all elements $\in F$ are individually diagonalizable

② There is invertible S such that $S^{-1}AS$ diagonal for every $A \in F$

Pf We prove Lemma

A diagonalizable $\Leftrightarrow A = \alpha_1 E_1 + \dots + \alpha_k E_k$
 for some idempotents s.t. $E_1 + \dots + E_k = I$
 (non-zero)

(\Rightarrow 2) Pf A diag-able $\Leftrightarrow S^{-1}AS$ diag $= P_1 F_1 + \dots + P_m F_m$, $F_i \rightarrow \text{idemp} \neq 0$
 $\Leftrightarrow F_1 + \dots + F_m = I$

$$\begin{aligned}
 \text{So } A &= S \left[\beta_1 F_1 + \dots + \beta_m F_m \right] S^{-1} \\
 &= \underbrace{\beta_1 S F_1 S^{-1} + \dots + \beta_m S F_m S^{-1}}_{\text{idemp}} \\
 &= \beta_1 E_1 + \dots + \beta_m E_m
 \end{aligned}$$

$$\text{Now, } E_1 + \dots + E_m = S^{-1} (F_1 + \dots + F_m) S = I$$

$\textcircled{2} \Rightarrow \textcircled{1}$ Suppose $A = \alpha_1 E_1 + \dots + \alpha_k E_k$ where $E_1 + \dots + E_k = I$, E_i idemp.
then

$$\oplus \text{Im } E_i = V = \mathbb{C}^P \quad (\star)$$

Let P_i = a basis of $\text{Im } E_i$, $P = P_1 \sqcup P_2 \sqcup \dots \sqcup P_k$ a basis of \mathbb{C}^P

Then $[A]_{P \times P}$? Express A as a block matrix w.r.t $\textcircled{2}$

$$A_{ii} = E_i \circ A \circ E_i \Big|_{\text{Im}(E_i)} = \alpha_i E_i \Big|_{\text{Im}(E_i)} = \alpha_i I_{\text{Ran } E_i}$$

So

$$A = \begin{pmatrix} \alpha_1 I \\ & \ddots \\ & & \alpha_k I \end{pmatrix} \xrightarrow{\text{w.r.t } P} [A]_{P \times P} = \begin{pmatrix} \alpha_1 & & & \\ & \ddots & & \\ & & \ddots & \\ & & & \alpha_k \end{pmatrix}$$

So A diagonalizable. \square

(Corollary) (for the proof) The α 's are exactly the eigenvalues of A

and $\text{Im } E_i = E_i(\alpha_i) \rightsquigarrow$ eigenspace of A corresponding to α_i

and the rep' of A as

$A = \alpha_1 E_1 + \dots + \alpha_k E_k$ is unique

↑ distinct α_i

If $A = \begin{pmatrix} \alpha_1 & & & \\ & \ddots & & \\ & & \alpha_k & \\ & & & \mathbb{I}_n \end{pmatrix}$ wrt $\oplus \text{Im } \varepsilon_i = \mathbb{C}^n$

then $p(A) = \begin{bmatrix} p(\alpha_1)\mathbb{I} & & & \\ & \ddots & & \\ & & p(\alpha_k)\mathbb{I} & \\ & & & \mathbb{I}_n \end{bmatrix}$ α_i are distinct

Consider $E_i = \begin{bmatrix} \mathbb{I} & \alpha \\ 0 & \ddots \\ & & \mathbb{I} \\ & & & 0 \end{bmatrix}$ Now, $p(z) = \frac{(z-\alpha_1) \cdots (z-\alpha_k)}{(x_1 - x_2) \cdots (x_k - x_n)}$
Send A to E_i ,

Thm Each spectral idempotent of a diagonalizable A is a polynomial in A , i.e. it is of the form $p(A)$ for some poly. p .

Pf of Thm t] \rightarrow by induction ... Induct on n

- (1) \Rightarrow (2)
 - Base case : $n=1$ trivially true
 - Suppose holds for all $1 \leq n \leq n_0$. Show holds for $n_0 + 1$.

If F contains only scalar multiples of identity \Rightarrow done ...

So, let us assume F contains $A \neq \alpha \mathbb{I}$

then $A = \sum_{i=1}^k \alpha_i \varepsilon_i$ for some ... distinct α_i

Since distinct α_i , $k \geq 2$

Recall $\varepsilon_i = p_i(A)$, for some poly p_i
and every $B \in F$ commutes with A

So B commutes with $p_i(A)$, i.e. with every ε_i .

So $B = \begin{bmatrix} & & & \\ & \ddots & & \\ & & B_{11} & \\ & & & \mathbb{I}_{n-k} \end{bmatrix} = \dots$

8:

$$B_{ij} = \left| \begin{matrix} E_i \circ B \circ E_j \\ \text{Im}(E_j) \end{matrix} \right| = \left| \begin{matrix} E_i \circ E_j \circ B \\ \text{Im}(E_i) \end{matrix} \right|$$

$\hookrightarrow B$ block diagonal : $B = \begin{bmatrix} B_{11} & & \\ & \ddots & \\ & & B_{kk} \end{bmatrix}$

A, B commute \Rightarrow A_{ij} commutes w/ B_{ij}

\rightarrow $B \in B_{ij}$ individually diagonalizable ... (by hypothesis)

\hookrightarrow orthonormal basis $\Rightarrow B$ diagonalizable $\Rightarrow //$

April 9, 2019

INNER PRODUCTS



Lots and lots of ways to measure distance... Pythagorean theorem
 $\left\{ \begin{array}{l} \text{sum } |x|^2 + |y|^2 \\ \dots \end{array} \right.$

But

Pythagorean distance is associated with the dot product...

$$\text{dist} \left(\begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} a \\ b \end{pmatrix} \right)^2 = \left(x-a \right)^2 + \left(y-b \right)^2$$

Also, $\left\| \begin{pmatrix} x \\ y \end{pmatrix} \right\| \cdot \left\| \begin{pmatrix} a \\ b \end{pmatrix} \right\| \cdot \cos \theta = \begin{pmatrix} x \\ y \end{pmatrix} \cdot \begin{pmatrix} a \\ b \end{pmatrix}$ as there's a angle captured and distance,

\rightarrow makes Pythagorean distance standard...

Defn An inner product $\varphi: V \times W \rightarrow \mathbb{C}$ on a vector space V is a function satisfying the following conditions...

- ① φ is ~~linear~~ partially linear in the first slot ~ partially conjugate linear in the second
i.e.

$$\left\{ \begin{array}{l} \varphi(\alpha v_1 + \beta v_2, w) = \alpha \varphi(v_1, w) + \beta \varphi(v_2, w) \text{ and} \\ \varphi(v, \alpha w_1 + \beta w_2) = \bar{\alpha} \varphi(v, w_1) + \bar{\beta} \varphi(v, w_2) \end{array} \right.$$

Standard inner product on \mathbb{C}^n :

$$\left\langle \begin{pmatrix} a \\ b \\ c \end{pmatrix}, \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} \right\rangle = a\bar{\alpha} + b\bar{\beta} + c\bar{\gamma} = \dots$$

$$\Rightarrow \text{that } \left\langle \begin{pmatrix} a \\ b \\ c \end{pmatrix}, \begin{pmatrix} a \\ b \\ c \end{pmatrix} \right\rangle = |a|^2 + |b|^2 + |c|^2 = \left\| \begin{pmatrix} a \\ b \\ c \end{pmatrix} \right\|^2$$

$$(2) \quad \varphi(v, w) = \varphi(w, v)$$

$$(3) \quad \varphi(v, v) \geq 0, \text{ equality holds iff } v = 0 \text{ (positive definite)}$$

An inner product space is a vector space with an inner product.

Observe

If (V, φ) is an inner product space, then

$\|v\| := \sqrt{\varphi(v, v)}$ defines a "norm" (i.e. a sizing function)

such that

$$(i) \quad \|\cdot\| : V \rightarrow [0, \infty)$$

$$(ii) \quad \|v\| = 0 \Leftrightarrow v = 0$$

$$(iii) \quad \|\alpha v\| = |\alpha| \cdot \|v\|$$

$$(iv) \quad \|u+v\| \leq \|u\| + \|v\| \text{ (triangle inequality)}$$

$$(v) \quad \|\langle v, u \rangle\| \leq \|v\| \cdot \|u\|$$

$$\text{If f (iv)} \quad \|u+v\| = \dots = \sqrt{\|u\|^2 + 2\operatorname{Re} \varphi(u, v) + \|v\|^2}$$

$$\left(\leq \sqrt{\|u\|^2 + 2|\varphi(u, v)| + \|v\|^2} \right)$$

$$\leq \sqrt{(\|u\| + \|v\|)^2} = \|u\| + \|v\|$$

Q Can it be that we always have $|\varphi(u, w)| \leq \|u\| \cdot \|w\|$?

