

Least Square Approximation / Regression

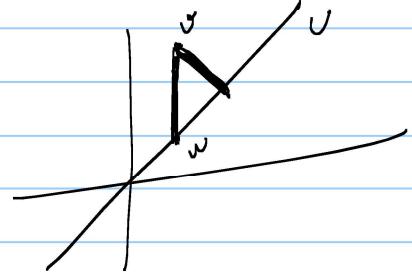
Theorem: Suppose U is a finite dimensional subspace of V , $v \in V$, $u \in U$. Then

$$\|v - P_U(v)\| \leq \|v - u\|$$

The inequality is an equality $\Leftrightarrow u = P_U(v)$.

$$\begin{aligned} \text{Proof: } \|v - P_U(v)\|^2 &\leq \|v - P_U(v)\|^2 + \|P_U(v) - u\|^2 \\ &= \|(v - P_U(v)) + (P_U(v) - u)\|^2 \quad \text{Pythagoras.} \\ &= \|v - u\|^2 \end{aligned}$$

The first inequality = an equality when $\|P_U(v) - u\|^2 = 0$ or $u = P_U(v)$.



Ex: The boiling point of water is known to vary depending on the barometric pressure.
Experiments show the following data

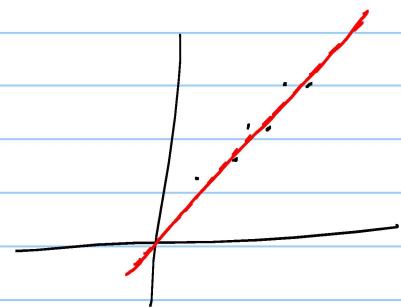
Barometric Pressure	20.2	22.1	24.5	27.3	30.1
Boiling Point	19.5	19.7	20.2	20.9	21.2

Find a linear equation of the form $T = C_0 + C_1 P$ that will make accurate prediction of the boiling point given the pressure.

$$C_0 + 20.2 C_1 = 19.5$$

$$C_0 + 22.1 C_1 = 19.7$$

$$C_0 + 30.1 C_1 = 21.2$$



This system has no solution!

So we try to get an approximate solution!

To find "approximate" solution to a system $Ax=y$ that has no solution

We find an approximate solution by replacing y by a suitable vector \hat{y} in range of A -
choose \hat{y} s.t $\|y-\hat{y}\|$ is minimal (or \hat{y} is closest to y)

Remark: The previous theorem says that the \hat{y} closest to y is $\hat{y} = P_U(y)$ where $U = \text{range } T_A$.

Def: Finding such a solution is called the method of Least square regression. The solution
 \hat{x} for the system $Ax=\hat{y}$ is called a least square solution

Ex: Find the least square solution to $Ax=y$

$$A = \begin{bmatrix} 1 & 4 \\ -3 & 3 \\ 5 & 1 \end{bmatrix} \quad y = \begin{bmatrix} -16 \\ 28 \\ 6 \end{bmatrix}$$

$$\begin{aligned} V &= U \oplus U^\perp \\ y &= u + w \\ P_U(y) &= u \end{aligned}$$

$$U = \text{range } T_A. \quad \hat{y} = P_U(y) = \langle a_1, y \rangle \frac{a_1}{\|a_1\|} + \langle a_2, y \rangle \frac{a_2}{\|a_2\|}$$

$a_1 \cdot a_2 = 0$

$$= \begin{bmatrix} 2 \\ 9 \\ -9 \end{bmatrix}$$

$$A \hat{x} = \hat{y} \Rightarrow \begin{bmatrix} 1 & 4 & 2 \\ -3 & 3 & 9 \\ 5 & 1 & -9 \end{bmatrix} \Rightarrow \hat{x} = \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$$

Definition: If A is a $n \times m$ matrix & $y \in \mathbb{R}^n$, then a least squares solution to $Ax=y$ is a vector \hat{x} s.t

$$\|A\hat{x} - y\| \leq \|Ax - y\|. \quad \forall x \in \mathbb{R}^m.$$

If $Ax=y$ has a solution, then $\hat{x}=x$. If A has linearly independent columns, then \hat{x} is unique. If not, there are infinite many choices of \hat{x} .