Yes! \rightarrow Cauchy-Schwarz inequality

In any product space (V, φ) $|\varphi(v, w)| \leq \sqrt{\varphi(v, v)} \cdot \sqrt{\varphi(w, w)}$

$$\hookrightarrow \alpha \in \varphi(tu + w, tw + w) = \varphi(tu, tw) + 2\operatorname{Re} \varphi(tu, w) + \varphi(w, w)$$

$$\text{for } t \in \mathbb{R} \quad = t^2 \varphi(u, u) + 2t \operatorname{Re} \varphi(u, w) + \varphi(w, w)$$

Hence no real roots except 0 $\Rightarrow \Delta < 0$

$$\hookrightarrow (2\operatorname{Re} \varphi(u, w))^2 - 4\varphi(u, u)\varphi(w, w) \leq 0 \quad \rightarrow \text{But we're not done...}$$

$$\hookrightarrow \operatorname{Re} \varphi(u, w) \leq \sqrt{\varphi(u, u)} \sqrt{\varphi(w, w)} \quad \square$$

know, for $|\alpha| = 1$,

$$\operatorname{Re}(\alpha \varphi(u, w)) \leq |\alpha| \sqrt{\varphi(u, u)} \sqrt{\varphi(w, w)}$$

$$\operatorname{Re}(\underbrace{\alpha \varphi(u, w)}_{r e^{i\theta}}) \leq \sqrt{\varphi(u, u)} \sqrt{\varphi(w, w)}$$

$r e^{i\theta}$

for $\alpha = e^{-i\theta}$, I get

$$\operatorname{Re}(r) \leq \sqrt{\varphi(u, u)} \sqrt{\varphi(w, w)}, \text{ but } r = |\varphi(u, w)| \text{ real}$$

$$\hookrightarrow \operatorname{Re}|\varphi(u, w)| \leq \sqrt{\varphi(u, u)} \sqrt{\varphi(w, w)}$$

$$\hookrightarrow |\varphi(u, w)| \leq \sqrt{\varphi(u, u)} \sqrt{\varphi(w, w)} \quad \square$$

► for $V = C[0, 1]$

we can def inner product of 2 fns $\varphi(f, g) = \int_0^1 f \cdot \bar{g}$

ORTHOGONALITY

Recall, Inner product spaces $(\mathcal{V}, \langle \cdot, \cdot \rangle)$

April 11, 2019

"Def" ORTHOGONAL $\Leftrightarrow \langle \cdot, \cdot \rangle = 0$

For any $S \subset \mathcal{V}$, let $S^\perp = \{v \in \mathcal{V} \mid \langle v, s \rangle = 0 \text{ for every } s \in S\}$

① Observe that $\theta_v \in S^\perp$

② $\mathcal{V}^\perp = \{v \in \mathcal{V} \mid \langle v, w \rangle = 0 \text{ for all } w \in \mathcal{V}\} \ni \theta_v$

Suppose $v_0 \in \mathcal{V}^\perp$, then $\langle v_0, v_0 \rangle = 0$, thus $v_0 = \theta$

③ $\cdot (\mathcal{V} \setminus \{\theta_v\})^\perp = \dots = \{\theta_v\}$

④ $\cdot (\mathcal{V} \setminus \{v_0\})^\perp = \dots = \{\theta_{v_0}\}$

⑤ $\cdot (\mathcal{V} \setminus \{v_0, v_1, \dots, v_n\})^\perp = \dots = \{\theta_{v_n}\}$

↑
countably many

⑥ For any $S \subset \mathcal{V}$: $S^\perp \not\propto \mathcal{V}$ pf $\{ \theta_v \in S^\perp$
 partial lin. in 1st slot of $\langle \cdot, v \rangle$

⑦ Thm If $W \not\propto \mathcal{V}$ then $W \oplus W^\perp = \mathcal{V}$ f.d. only.

* If we know $W + W^\perp = \mathcal{V}$ then the sum is direct, since
 $W \cap W^\perp = \{\theta\}$

* finite-dim diff w_1, w_2, \dots, w_n be a basis for W and let
 $v \in \mathcal{V}$

To show: $\exists \alpha_1, \dots, \alpha_k \in \mathbb{C}$ s.t. $\alpha_1 w_1 + \dots + \alpha_n w_n + y = v$
 for some $y \in W^\perp$

i.e. $\exists \alpha_1, \dots, \alpha_k \in \mathbb{C}$ such that $v - \alpha_1 w_1 - \dots - \alpha_n w_n \in W^\perp$

i.e. $\exists \alpha_1, \dots, \alpha_k \in \mathbb{C}$ s.t.

$$\langle v - \alpha_1 w_1 - \dots - \alpha_n w_n, w_1 \rangle = 0$$

$$\langle v - \alpha_1 w_1 - \dots - \alpha_n w_n, w_2 \rangle = 0$$

$$\text{So } \langle v - \alpha_1 w_1 - \dots - \alpha_k w_k, w_j \rangle = 0$$

$$\text{So } \langle v, w_j \rangle = \alpha_1 \langle w_1, w_j \rangle + \alpha_2 \langle w_2, w_j \rangle + \dots + \alpha_k \langle w_k, w_j \rangle$$

i.e.

$$\begin{bmatrix} \langle w_1, w_1 \rangle & \langle w_2, w_1 \rangle & \dots & \langle w_k, w_1 \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle w_1, w_k \rangle & \langle w_2, w_k \rangle & \dots & \langle w_k, w_k \rangle \end{bmatrix} \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_k \end{pmatrix} = \begin{pmatrix} \langle v, w_1 \rangle \\ \vdots \\ \langle v, w_k \rangle \end{pmatrix}$$

GRAMIAN
MATRIX

\rightsquigarrow we shall show it is invertible. But since it's square, we can just show it is injective.

Suppose $G(\gamma) = 0$ then

$$\sum_i \gamma_i \langle w_i, w_i \rangle = \sum_i \gamma_i \langle w_i, w_1 \rangle = \dots = \sum_i \gamma_i \langle w_i, w_k \rangle = 0$$

So $\sum_i \gamma_i w_i$ is $\perp w_1, \dots, w_k$

In particular, $\sum_i \gamma_i w_i \perp \sum_i \gamma_i w_i$, i.e. $\sum_i \gamma_i w_i = 0$

So $\gamma_i \cdot \gamma_i = 0 \forall i$ (since w_i form a basis)

So, the Gramian is injective, hence it is invertible. \checkmark



\Rightarrow Note that in infinite dim \rightarrow this theorem does not hold

Ex \hookrightarrow let $V = \mathbb{C}^{\mathbb{N}}: \{f: \mathbb{N} \rightarrow \mathbb{C} \mid f = \text{func}\}$ the space of complex numbers

$$= \boxed{\begin{array}{cccccc} 1 & 2 & 3 & \dots & & \\ x_1 & x_2 & x_3 & \dots & & \end{array}} \quad x_i \in \mathbb{C}$$

We claim that V is a vector space (this is easy to check)

Let $V_2 \subset V$ be defined by

$$(\alpha, \dots, \alpha, \dots) \in V_2 \Leftrightarrow \sum_{i=1}^{\infty} |\alpha_i|^2 < \infty$$

Define $\langle (\alpha, \dots, \alpha, \dots), (\beta, \dots, \beta, \dots) \rangle = \alpha_1 \bar{\beta}_1 + \alpha_2 \bar{\beta}_2 + \dots$

Now, we show that this sum doesn't diverge.

Consider $|\alpha_1 \bar{\beta}_1| + |\alpha_2 \bar{\beta}_2| + \dots$ i.e. $|\alpha_1| |\bar{\beta}_1| + |\alpha_2| |\bar{\beta}_2| + \dots$

Consider partial sums $|\alpha_1| |\bar{\beta}_1| + \dots + |\alpha_n| |\bar{\beta}_n|$

$$= \left\langle \begin{pmatrix} |\alpha_1| \\ \vdots \\ |\alpha_{34}| \end{pmatrix}, \begin{pmatrix} |\beta_1| \\ \vdots \\ |\beta_{34}| \end{pmatrix} \right\rangle \leq \sqrt{\left\| \begin{pmatrix} |\alpha_1|^2 \\ \vdots \\ |\alpha_{34}|^2 \end{pmatrix} \right\| \cdot \left\| \begin{pmatrix} |\beta_1|^2 \\ \vdots \\ |\beta_{34}|^2 \end{pmatrix} \right\|}$$

But $\sum |\alpha_i|^2 < \infty$, so both A, B

\rightarrow any inner product converges...