Theorem: The set of least square solutions to $\underline{Ax=y}$ is equal to the set of solutions to

$$A^T A x = A^T y$$

(equation)
The solutions to above system are called normal equations / solutions for $Ax=y$.

If A has linearly independent columns, then \exists a unique solution given by

$$\hat{x} = (A^T A)^{-1} A^T y.$$

Proof: Suppose \hat{x} is a least square solution to $Ax=y$.

$\Leftrightarrow \hat{x}$ is a solution to $Ax=\hat{y}$ where $\hat{y} = P_{\mathcal{V}}(y)$ $\mathcal{V} = \text{range } T_A$.

$\Leftrightarrow y - \hat{y} = y - P_{\mathcal{V}}(y) \in \mathcal{V}^\perp = (\text{range } T_A)^\perp = \text{null}(A^T)$.

$$\Leftrightarrow A^T(y - \hat{y}) = 0$$

$$A^T(y - A\hat{x}) = 0 \Rightarrow A^T A \hat{x} = A^T y$$

(One can proceed the proof backwards to get viceversa)

$A^T A$ is invertible $\Leftrightarrow A$ has linearly independent columns

\Rightarrow the least square solution is unique if A has linearly independent columns

Ex: back to bivariate penne/Böhrig example

$$A = \begin{pmatrix} 1 & 20.2 \\ 1 & 22.1 \\ 1 & 24.5 \\ 1 & 27.3 \\ 1 & 30.1 \end{pmatrix} \quad x = \begin{bmatrix} x_0 \\ x_1 \end{bmatrix} \quad y = \begin{bmatrix} 19.5 \\ 19.7 \\ 20.2 \\ 20.9 \\ 21.2 \end{bmatrix}$$

$V = \text{range } T_A = \text{space spanned by col of } A$

The column of A are linearly independent!

$$\Rightarrow \text{solution is given by } (A^T A)^{-1} A^T y = \begin{bmatrix} 157.17 \\ 1.849 \end{bmatrix}$$

Ex: A data is given by an Economist to correlate consumer confidence.

Quarter	1	2	3	4	5	6
confidence index	5	9	8	4	6	8

Find a cubic polynomial that approximates this data.

$$g(t) = c_0 + c_1 t + c_2 t^2 + c_3 t^3$$

$$c_0 + c_1 + c_2 + c_3 = 5$$

$$c_0 + 4 \cdot 1^2 + c_2 \cdot 2^2 + c_3 \cdot 3^2 = 9$$

$$c_0 + 6c_1 + 6^2 c_2 + 6^3 c_3 = 8$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 2^2 & 2^3 \\ 1 & 6 & 6^2 & 6^3 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \end{bmatrix} = \begin{bmatrix} 5 \\ 9 \\ 8 \end{bmatrix}$$

In this case the column are linearly independent

\Rightarrow the least square solution $(A^T A)^{-1} A^T y$

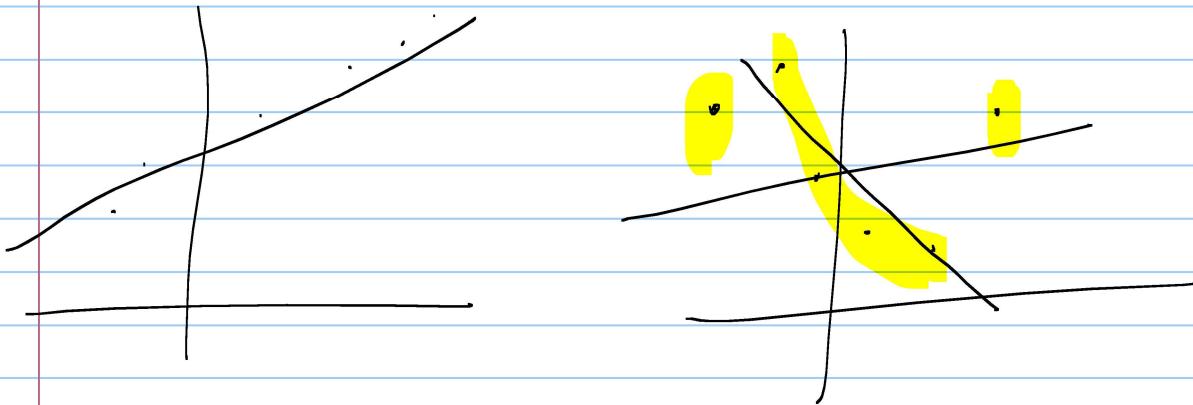
Ex: The times from sunrise to sunset in a certain place is given by the following data -