Therefore the inner product is well defined.

q.e.d.

Back to proof. Let $W \subset \ell^2$ be defined by

W the set of all terminating sequences in ℓ^2

So (we can check) $W \subset \ell^2$, and $\begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \in W$, and

$$\begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \in W, \quad \begin{pmatrix} 0 \\ 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} \in W \dots$$

$$W + W^\perp = W$$

So $\begin{pmatrix} \alpha_j \\ \vdots \\ \alpha_1 \end{pmatrix} \in W^\perp \Rightarrow \alpha_j = 0 \forall j \Rightarrow W^\perp = \{0\}$

Yet $\begin{pmatrix} 1/2 \\ 1/4 \\ \vdots \end{pmatrix} \in \ell^2, \notin W \subseteq [W + W^\perp = \ell^2]$

(read Axler)

April 16, 2018

Recall, $W \subset V$, V finite dimensional inner product space
 $W \oplus W^\perp = V$

Observation

Orthonormal list in inner product space is linearly independent.

Suppose $\alpha_1 v_1 + \dots + \alpha_k v_k = 0$

where v_1, \dots, v_k mutually orthogonal unit vectors

Then $0 = \langle 0, v_i \rangle$

$$= \alpha_i \langle v_i, v_i \rangle + 0 = \alpha_i$$

So all v_i 's are linearly independent.

Thm

Every finite dimensional inner product space (non-trivial)
 has an orthonormal basis, and in fact any ortho-
 normal list can be enlarged to one.

Riesz Representation Theorem

a linear functional on V

{ Suppose V is f.d. and $\rho : V \xrightarrow{\text{lin}} \mathbb{C}$.
 Then there exists w_0 s.t. exactly one $w \in V$ s.t.

$$\rho(v) = \langle v, w_0 \rangle, \forall v, \text{ i.e. } \rho(\cdot) = \langle (\cdot), w_0 \rangle$$

Proof $\text{rank}(\rho) + \text{Nullity } \rho = \dim(V)$

0 or 1

If $\text{rank}(\rho) = 0$, then $\rho \equiv 0$, so $w_0 = 0$ (unique) (easy)

If $\text{rank}(\rho) = 1$

$$\text{then } \ker(\rho \oplus \ker(\rho))^\perp = V$$

then $\ker(p)^\perp$ has dimension one. Thus $\ker(p)^\perp = \text{span}(z_0, z_1)$, unit vector.

$$\text{Then } p(z_0) \neq 0 \Rightarrow \frac{1}{p(z_0)} z_0 \in \text{span}(z_1)$$

$$\text{and } g\left(\frac{1}{p(z_0)} z_0\right) = \underbrace{\frac{1}{p(z_0)}}_{z_1} p(z_1) = 1$$

$$\text{And so, } \text{span}(z_0) = \text{span}(z_1)$$

Thus, any $v \in V$, $v = w + \alpha z_1$ (uniquely)

$$\begin{array}{ccc} \cap & \cap \\ \ker(p) & \ker(p)^\perp \\ \xrightarrow{\quad} \\ \text{So } p(v) = p(w + \alpha z_1) = 0 + \alpha p(z_1) = \alpha \stackrel{?}{=} \langle v, ?? \rangle \end{array}$$

So now, $\langle w + \alpha z_1, ?? \rangle$ how to get α ?

$$= \langle w + \alpha z_1, ?? \rangle$$

$$= \langle w, ?? \rangle + \alpha \langle z_1, ?? \rangle$$

$$0 \quad ? \quad ? = \frac{1}{\|z_1\|^2}$$

$$\text{And thus, } ?? = \frac{1}{\|z_1\|^2} z_1$$

$$\text{So, let } w_0 = \frac{z_1}{\|z_1\|^2}$$

$$\text{then } g(v) = g(w + \alpha z_1) = \left\langle v, \frac{z_1}{\|z_1\|^2} \right\rangle = (\alpha) = \langle v, w_0 \rangle$$

for all v . //

$$\text{Reap, if } v = w + \alpha z_1, \text{ then } w_0 = \frac{z_1}{\|z_1\|^2} + v$$

Uniqueness Assume $\beta = \langle (\cdot), w_0 \rangle = \langle (\cdot), w_1 \rangle$ then $0 = \langle (\cdot), w_0 - w_1 \rangle \Leftrightarrow w_0 = w_1$ //

Consequences?

① Suppose $L: V \xrightarrow[\text{lin}]{\text{dip}} V$. Consider

$f(): \langle L(), y_0 \rangle, f: V \xrightarrow{\text{lin}} \mathbb{C}$, so, f must be like

$f() = \langle (), w_0 \rangle$ for some $w_0 \in V$, w_0 depends on UNIQUE
 y_0 and L .

So let us write $w_0 = w_L(y_0)$

② Consider $w_L: V \xrightarrow{\text{lin}} V$

$$\text{Now } f() = \langle 0, w_0 \rangle = \langle L(), y_0 \rangle = \bar{x} \langle x, w_L(y_0) \rangle$$

$$\text{So } \langle x, w_L(y_0) \rangle = \langle Lx, y_0 \rangle = \bar{x} \langle Lx, y_0 \rangle = \text{soft bold}$$

$$\therefore \boxed{w_L(\alpha y_0) = \bar{\alpha} w_L(y_0)} \quad (\text{scalar linear})$$

• what about additivity?

$$w_L(y_1 + y_2) = \langle L(x), y_1 + y_2 \rangle = \langle x, w_L(y_1) \rangle + \langle x, w_L(y_2) \rangle$$

thus, we can show $w_L(y_1 + y_2) = w_L(y_1) + w_L(y_2)$
 by subtracting ...

$$\left. \begin{array}{l} \langle x, y \rangle - \langle x, z \rangle = 0 \\ \langle x, y-z \rangle = 0 + x \\ \Rightarrow y-z = 0 \end{array} \right\}$$

And so, additivity checks.

$w_L: V \xrightarrow{\text{lin}} V$ has the property

$$\langle L(), y \rangle = \langle (), w_L(y) \rangle$$

w_L is the adjoint of L , denoted L^*

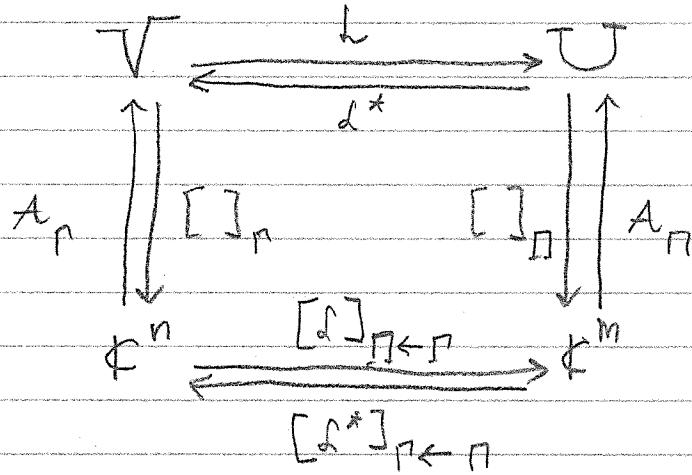
Note L depends on the inner product, but not on the basis. □



Let Γ be orthonormal basis of V v_1, \dots, v_n

Π be orthonormal basis of U u_1, \dots, u_m

then



What is the relationship between $[f]$ and $[f^*]$? Claim:

$$[f^*]_{\Gamma \leftarrow \Pi} = ([f]_{\Pi \leftarrow \Gamma})^\top$$

Let write this $A = (\bar{B})^\top$

do this
entry-wise
...

$$A_{ij} = A[i;j] = A(e_j) \cdot e_i = A(e_j) \circ e_i$$



Observation $\alpha_1 V_1 + \dots + \alpha_n V_n$, V_i orthonormal

then $\langle \alpha_1 V_1 + \dots + \alpha_n V_n, \beta_1 V_1 + \dots + \beta_n V_n \rangle$

$$= \alpha_1 \bar{\beta}_1 + \dots + \alpha_n \bar{\beta}_n = \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} \circ \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_n \end{pmatrix} = \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} \bar{\circ} \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_n \end{pmatrix}$$

standard inner product
↳ like dot product!

April 18, 2019

Def: $L: \mathcal{V} \xrightarrow{\text{flip}} \mathcal{V}$ f.d.p.s is normal if $LL^* = L^*L$; $L^* \leftrightarrow L$

is self-adjoint if $L = L^*$; $L^* = L$

is unitary if $LL^* = L^*L = I$ i.e. $L^* = L^{-1}$

NORMAL
OPERATOR

SELF-ADJOINT
UNITARY

(Peak into Future) (1) L is normal \Rightarrow there is an ortho-basis Γ of \mathcal{V} such that

(1)

SPECTRAL

THEOREM

$[L]_{\Gamma \times \Gamma}$ is diagonal

\Rightarrow there is an ortho-eigenbasis of L for \mathcal{V} (eigenvect.)

(2)

$L = L^* \Rightarrow L$ is normal $\Rightarrow \tilde{\sigma}_\phi(L) \subset \mathbb{R}$

(3)

$L^* = L^{-1} \Rightarrow L$ is normal $\Rightarrow \tilde{\sigma}_\ell(L) \subset \mathbb{T}^1 = \{e^{i\theta} | \theta \in \mathbb{R}\}$

(all eigenvalues have modulus 1)

(*)

Schur's Thm

(For each $L: \mathcal{V} \xrightarrow{\text{flip}} \mathcal{V}$ f.d.p.s there is an ortho-basis Γ of \mathcal{V} s.t. $[L]_{\Gamma \times \Gamma}$ is upper-triangular)

1-dim ✓

1-dim ✓

dim = 1

Let $W = \text{span}(v)$ \rightsquigarrow 1 dimensional

then $V = W \oplus W^\perp$. Let v_0 be a unit vect in W^\perp

Then $\Gamma = (v_0, v_1)$ is an ortho-basis of \mathcal{V}

thus

$$[L]_{\Gamma \times \Gamma} = \begin{pmatrix} v_0 & \\ ? & M \end{pmatrix} \rightsquigarrow \text{upper triangular}$$

3-dim

$$[L]_{\Gamma \times \Gamma} = \left[\begin{array}{c|c} ? & ? \\ \hline 0 & M \\ 0 & \end{array} \right] \rightsquigarrow M = E_{w^\perp} \circ L \mid_{W^\perp}$$

By 2-dim, there is an basis v_1, v_2 of w^\perp such that

$$[N]_{\Gamma \times \Gamma} = \left[\begin{array}{c|c} ? & ? \\ \hline 0 & ? \end{array} \right] \text{ortho}$$

Thus, let $\Gamma = (v_0, v_1, v_2)$. Then Γ = ortho.basis of \mathcal{V}

$$\text{Then } [L]_{\Gamma \times \Gamma} = \begin{bmatrix} ? & ? & ? \\ 0 & [M] \end{bmatrix} = \begin{bmatrix} ? & ? & ? \\ 0 & ? & ? \\ 0 & 0 & ? \end{bmatrix} \rightsquigarrow \text{upper triangular}$$

Towards the Spectral Theorem

$$\begin{array}{|c|c|} \hline A^* & C^* \\ \hline B^* & D^* \\ \hline \end{array} \quad \text{II}$$

① Suppose $M \in M_{n \times n}$, and $M = \begin{pmatrix} A & B \\ C & D \end{pmatrix}_{n \times n}$ Then $M^* = (\bar{M})^T = \begin{pmatrix} \bar{A} & \bar{B} \\ \bar{C} & \bar{D} \end{pmatrix}^T = \begin{pmatrix} \bar{A}^T & \bar{C}^T \\ \bar{B}^T & \bar{D}^T \end{pmatrix}$

② Suppose M is normal, then $M \leftrightarrow M^*$ then

$$MM^* = \begin{pmatrix} A & B \\ C & D \end{pmatrix} \begin{pmatrix} A^* & C^* \\ B^* & D^* \end{pmatrix} = \begin{pmatrix} AA^* + BB^* & \cdots \\ \cdots & \cdots \end{pmatrix}$$

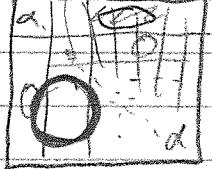
$$M^*M = \begin{pmatrix} A^* & C^* \\ B^* & D^* \end{pmatrix} \begin{pmatrix} A & B \\ C & D \end{pmatrix} = \begin{pmatrix} A^*A + C^*C & \cdots \\ \cdots & \cdots \end{pmatrix}$$

$$\therefore AA^* + BB^* = A^*A + C^*C$$

$$\therefore \text{Trace}(AA^* + BB^*) = \text{Trace}(A^*A + C^*C)$$

$$\therefore \text{Trace}(AA^*) + \text{Trace}(BB^*) = \text{Trace}(A^*A) + \text{Trace}(C^*C)$$

Note $\text{Trace}(AA^*) = 0 \Leftrightarrow A = 0$ and $\text{Tr}(BB^*) = \sum |B_{ij}|^2$

③ Suppose  is normal, then  is diagonal.
 Every normal has a representation that is diagonal.

④ Converse?

—

SPECTRAL THEOREM

* Equivalent statements:

① A is normal then there is an ortho-basis of the underlying space made up entirely of the eigenvectors of A

i.e. ② $\lambda: \sqrt{\frac{\lambda}{n}} \xrightarrow{\text{norm}} \sqrt{\frac{\lambda}{n}}$, then \exists ortho basis s.t. $[A]_{\mathbb{R}^n}$ is diagonal

i.e. ③ $E_{\lambda_1, j_1} \oplus E_{\lambda_2, j_2} \oplus \dots \oplus E_{\lambda_k, j_k} = \mathbb{V} \rightarrow$ direct sum +
 \uparrow direct sum + $\left\{ \overbrace{E_{\lambda_1, j_1} = \bigoplus_2^k E_{\lambda_i, j_i}}^{\perp} \right\}$ ortho normal

If A normal

i.e. (4) $A = \lambda_1 F_1 + \dots + \lambda_n F_n$ → idempotent \rightarrow also self adjoint

↑
ortho-
resolution of identity

$A \sim \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$

where $F_i^2 = F_i = F_i^*$ and $\sigma(F_i) \subset \{0, 1\}$

⇒ $\{F_i \text{ is a non-negative self-adjoint}\}$ ↑ properties...
 or positive-semidefinite

Terminology

Non-negative self-adjoint idempotents are called ORTHO-PROJECTION

So diagonalizable \Rightarrow resolution of identity = lin. comb. of idempotents

normal \Rightarrow resolution of identity = lin. comb. of idempotents
 that are ORTHO-PROJECTIONS.

Result for idempotent. $V = \text{Im}(E) \oplus \text{ker}(E)$

For ORTHO-PROJECTIONS $\Rightarrow \text{Im}(E) \perp \text{ker}(E)$



SPECTRAL THEOREM FOR MATRICES

$\Gamma = (f_1, \dots, f_n) = \text{orthonormal f.s}$

$$\begin{matrix} \mathbb{C}^n & \xrightarrow{\text{normal}} & \mathbb{C}^n \\ \Gamma & \uparrow \text{normal} & \downarrow [\Gamma]_\rho \\ \Gamma & \xrightarrow{\text{ortho}} & \mathbb{C}^n \end{matrix}$$

$$A_\rho = [f_1, f_2, \dots, f_n] : \mathbb{C}^n \xrightarrow{\text{bij}} \mathbb{C}^n$$

Square matrix

bijective

orthonormal columns

Unitary

Unitary Matrix

$$U^\dagger U = I \Leftrightarrow \begin{bmatrix} \bar{c}_1 \\ -\bar{c}_2 \\ \vdots \\ c_n \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{pmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{pmatrix}$$

$\Leftrightarrow \langle c_1, c_1 \rangle = 1 ; \langle c_2, c_1 \rangle = 0 \Leftrightarrow U \text{ normal} \Leftrightarrow \overline{U} \in \{e^{i\theta} | \theta \in \mathbb{R}\}$