Date	Nov 20	June 1	Aug 12	Oct 25
Day length	12.17	16.23	12.12	8.18
	0	0.2	0.4	0.6

The length of the day is modelled on the function $L(t) = c_1 + c_2 \sin(2\pi t)$ when $t = \text{time}$ in years.

$t=0$ is evaluated as March 20. Then rest of the data occur in $t=0.2, 0.4, 0.6$

check -



Def: $V = \mathbb{R}^n$ t_1, \dots, t_n are positive numbers in \mathbb{F} .

$$u, v \in V \quad \langle u, v \rangle = t_1 u_1 v_1 + t_2 u_2 v_2 + \dots + t_n u_n v_n.$$

is an inner product on \mathbb{R}^n (check!)

$$Ax = y \quad Ax = \hat{y} \quad \hat{y} = P_U(y) \quad U = \text{range } A$$

$\|y - \hat{y}\|$ is least

$$\|y_1 - \hat{y}_1\|^2 + \dots + \|y_n - \hat{y}_n\|^2$$

Ex.: Use the least square regression to fit a line to the data set

$$(-6, 2.9), (-3, 1.5), (-2, 2), (2, 2.7), (3, 3.3), (6, 1.1)$$

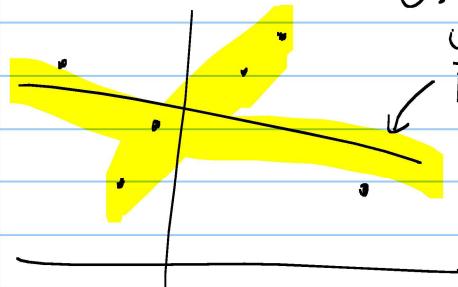
Then use the weighted inner product to solve this problem s.t. middle four pts are more emphasized than the extremal pts.

Sol.: $f(t) = c_0 + c_1 t$

$$c_0 + c_1(-6) = 2.9$$

use usual
inner product

$$c_0 + c_1(6) = 1.1$$



$$\begin{bmatrix} a_1 & a_2 \\ -6 & -3 \\ -2 & 2 \\ 2 & 3 \\ 3 & 6 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \end{bmatrix} = \begin{bmatrix} 2.9 \\ 1.5 \\ 2 \\ 2.7 \\ 3.3 \\ 1.1 \end{bmatrix}$$

$$\hat{y} = P_U(y) = \langle a_1, y \rangle \frac{a_1}{\|a_1\|} + \langle a_2, y \rangle \frac{a_2}{\|a_2\|}$$

$$= 2.25 a_1 - 0.0408 a_2$$

$$U = \text{range } U = \langle a_1, a_2 \rangle$$

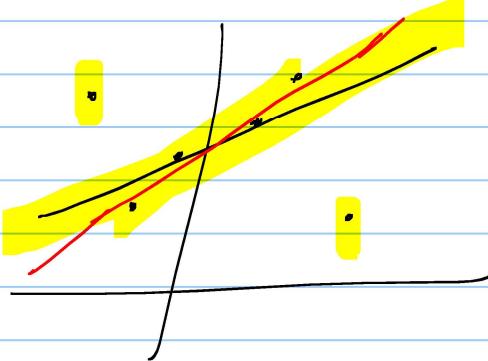
$$Ax = \hat{y} \quad f(t) = 2.25 - 0.0408x.$$

weighted inner product. $t_1 = t_2 = 1 \quad t_2 = t_3 = t_4 = t_5 = 5$

$$(t_1, \dots, t_5) = (1, 5, \dots, 5, 1)$$

Using this inner product. $\hat{y} = P_1(y) = \langle a_1, y \rangle \cdot \frac{a_1}{\|a_1\|} + \langle a_2, y \rangle \cdot \frac{a_2}{\|a_2\|}$

$$= 2.34a_1 + 0.115a_2.$$



§ Fourier Approximations

$$V = C([-π, π])$$

$$\langle f, g \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) g(x) dx$$

Theorem: If $n \geq 1$, the set

$$\{1, \cos x, \cos 2x, \dots, \cos nx, \sin x, \sin 2x, \dots, \sin nx\}$$

is orthogonal on V .

(check): $\langle 1, \cos nx \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} 1 \cdot \cos nx dx = 0$

$$\langle \sin jx, \cos kx \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \sin jx \cdot \cos kx dx = 0 \quad \text{. } \cos kx \cdot \sin jx \text{ is odd!}$$

For any $f(x) \in V$,

$$f_n(x) = P_{F_n}(f(x)) = a_0 + a_1 \cos x + \dots + a_n \cos nx + b_1 \sin x + \dots + b_n \sin nx.$$

F_n = subspace spanned by $\{1, \cos x, \dots, \cos nx, \sin x, \dots, \sin nx\}$

$f_n(x)$ is called the n^{th} Fourier approximation to $f(x)$.

$$a_k = \frac{\langle f, \cos kx \rangle}{\langle \cos kx, \cos kx \rangle}$$

$$= \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \cos kx dx$$

$$b_k = \frac{\langle f, \sin kx \rangle}{\langle \sin kx, \sin kx \rangle}$$

$$= \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \sin kx dx$$

a_k, b_k 's are called Fourier coefficients of $f(x)$.

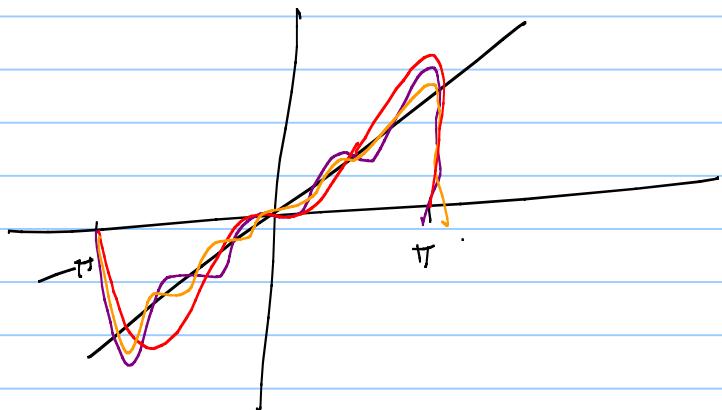
Ex: $f(x) = x$

$$a_0 = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \cdot 1 \cdot dx = 0$$

$$a_k = \frac{1}{\pi} \int_{-\pi}^{\pi} x \cdot \cos kx dx = 0$$

$$b_k = \frac{1}{\pi} \int_{-\pi}^{\pi} x \cdot \sin kx dx = \frac{2}{k} (-1)^{k+1}$$

$$f_n(x) = \sum_{k=1}^n \frac{2}{k} (-1)^{k+1} \sin kx.$$



If the Fourier coefficients decrease in size quickly then we can extend the n^{th} Fourier approximation to

$$a_0 + \sum_{k=1}^{\infty} (a_k \cos kx + b_k \sin kx).$$

is called a Fourier Series.

$$f_2(x) \quad f_3(x) \quad f_4(x)$$

§ Positive Operators

Def: $T \in L(V)$ is called positive if T is selfadjoint &

$$\langle Tu, u \rangle \geq 0$$

Remark: If V is a complex vector space then T being self adjoint can be dropped.

Def: An operator R is called a square root of an operator T if $R^2 = T$.

Ex: ① If U is a subspace of V . Then P_U is a positive operator.

Theorem: $T \in L(V)$. Then T FAE

- ① T is positive
- ② T is selfadjoint & all the eigenvalues of T are nonnegative.
- ③ T has a positive square root.
- ④ T has a selfadjoint square root.
- ⑤ \exists an operator $R \in L(V)$ s.t. $T = R^*R$.

Proof: ① \Rightarrow ② T is already selfadjoint.