Unitary Matrix, cont

$$\text{B} \quad U^*U = I \Leftrightarrow \langle U^*Ux, y \rangle = \langle Ix, y \rangle = \langle x, y \rangle \quad \forall x, y$$

$$(\text{proof: } \langle Ax, y \rangle = 0 \quad \forall x, y \Leftrightarrow A = 0)$$

$$\Leftrightarrow \langle U^*Ux, x \rangle = \langle x, x \rangle \quad \forall x$$

$$(\text{proof: } \& \text{ Math } \|Ux\|^2 = \|x\|^2)$$

$$\text{then note that } \langle x, y \rangle = \frac{1}{4} \left[\|x+y\|^2 - \|x-y\|^2 + i\|x+iy\|^2 - i\|x-iy\|^2 \right]$$

then $\langle Ux, Uy \rangle = \dots$ then linearity, then preserve by th

then

$$\langle x, y \rangle = \langle Ux, Uy \rangle \quad \forall x, y$$

$$\text{Now } \langle Ux, Uy \rangle = \langle x, y \rangle \Leftrightarrow \|U(x)\| = \|x\| \Leftrightarrow \langle Ux, Ux \rangle = \langle x, x \rangle$$

$$\langle U^*Ux, y \rangle = \langle x, y \rangle \Leftrightarrow U^*U = I$$

↑

Orthonormal \Leftrightarrow columns form an orthonormal basis \Leftrightarrow Unitary

$$\Rightarrow \det(U) = 1$$

\uparrow \uparrow

U^* unitary \Leftrightarrow rows of U form an orthonormal basis

$$\text{Back to } A_P = \begin{bmatrix} f_1 & \dots & f_n \end{bmatrix} : \mathbb{C}^n \xrightarrow{\text{bijective}} \mathbb{C}^n \quad \overset{*}{A_P}$$

orthonormal $\Rightarrow A_P$ unitary, so $[A_P]$ unitary

$$\text{Op}_A(M) \xrightarrow{\text{unitary}} [M]_{\text{per}}$$

$[M]_{\text{per}}$

$$U^* \circ M \circ U = [M]_{\text{per}}$$

unitary
diagonal

\hookrightarrow spectral theorem for matrices

comes from P

Diagonalizable, $S^{-1}AS = B$
such that S is unitary

\hookrightarrow M is normal, then M is unitarily equivalent to diagonal

\exists What if $W^* M W =$ diagonal, W unitary?

Does this mean M is normal? \rightarrow Yes

$M = W \circ D \circ W^*$, then $M^* = W \circ D^* \circ W^*$ \rightarrow diagonal

$$\text{So } M^* M = W D^* W^* W D W^* = W (D^* D) W^*$$

$$M M^* = W D W^* W D^* W^* = W (D D^*) W^*$$

$$\text{So } M^* M = M M^*$$

Spectral theorem for matrices

$[M] \in M_{n \times n}$ normal \Leftrightarrow it is unitarily equivalent to diagonal

\exists Schur's Theorem - extension

Every $n \times n$ matrix is unitarily similar to a triangular

Back to normal matrices

A normal, then $A = \gamma_1 E_1 + \dots + \gamma_k E_k$

\uparrow \uparrow

ortho-resolution

of identity

Q: What else do we get from Spectral Theorem?

A normal, then $A = UDU^*$ \rightsquigarrow unitary
 \downarrow
 $[d_1 \dots d_n]$

then $A^2 = U^* D^2 U \dots$

$\hookrightarrow A^n = U^* D^n U \dots \rightsquigarrow$ polynomial.

In particular, $P(A) = U^* P(D) U$

Defn $f(A) := U^* [f(d_1) \dots f(d_n)] U \rightsquigarrow$ whenever this makes sense ..

Then But what if we have different U 's?

Better definition

$$f(A) := f(\gamma_1) E_1 + \dots + f(\gamma_k) E_k$$

\uparrow this definition removes ambiguity

$$E \sim \begin{pmatrix} 1 & & \\ & \ddots & \\ & & 0 \end{pmatrix}$$

This becomes a theorem.

Then

$$f(A) = U^* \begin{bmatrix} f(d_1) & & \\ & \ddots & \\ & & f(d_k) \end{bmatrix} U$$

Q Suppose A is normal $\Leftrightarrow \sigma_A \subset [0, \infty) \subset \mathbb{R}$

$$\text{Then } \sqrt{A} \cdot \sqrt{A} = (\sum \sqrt{\lambda_i} E_i) \cdot (\sum \sqrt{\lambda_j} E_j)$$

$$= \lambda_1 E_1 + \lambda_2 E_2 + \dots + \lambda_n E_n = A$$

S if A is normal $\Leftrightarrow \sigma_A \subset [0, \infty) \subset \mathbb{R}$

then

$$\sqrt{A} \circ \sqrt{A} = A, \text{ where } \sqrt{A} = f(\lambda_1) E_1 + \dots + f(\lambda_k) E_k$$

and so \sqrt{A} is also normal, with non-negative spectrum
and so ...

E

Every non-negative normal matrix A has a nonnegative
normal square root \sqrt{A}

In fact

A normal + σ_A real $\Rightarrow A$ self adjoint

$$\hookrightarrow \text{pf } A = \lambda_1 E_1 + \dots + \lambda_n E_n$$

$$\left\{ \begin{array}{l} A^* = \overline{\lambda}_1 E_1^* + \dots + \overline{\lambda}_n E_n^* \\ A^* = \lambda_1 E_1 + \dots + \lambda_n E_n < A \text{ so } A \text{ self adj} \end{array} \right.$$

$$A^* = \overline{\lambda}_1 E_1^* + \dots + \overline{\lambda}_n E_n^*$$

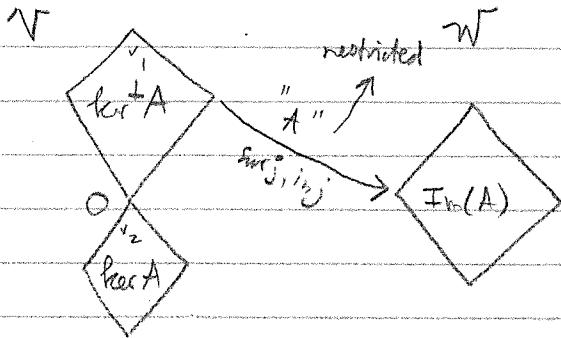
$$A^* = \lambda_1 E_1 + \dots + \lambda_n E_n < A \text{ so } A \text{ self adj}$$

"normal matrices are like numbers... we can apply functions to them... they're like a number system... not really, but they're quite nice"

Spatial of Adjoint

May 7, 2019

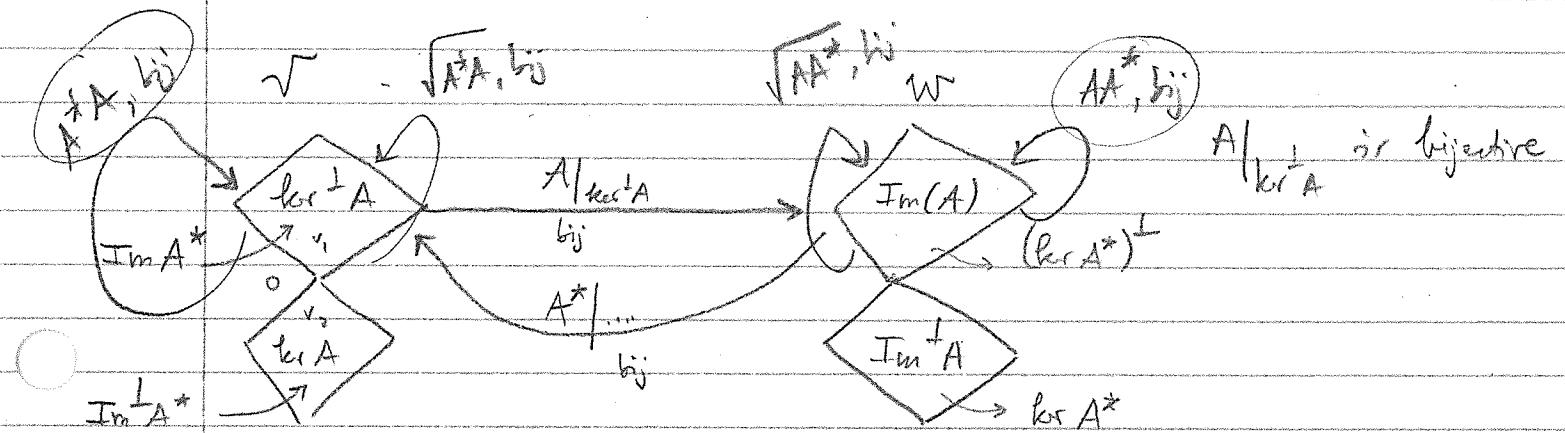
$$A: V \rightarrow W, V = \ker A \oplus \ker A^\perp$$



But if $Av_1 = 0$, then $v_1 \in \ker A$
But $v_1 \in \ker A^\perp$

$$\therefore v = 0$$

\Rightarrow "A" is a bijection



$$\text{But } \ker(A^*) = (\text{Im } A)^{\perp}$$

$$(\ker(A))^{\perp} = (\ker(A^*)^*)^{\perp} = \text{Im}(A^*)$$

$$\text{Now } \text{Im}(A^*A) \stackrel{?}{=} \text{Im}(A^*) = \ker A^\perp$$

$$\text{It's true that } \text{Im}(A^*A) \subset \text{Im}(A^*) = \ker A^\perp$$

$$\text{and } \ker(A^*A) \supset \ker A$$

$$\text{But notice } A^*A(v_1 + v_2) = A^*Av_1 = 0 \Leftrightarrow v_1 = 0$$

$$\therefore \ker(A^*A) = \ker A$$

$$\text{Alternatively, } \langle A^*Ax, x \rangle = \langle Ax, Ax \rangle = \|Ax\|^2$$

$$\text{if } A^*Ax = 0, \text{ then } \|Ax\|^2 = 0, \text{ then } Ax = 0$$

$$\text{and if } Ax = 0 \rightarrow A^*Ax = 0 \rightarrow \langle , \rangle = 0$$