Let λ be an eigenvalue of $T \rightarrow \exists v \in V$ s.t. $Tv = \lambda v$.

$$0 \leq \langle Tv, v \rangle = \langle \lambda v, v \rangle = \lambda \langle v, v \rangle = \lambda \|v\|^2 \Rightarrow \lambda \geq 0$$

② \Rightarrow ③ By spectral theorem \exists orthonormal basis e_1, \dots, e_n of V with eigenvalues $\lambda_1, \dots, \lambda_n \geq 0$.

$$\text{Define } R e_i = \sqrt{\lambda_i} e_i \quad R \in L(V) \quad R^2 = T$$

$$\textcircled{3} \Rightarrow \textcircled{4} \quad \checkmark$$

$\textcircled{4} \Rightarrow \textcircled{5}$ T has a selfadjoint square root $R \Rightarrow R^2 = T$

$$R^* R = RR = R^2 = T$$

$$\textcircled{5} \Rightarrow \textcircled{1} \quad T = R^* R.$$

$$T^* = (R^* R)^* = R^*(R^*)^* = R^* R = T \Rightarrow \text{selfadjoint}$$

$$\langle T\varphi, \varphi \rangle_- = \langle R^* R \varphi, \varphi \rangle = \langle R\varphi, R\varphi \rangle \geq 0 \quad \forall \varphi \in V \Rightarrow \text{positive}$$

Theorem: Every positive operator on V has unique positive square root.

Proof: Suppose $T \in \mathcal{L}(V)$ is positive. Suppose v is an eigenvector for $T \Rightarrow \exists \lambda \geq 0$ s.t. $Tv = \lambda v$.

Let R be a positive square root of T . (Claim: $Rv = \sqrt{\lambda} v$).

By spectral theorem \exists orthonormal basis e_1, \dots, e_n of V s.t. e_1, \dots, e_n are eigenvectors of T .
 $\lambda_1, \dots, \lambda_n$ are eigenvalues ≥ 0 .

e_1, \dots, e_n is a basis $\Rightarrow v = a_1 e_1 + \dots + a_n e_n$.

$$\begin{aligned} Rv &= R(a_1 e_1 + \dots + a_n e_n) = a_1 R e_1 + \dots + a_n R e_n \\ &= a_1 \sqrt{\lambda_1} e_1 + \dots + a_n \sqrt{\lambda_n} e_n. \end{aligned}$$

$$\begin{aligned} R^2 v &= R(Rv) = R(a_1 \sqrt{\lambda_1} e_1 + \dots + a_n \sqrt{\lambda_n} e_n) \\ &= a_1 \sqrt{\lambda_1} R e_1 + \dots + a_n \sqrt{\lambda_n} R e_n. \end{aligned}$$

Betr. $R^2 v = T v \Leftrightarrow T v = \lambda v$ \Rightarrow $= \alpha_1 \lambda_1 e_1 + \dots + \alpha_n \lambda_n e_n$

$$\alpha_1 \lambda_1 e_1 + \dots + \alpha_n \lambda_n e_n = \alpha_1 \lambda_1 e_1 + \dots + \alpha_n \lambda_n e_n$$

$$\alpha_1 (\lambda - \lambda_1) e_1 + \dots + \alpha_n (\lambda - \lambda_n) e_n = 0$$

$$\Rightarrow \alpha_i (\lambda - \lambda_i) = 0 \text{ für } i = 1, \dots, n \text{ da } \lambda_i \neq \lambda$$

$$\Rightarrow v = \sum_{\{j : \lambda_j = \lambda\}} \alpha_j e_j$$

$$\Rightarrow Rv = \sum_{\{j : \lambda_j = \lambda\}} \alpha_j \sqrt{\lambda} e_j = \sqrt{\lambda} v$$

§ Isometries.