$\boxed{\text{So } Ax = 0 \Leftrightarrow A^*Ax = 0, \text{ ie. } \ker(A) = \ker(A^*A)}$

$\boxed{\text{So now } \ker(A^*A) = \ker(A)}$

$$\hookrightarrow (\ker(A^*A))^\perp = (\ker(A))^\perp$$

$$\hookrightarrow \text{Im}((A^*A)^*) = \text{Im}A^* \quad \text{But } A^*A \text{ non negative}$$

$\boxed{\text{Im}(A^*A) = \text{Im}(A^*) = \ker(A)}$

$\boxed{\text{Now,}}$

$$x \in \ker(A^*) \Leftrightarrow A^*x = 0 \Leftrightarrow \|A^*x\|^2 = 0$$

$$\Rightarrow \langle A^*x, A^*x \rangle = 0 \Leftrightarrow \langle AA^*x, x \rangle = 0$$

$$x \in \ker(A) \Leftrightarrow \dots \Rightarrow \langle A^*Ax, x \rangle = 0$$

So, if A normal $\Rightarrow \ker(A) = \ker(A^*)$

$\boxed{\text{So } A \text{ normal} \Rightarrow \ker(A) = (\text{Im } A)^\perp}$

or $\boxed{A \Leftrightarrow A^* \Rightarrow \ker(A) = (\text{Im } A)^\perp}$

Now since A^*A nonnegative, there is exactly one $\sqrt{A^*A}$ nonnegative

Observe $x \in \ker \sqrt{A^*A} \Leftrightarrow \|\sqrt{A^*A}x\|^2 = 0$

$$\Rightarrow \langle \sqrt{A^*A}x, \sqrt{A^*A}x \rangle = 0 \quad \text{But } \sqrt{A^*A} \text{ self adj}$$

$$\Rightarrow \langle (\sqrt{A^*A})^*(\sqrt{A^*A})x, x \rangle = 0 \Leftrightarrow \langle A^*Ax, x \rangle = 0 \Leftrightarrow \boxed{\|Ax\|^2 = 0}$$

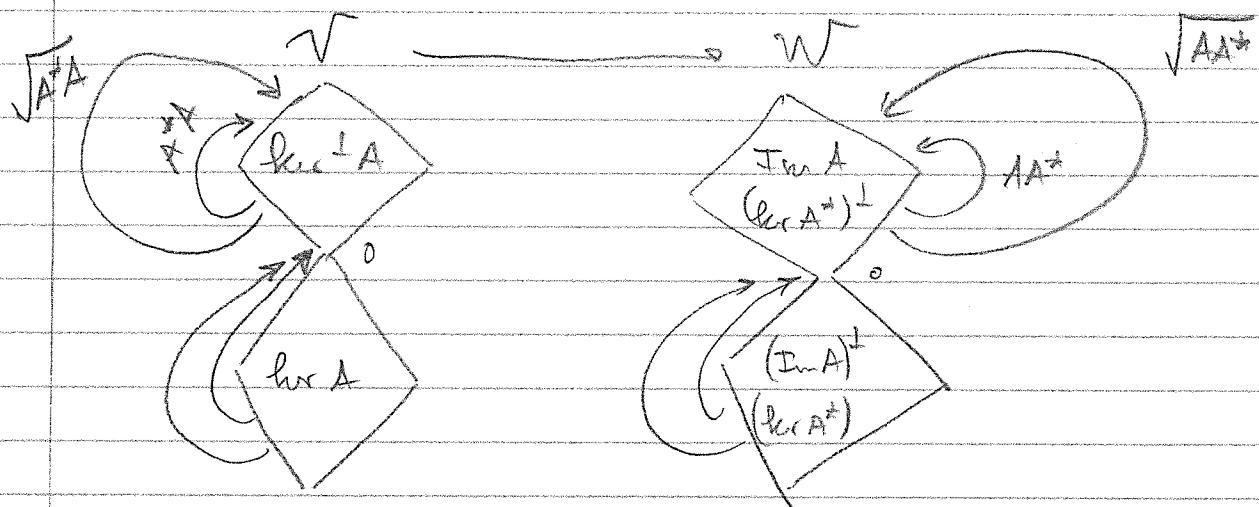
$$\boxed{\text{P} \quad \ker(\sqrt{A^*A}) = \ker(A)}$$

Now

$$\begin{aligned}\text{Im}(\sqrt{A^*A}) &= (\ker(\sqrt{A^*A})^*)^\perp \\ &= (\ker(\sqrt{A^*A}))^\perp \\ &= (\ker(A))^\perp\end{aligned}$$

$$\boxed{\text{Im } \sqrt{A^*A} = \text{Im}(A^*)}$$

Before, again



Repeat argument

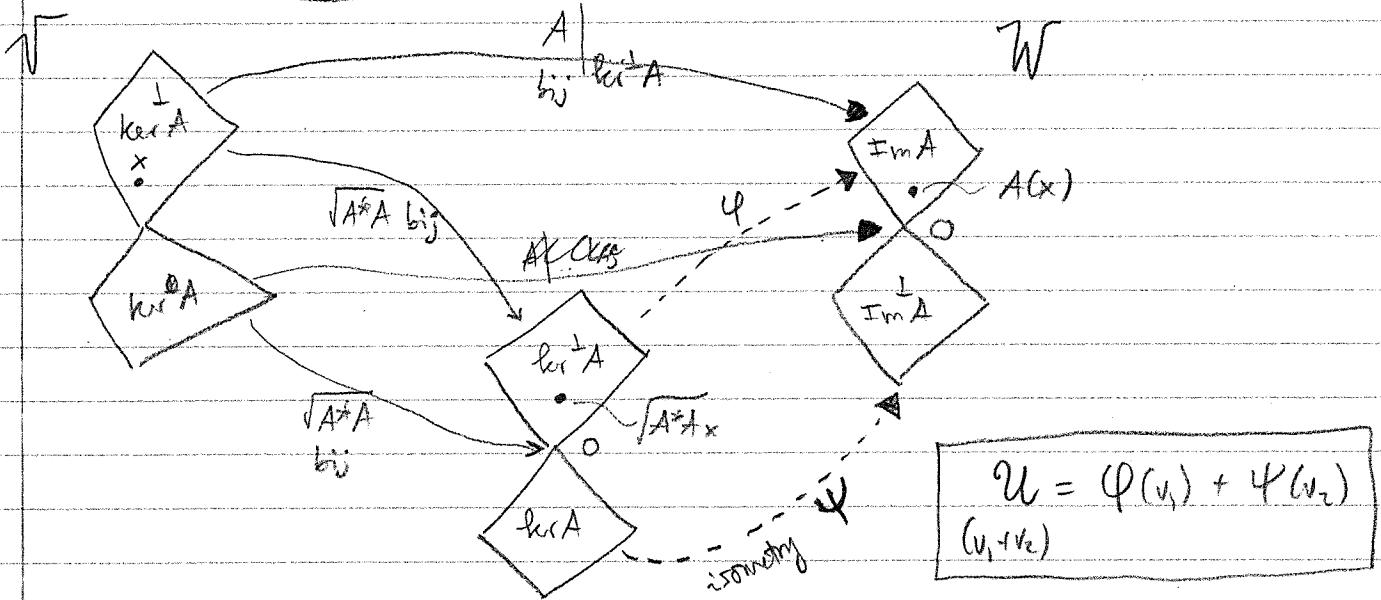
Theorem

$$\|\sqrt{A^*A}x\| = \|Ax\|$$

\downarrow

$\epsilon V \quad \epsilon W$

Another picture



$$\text{Now, } \|A^T A x\| = \|Ax\|$$

A small circular logo or seal containing a stylized five-pointed star.

Define new function

$$\boxed{4} \quad q: \ker A \longrightarrow \text{Im } A$$

۱۰

$$f: \ker^{\perp} A \longrightarrow \text{Im } A$$

as $\psi \left[\sqrt{A^* A}(x) \right] := A(x)$

$\lim_{i \rightarrow j}$ $\lim_{i \rightarrow j}$

Claim 1 φ is linear... because $\varphi = A_0 \sqrt{A^* A}$

Claim 2 φ is an isometry (length preserving)

Since depth preserving...

If W is V then $\dim(\ker A) = \dim(\text{Ran } A)$

then, we can define an isometry for $A \mapsto \text{Im}^{-1}A$

$$\text{So } z = \sum \alpha_i u_i \rightarrow \|z\|^2 = \sum |\alpha_i|^2$$

$$f(z) = \sum \alpha_i f(u_i) = \sum \alpha_i w_i \rightarrow \|f(z)\|^2 = \sum |\alpha_i|^2 \|w_i\|^2 = \|z\|^2$$

\hookrightarrow Isometry!!

This means we can extend Φ to $\ker A \mapsto \text{Im}^\perp(A)$

* By rank-nullity, $\dim(\ker(A)) = \dim(\text{Im}^\perp(A)) = \dim V - \dim(\text{Im} A)$

Pick any isometry Ψ from $\ker A$ to $\text{Im}^\perp A$
(linear)

and define

$\in \ker A$

$$\boxed{U : V \mapsto V \text{ by } U(v_1 + v_2) = \Psi(v_1) + \Psi(v_2)}$$

\downarrow

$\in \ker^\perp A$

Now

A

$$x \longrightarrow Ax$$

Let $x \in \ker^\perp A$

$$x \xrightarrow{\sqrt{A^*A}} \sqrt{A^*A} x \xrightarrow{\Psi} Ax$$

Let $a \in \ker A$

$$x \xrightarrow{A} 0$$

$$x \xrightarrow{\sqrt{A^*A}} 0 \xrightarrow{\Psi} 0$$

$\nearrow \text{unitary}$

\hookrightarrow we have just shown that $\boxed{A = U \circ \sqrt{A^*A}}$

This U is an isometry, surj, $\Rightarrow U$ is unitary

What if start at adjoint?

$$A^* = \tilde{U} \circ \sqrt{AA^*}$$

$$\hookrightarrow A = \sqrt{AA^*} \circ (\tilde{U})^*$$

$$\therefore A = \boxed{\sqrt{AA^*} \circ W} \rightarrow \text{unitary ...}$$

For any $A : V \xrightarrow{\text{lin}} V$, there exists a unitary U_1 , $U_1 : V \rightarrow V$ such that

$$A = \sqrt{AA^*} \circ U_1$$

and there exists a unity $U_2 : V \rightarrow V$ such that

$$A = U_2 \circ \sqrt{A^*A}$$

Now, think complex numbers... $z = |z|e^{i\theta} = \sqrt{z\bar{z}} e^{i\theta}$

→ This called **POLAR DECOMPOSITION**

$$U^*A = \sqrt{A^*A}$$

Now

$$A = U \circ \sqrt{A^*A} \longrightarrow$$

$\sqrt{A^*A}$

$$= \underbrace{(U \circ \sqrt{A^*A} U^*)}_{\text{nonnegative, called } N} U$$

$$A = \sqrt{A^*A} U$$

nonnegative, called N

$$N^2 = (U A^*) (A U^*)$$

$$\therefore \text{Now } A A^* = U (A^* A) U^* = (U A^*) (A U^*) = N^2 \Rightarrow N = \sqrt{A A^*}$$

$$\therefore A = N u = \sqrt{A^* A} u$$

$$\therefore \boxed{A = u \sqrt{A^* A} = \sqrt{A^* A} u}$$

S We can use the same unitary in the polar decomp.
so can just use $u_1 = u_2$.

Note u is not necessarily unique

(*) If A is invertible, then there's a unique choice for u .

→

$$\text{Now, if } A \in M_n, \text{ then } A = u \underbrace{\sqrt{A^* A}}$$

nonnegative, self adj, normal

$$= u (\tilde{u} D \tilde{u}^{-1})$$

↑ diag, nonnegative.

This says

$$A = (u \tilde{u}) \overset{\text{diag}}{\underset{\uparrow}{D}} \tilde{u}^{-1}$$

SVD

for

square
matrix

$$A = \underset{T}{u} \underset{T}{D} \underset{T}{u}^T, \text{ and } D = \begin{bmatrix} s_1 & & \\ & \ddots & \\ & & s_n \end{bmatrix}$$

unitary

where s_i are eigenvalues of $\sqrt{A^* A}$

→ s_i are called "singular values" of A

→ This is SINGULAR VALUE DECOMPOSITION

for square matrices

⑦ What about $\sqrt{AA^*}$?

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SVD

For any $A \in M_n$ we can always find an orthonormal basis v_1, \dots, v_n such that $A(v_1), \dots, A(v_n)$ orthogonal.

i.e. there is a unitary $W \in M_n$ (whose columns are v_1, \dots, v_n) such that

$$A \circ W = A \cdot [v_1, \dots, v_n] = [Av_1, Av_2, \dots, Av_n]$$

$$= \underbrace{\begin{bmatrix} 1 & & & \\ & \ddots & & \\ & & 1 & \\ & & & 1 \end{bmatrix}}_{\text{unitary}} \begin{bmatrix} d_1 & & & \\ & \ddots & & \\ & & d_n & \\ & & & 0 \end{bmatrix} \rightarrow (d_i \geq 0)$$

i.e. $A \circ W = U \circ \begin{bmatrix} \alpha_1 & & & \\ & \ddots & & \\ & & \alpha_n & \\ & & & 0 \end{bmatrix}$ "scaled each column by a nonnegative scalar"

s.t. $A = U \circ \begin{bmatrix} \alpha_1 & & & \\ & \ddots & & \\ & & \alpha_n & \\ & & & 0 \end{bmatrix} \circ W^*$

If $A = U D W$, U, W unitary, D diagonal

then $A^* = W^* D U^*$

$\therefore A^* A = W^* D^2 W = \underbrace{(W^* D W)}_W^2 \rightarrow$ Similarly $U^* D U = \sqrt{AA^*}$

$\therefore W^* D W = \sqrt{A^* A} \leftrightarrow \alpha_i$'s are exactly the eigenvalues of $\sqrt{A^* A}$

Now, $U W W^* D W = U W \sqrt{A^* A}$

$\therefore A = U W \sqrt{A^* A} \rightarrow$ polar decomposition

$$\text{Now } \left\{ \begin{array}{l} W^* D W = \sqrt{A^* A} \\ U D U^* = \sqrt{A A^*} \end{array} \right\}$$

$$\Rightarrow \boxed{\sqrt{A^* A} \simeq \sqrt{A A^*} \text{ (unitarily equivalent)}}$$

$$\text{In particular, } U D U^* = A A^*$$

$$W^* D^2 W = A^* A$$

So

$$\boxed{A^* A \simeq A A^* \text{ (unitary equivalent)}}$$

$\left\{ \begin{array}{l} \text{Since } \sqrt{A^* A} \simeq \sqrt{A A^*} \text{ have the same eigenvalues (eigenprop...)} \\ \text{same with } A^* A = A A^* \end{array} \right\}$

(what if A is not square, i.e. $M_{m \times n}$?)

$$\hat{A}(v_1) \quad \hat{A}(v_2)$$

$$\text{Say, } \hat{A}^* = \boxed{A_{4 \times 7}} \quad \underbrace{\left[\begin{array}{c|c|c|c|c|c|c} v_1 & v_2 & \dots & v_7 \end{array} \right]}_{7 \times 7} = \left(\begin{array}{c} A(v_1) \\ 0 \\ \vdots \\ 0 \end{array} \right) \dots \left(\begin{array}{c} A(v_7) \\ 0 \\ \vdots \\ 0 \end{array} \right)$$

$\left\{ \begin{array}{l} A: \mathbb{C}^7 \rightarrow \mathbb{C}^4 \quad \text{orthonormal} \\ \hat{A}: \mathbb{C}^7 \rightarrow \mathbb{C}^7 \Rightarrow \text{ basis s.t.} \quad \text{orthogonal} \end{array} \right.$

And $\hat{A}(v_1) \dots \hat{A}(v_7)$ orthogonal exactly when $A(v_1) \dots A(v_7)$ orthogonal ...

$$\rightarrow \boxed{A_{4 \times 7} \left[\begin{array}{c|c|c|c} v_1 & \dots & v_7 \end{array} \right] = \left[\begin{array}{c} A(v_1) \dots A(v_7) \end{array} \right]_{4 \times 7}}$$

$$A: \mathbb{C}^7 \rightarrow \mathbb{C}^4$$

↳ to put 0's at the back

$$A = \underbrace{W \cdot \text{Permutation}}_{\substack{\text{unitary}, W \\ \text{unitary}}} = \underbrace{\begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}}_{4 \times 4} \underbrace{\begin{bmatrix} \alpha_1 & & & \\ & \alpha_2 & & \\ & & \ddots & \\ & & & \alpha_7 \end{bmatrix}}_{4 \times 7}$$