Def: An operator $S \in \mathcal{L}(V)$ is called an isometry if

$$\|Sv\| = \|v\| \quad \forall v \in V.$$

Ex: ① $d\mathbb{I}$ is an isometry when $(\lambda) = 1$

② $\lambda_1, \dots, \lambda_n$ are scalars. & $|\lambda_i| = 1 \forall i$ $S \in \mathcal{L}(V)$ $S e_j = \lambda_j e_j$ for some orthonormal basis e_1, \dots, e_n .

$$v \in V \quad v = \langle v, e_1 \rangle e_1 + \dots + \langle v, e_n \rangle e_n.$$

$$\begin{aligned} \|v\|^2 &= |\langle v, e_1 \rangle|^2 + \dots + |\langle v, e_n \rangle|^2 \\ &= |a_1|^2 + \dots + |a_n|^2 \end{aligned}$$

$$\begin{aligned} S v &= S(\langle v, e_1 \rangle e_1 + \dots + \langle v, e_n \rangle e_n) \\ &= \langle v, e_1 \rangle S e_1 + \dots + \langle v, e_n \rangle S e_n \\ &= \langle v, e_1 \rangle \lambda_1 e_1 + \dots + \langle v, e_n \rangle \lambda_n e_n. \end{aligned}$$

$$\|Sv\|^2 = |\langle v, e_1 \rangle \lambda_1|^2 + \dots + |\langle v, e_n \rangle \lambda_n|^2$$

$$= |\langle v, e_1 \rangle|^2 + \dots + |\langle v, e_n \rangle|^2$$

$$\Rightarrow \|Sv\| = \|v\| \quad \forall v \in V$$

$\Rightarrow S$ is an isometry.

Theorem: $S \in L(V)$ TFAE

① S is an isometry

② $\langle Su, Sv \rangle = \langle u, v \rangle \quad \forall u, v \in V$

③ $S e_1, \dots, S e_n$ is orthonormal if orthonormal list of vectors e_1, \dots, e_n in V

④ \exists an orthonormal basis e_1, \dots, e_n of V s.t. $S e_1, \dots, S e_n$ is orthonormal.

⑤ $S^* S = I$

⑥ $S S^* = I$

(7) S^* is also an isometry.

(8) S is invertible $\Leftrightarrow S^{-1} = S^*$.

Proof: (1) \Rightarrow (2) - Suppose S is an isometry: $\|Su\| = \|v\| \quad \forall v \in V$.

Suppose V is an inner product space over \mathbb{R} :

$$\begin{aligned}\langle u, v \rangle &= \|u+v\|^2 - \|u-v\|^2 \\ &\quad + \|u+v\|^2 - \|u-v\|^2 \\ &= \|u+v\|^2 - \|u-v\|^2 \\ &= \|u+v\|^2 - \|u-v\|^2\end{aligned}$$

$$\begin{aligned}\langle Su, Sv \rangle &= \frac{\|Su+Sv\|^2 - \|Su-Sv\|^2}{2} \\ &= \frac{\|S(u+v)\|^2 - \|S(u-v)\|^2}{2} \\ &= \frac{\|u+v\|^2 - \|u-v\|^2}{2} = \langle u, v \rangle\end{aligned}$$

Suppose V is an inner product space over \mathbb{C}

Check use the formula

$\textcircled{2} \Rightarrow \textcircled{3}$ Suppose e_1, \dots, e_n forms an orthonormal basis of V

$$\langle S e_k, S e_j \rangle = \langle e_k, e_j \rangle \quad (\text{by assumption } \textcircled{2})$$

$= 0 \Rightarrow S e_k, S e_j \text{ are orthogonal.}$

$\Rightarrow S e_1, \dots, S e_n$ forms an orthonormal basis $\|S e_i\| = \|e_i\| = 1$

$\textcircled{3} \Rightarrow \textcircled{4} \checkmark$

$\textcircled{4} \Rightarrow \textcircled{5}$ Suppose e_1, \dots, e_n is an orthonormal basis of S & $S e_1, \dots, S e_n$ forms orthonormal basis.

$$\langle S^* S e_j, e_k \rangle$$

//

$$\langle S e_j, S e_k \rangle = \langle e_j, e_k \rangle$$

If vectors $u, v \in V$. they can be written as a linear combination of vectors e_1, \dots, e_n

$$\Rightarrow \langle S^* S u, v \rangle = \langle u, v \rangle. \quad \forall u, v \in V$$

$$\Rightarrow S^* S = I$$

$$\textcircled{5} \Rightarrow \textcircled{6} \checkmark$$

$$\textcircled{6} \Rightarrow \textcircled{7} \text{ Supp } S^* S^T = I. \quad v \in V. \quad (S^*)^* = S$$

$$\|S^* v\|^2 = \langle S^* v, S^* v \rangle = \langle S S^* v, v \rangle = \langle v, v \rangle = \|v\|^2.$$

$\Rightarrow S^*$ is an isometry.

$\textcircled{7} \Rightarrow \textcircled{8}$ Supp S^* is an isometry. (use the implications $\textcircled{1} \Rightarrow \textcircled{5} \Rightarrow \textcircled{6}$) to see that

$$S S^* = S^* S = I$$

$$\Rightarrow S^{-1} = S^*.$$

$\textcircled{8} \Rightarrow \textcircled{1}$ S is invertible $\Leftrightarrow S^{-1} = S^*$

$v \in V.$

$$\|Sv\|^2 = \langle S v, S v \rangle = \langle S^* S v, v \rangle = \langle v, v \rangle = \|v\|^2$$

$$\Rightarrow \|Sv\| = \|v\| \Rightarrow S \text{ is an isometry.}$$

Theorem : Suppose V is a complex inner product space $s \in L(V)$. Then TFAE

① s is an isometry

② \exists an orthonormal basis of V consisting of eigenvectors of s which corresponds to eigenvalues all have absolute value 1.

Proof: ① \Rightarrow ②

Suppose s is an isometry.

By the spectral theorem \exists an orthonormal basis e_1, \dots, e_n which are all eigenvectors.
 $\lambda_1, \dots, \lambda_n$ \rightarrow eigenvalues.

$$|\lambda_j| = \|s e_j\| = \|s^* s e_j\| = \|s e_j\| = \|e_j\| = 1$$

② \Rightarrow ① Ex ② above.



§ Polar & Singular value Decomposition.

Remark: $T \in L(V) \Rightarrow T^*T$ is self adjoint & positive $\langle T^*T v, v \rangle = \langle T v, T v \rangle \geq 0$.
 $\Rightarrow -\sqrt{T^*T}$ exists!

$$\begin{aligned} T \in L(V) \\ \sqrt{T^*T} \end{aligned}$$

Theorem: Polar Decomposition: $T \in L(V)$ Then \exists isometry $S \in L(V)$ s.t

$$T = S \sqrt{T^*T}$$

Proof: $v \in V$

$$\begin{aligned} \|Tv\|^2 &= \langle Tv, Tv \rangle = \langle T^*T v, v \rangle = \langle \sqrt{T^*T} \sqrt{T^*T} v, v \rangle \\ &= \langle \sqrt{T^*T} v, \sqrt{T^*T} v \rangle \quad \begin{matrix} \sqrt{T^*T} v \\ \text{self adjoint} \end{matrix} \\ &= \|\sqrt{T^*T} v\|^2 \quad \begin{matrix} (\text{self adjoint}) \end{matrix} \end{aligned}$$

$$\|Tv\|^2 = \|\sqrt{T^*T}v\|^2$$

$$S_1: \text{range } \sqrt{T^*T} \rightarrow \text{range } T$$

$$\sqrt{T^*T}v \mapsto Tv$$

check S_1 is linear.

Suppose $v_1, v_2 \in V$ & $\sqrt{T^*T}v_1 = \sqrt{T^*T}v_2$.

$$\begin{aligned}\|Tv_1 - Tv_2\|^2 &= \|T(v_1 - v_2)\|^2 = \|\sqrt{T^*T}(v_1 - v_2)\|^2 \\ &= \|\sqrt{T^*T}v_1 - \sqrt{T^*T}v_2\|^2 = 0\end{aligned}$$

$$Tv_1 = Tv_2$$

S_1 is clearly onto (surjective)

S_1 is also injective : $\|S_1v\| = \|v\| \quad v \in \text{range } \sqrt{T^*T} - \text{domain } S_1$