So $A = \begin{bmatrix} A \\ 4 \times 7 \end{bmatrix} = \begin{bmatrix} U & O \\ 4 \times 7 & 4 \times 7 \end{bmatrix} \begin{bmatrix} \alpha_1 & & & \\ & \alpha_2 & & \\ & & \ddots & \\ & & & \alpha_7 \end{bmatrix} \begin{bmatrix} W \\ 7 \times 7 \end{bmatrix}$

$$= \begin{bmatrix} U \\ 4 \times 4 \end{bmatrix} \begin{bmatrix} \alpha_1 & & & \\ & \alpha_2 & & \\ & & \ddots & \\ & & & \alpha_7 \end{bmatrix} \begin{bmatrix} W \\ 7 \times 7 \end{bmatrix}$$

Same shape

as A

For $\begin{bmatrix} A \end{bmatrix}$, do same thing, with A^*

What are the α_i 's?

$$D = \begin{bmatrix} S & O \end{bmatrix}$$

$$A = U D W$$

$$A^* = W^* D^* U^*$$

$$D^* = \begin{bmatrix} S \\ O \end{bmatrix}$$

$$\rightarrow \sqrt{A^* A} = W^* \begin{bmatrix} S & O \\ O & O \end{bmatrix} W \quad \rightarrow D^* D = \begin{bmatrix} S^2 & O \\ O & O \end{bmatrix} \rightarrow \sqrt{D^* D} = \begin{bmatrix} S & O \\ O & O \end{bmatrix}$$

$$\begin{array}{c} A^* \\ \downarrow \\ \begin{bmatrix} A^* & A \end{bmatrix} \end{array}$$

The point is of $A^* A, AA^*$

→ Eigenvalues are still the same, 4 of them...
 \rightarrow SVD for non square

$$\sqrt{AA^*} \rightarrow \begin{bmatrix} A \\ A^* \end{bmatrix} \rightarrow D D^* = \begin{bmatrix} S \end{bmatrix}$$

→ In fact, even

In fact, can derive the non-square version of polar decomposition using the same recipe.

Consequences

/ $\|\cdot\|_{\text{Frobenius}}$

① Recall Hilbert-Schmidt norm

$$\begin{aligned}\|A\|_{\text{HS}}^2 &= \text{tr}(A^* A) \\ &= \text{tr}((W^* D^* U^*)(U D W)) \\ &= \text{tr}(\underbrace{W^* D^* D W}_{\text{no need square}})\end{aligned}$$

Now, since $\text{tr}(BC) = \text{tr}(CB)$, so...

matrices for

$$\begin{aligned}\sum \|A\|_{\text{HS}}^2 &= \text{tr}[D^* D W W^*] = \text{tr}(D^* D) \\ &= \sum s_i^2(A)\end{aligned}$$

↙ sum of squares of singular values of A

② Easy to check...

$$\left\| U_1 A U_2 \right\|_{\text{HS}}^2 = \dots = \|A\|_{\text{HS}}^2$$

↙ "Contracting A with unitary don't change size."

③ Let F_c be $n \times n$ matrices of rank k . Given $A \in M_n$, what is the inf $\|A - F\|_{\text{HS}}$ for $F \in F_c$.

i.e. How small can $\|A - F\|_{\text{HS}}$ be?

Observe..

$$\text{Desire } A = UDW$$

$$\text{Then } \|UDW - F\|_{HS} = \|UDW - UW^*FW^*W\|_{HS}$$

$$\begin{aligned} &= \|U(D - U^*FW^*)W\|_{HS} \\ \xrightarrow[\substack{\text{HS norm} \\ \text{is unitarily} \\ \text{invariant}}]{} &= \|D - \underbrace{U^*FW^*}_T\|_{HS} \\ &\quad \downarrow \\ &\quad \text{R} \\ &\quad F \\ &\quad T \end{aligned}$$

\square Now the question becomes how small can we make $\|D - U^*FW^*\|_{HS}$.

$$\underline{\delta} \quad \left\| \underbrace{\begin{pmatrix} a_1 \\ \vdots \\ a_5 \end{pmatrix}}_D - T \right\|_{HS}^2 \leq \left\| \begin{pmatrix} a_1 \\ \vdots \\ a_5 \end{pmatrix} - \begin{pmatrix} a_1 \\ \vdots \\ a_3 \\ 0_0 \end{pmatrix} \right\|_{HS}^2$$

$$\text{Let } T \text{ rank 3, } D \text{ rank 5} \quad = \boxed{a_4^2 + a_5^2}$$

\square Turns out there's a minimum approximation

Detour... if $V = W \oplus W^\perp$, $P_W(z)$ is an element of W

(ortho-projection)

and

$$z = P_W(z) + (I - P_W)(z)$$



orthogonal

$$\text{then } \|z\|^2 = \|P_W(z)\|^2 + \|(I - P_W)(z)\|^2$$

(Pythagorean)

$$\|z\|^2 = \|P_W(z)\|^2 + \|(I - P_W)(z)\|^2$$

GW

\square $P_W(z)$ is an element of W that is closest to z .

Strange Question

Suppose $v_1, \dots, v_7 \in \mathbb{C}^7$. How do we find a 3D subspace $W \subset \mathbb{C}^7$

st

$$(i) \|P_W(v_1)\|^2 + \dots + \|P_W(v_7)\|^2 \text{ is } \begin{cases} \max? \\ \min? \end{cases}$$

+

+

$$(ii) \|P_{W^\perp}(v_1)\|^2 + \dots + \|P_{W^\perp}(v_7)\|^2 \text{ is } \begin{cases} \min? \\ \max? \end{cases}$$

$$\|v_1\|^2 + \dots + \|v_7\|^2$$

B What if (v_i) is $(s_i e_i) \rightarrow$ multiples of standard basis?

$$s_1 \geq s_2 \geq \dots \geq s_7 \geq 0$$

Given $W \subset V$, $\dim W = 3$. Take an

ONB of W , say w_1, w_2, w_3

$$(w_1, \dots, w_7)^T \quad (w_2, \dots, w_7)^T \quad (w_3, \dots, w_7)^T$$

$$P_W(s_i e_i) = \langle s_i e_i, w_1 \rangle w_1 + \langle s_i e_i, w_2 \rangle w_2 + \langle s_i e_i, w_3 \rangle w_3$$

$$= s_i \left[\overline{w_{1j}} w_1 + \overline{w_{2j}} w_2 + \overline{w_{3j}} w_3 \right]$$

$$\text{So } \|P_W(s_i e_i)\|^2 = \underbrace{|s_i|^2}_{s_i^2} \left(|w_{1j}|^2 + |w_{2j}|^2 + |w_{3j}|^2 \right)$$

$$s_i^2$$

$$t_j$$

$$\|P_W(s_j e_j)\|^2 = s_j^2 \left(|w_{1j}|^2 + |w_{2j}|^2 + |w_{3j}|^2 \right)$$

Check $\boxed{t_1 + \dots + t_7 = 3}$

$$\sum_j \|P_W(s_j e_j)\|^2 = \sum s_j^2 t_j = s_1^2 t_1 + s_2^2 t_2 + \dots + s_7^2 t_7$$

To make this as small as possible ... \rightarrow need all the weight on $s_4, s_5, s_6 \dots$
 \rightarrow need $W = \text{span}(e_4, e_5, e_6)$

\hookrightarrow Each $\left\| \begin{pmatrix} s_1 \\ \vdots \\ s_7 \end{pmatrix} - \begin{pmatrix} c_1 & c_2 \\ \vdots & \vdots \end{pmatrix} \right\|^2$
 to exp

$$= \left\| \begin{bmatrix} s_1 e_1 - c_1 & s_2 e_2 - c_2 & \dots & s_5 e_5 - c_5 \\ \vdots & \vdots & & \vdots \end{bmatrix} \right\|^2$$

$$= \|s_1 e_1 - c_1\|^2 + \dots + \|s_5 e_5 - c_5\|^2$$

now, $\underbrace{\text{span}(c_1, c_2, c_3)}_{W}$ is 3D

The closest to $s_i e_i$ for e_i is to set $P_W(s_i e_i) = c_i$

$$\text{so } c_j = P_W(s_j e_j)$$

$$\text{So } \left\| (I - P_W)(s_i e_i) \right\|^2 + \dots + \left\| (I - P_W)(s_5 e_5) \right\|^2$$

$$= \left\| P_W^\perp(s_i e_i) \right\|^2 + \dots + \left\| P_W^\perp(s_5 e_5) \right\|^2$$

"minimizing one is maximizing the other"