By rank-Nullity theorem $\dim \text{range } \sqrt{T^*T} = \dim \text{range } T$

$$\dim (\text{range } \sqrt{T^*T})^\perp = \dim (\text{range } T)^\perp$$

$$V = T \oplus L(V)$$

$$V \oplus V^\perp$$

$S_2 \in \mathcal{L}((\text{range } \sqrt{T^*T})^\perp)$. Choose an orthonormal basis e_1, \dots, e_m of $(\text{range } \sqrt{T^*T})^\perp$
 f_1, \dots, f_m of $(\text{range } T)^\perp$

$$S_2 \left(\sum c_i e_i \right) = \sum c_i f_i$$

$$\forall w \in (\text{range } \sqrt{T^*T})^\perp \Rightarrow \|\sum c_i f_i\| = \|w\| = \sqrt{|c_1|^2 + \dots + |c_m|^2}$$

$$\frac{\|\sum c_i f_i\|}{\sqrt{|c_1|^2 + \dots + |c_m|^2}}$$

$$S : V \longrightarrow V$$

$$S(v) = \begin{cases} S_1(v) & \text{if } v \in \text{range } \sqrt{T^*T} \\ S_2(v) & \text{if } v \in (\text{range } \sqrt{T^*T})^\perp \end{cases}$$

$$\text{Let } v \in V = \text{range } \sqrt{T^*T} \oplus (\text{range } \sqrt{T^*T})^\perp$$

$$v = u + w$$

$$S\omega = S(u + \omega) = Su + Sw = s_1 u + s_2 \omega.$$

$$S(\sqrt{T^*T}\omega) = s_1(\sqrt{T^*T}\omega) = T\omega \quad \forall \omega \in V.$$

$$\Rightarrow T = S\sqrt{T^*T}$$

$$\begin{aligned} S \text{ is an isometry!} : \|S\omega\|^2 &= \|s_1 u + s_2 \omega\|^2 = \|s_1 u\|^2 + \|s_2 \omega\|^2 = \|u\|^2 + \|\omega\|^2 \\ &= \|u + \omega\|^2 = \|v\|^2. \end{aligned}$$

Remark: s_1 is an isometry $\Rightarrow \exists$ orthonormal basis B_1 , $s_1 \perp \mu(s_1, B_1)$ is diagonal

$\sqrt{T^*T}$ is positive \Rightarrow " B_2 s.t. $\mu(\sqrt{T^*T}, B_2)$ is diagonal.

But B_1 may not be the same as B_2 .

Def: $T \in L(V)$. The singular values of T are eigenvalues of $\sqrt{T^*T}$ with each eigenvalue λ repeated $\dim E(\lambda, \sqrt{T^*T})$ times.

$$\text{Ex: } T \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} 3z_1 \\ 2z_2 \\ 0 \\ 3z_4 \end{bmatrix} \quad T^*T \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} 9z_1 \\ 4z_2 \\ 0 \\ 9z_4 \end{bmatrix} \quad \sqrt{T^*T} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} 3z_1 \\ 2z_2 \\ 0 \\ 3z_4 \end{bmatrix}$$

Eigenvalues of $\sqrt{T^*T}$ is $3, 2, 0$. $\dim E(3, \sqrt{T^*T}) = 2$

$$\dim E(2, \sqrt{T^*T}) = 1$$

$$\dim E(0, \sqrt{T^*T}) = 1$$

Theorem: $T \in \mathcal{L}(V)$ have singular values s_1, \dots, s_n . Then \exists orthonormal basis e_1, \dots, e_n & f_1, \dots, f_n of V s.t.

$$T\varphi = s_1 \langle v, e_1 \rangle f_1 + \dots + s_n \langle v, e_n \rangle f_n$$

Proof: $\sqrt{T^*T}$ is a positive operator ($\&$ hence self-adjoint)

By Spectral Theorem \exists a basis e_1, \dots, e_n of V s.t. $\sqrt{T^*T} e_i = s_i e_i$

$$\varphi = \langle \varphi, e_1 \rangle e_1 + \dots + \langle \varphi, e_n \rangle e_n$$

$$\begin{aligned} \sqrt{T^*T} \varphi &= \langle \varphi, e_1 \rangle \sqrt{T^*T} e_1 + \dots + \langle \varphi, e_n \rangle \sqrt{T^*T} e_n \\ &= \langle \varphi, e_1 \rangle s_1 e_1 + \dots + \langle \varphi, e_n \rangle s_n e_n \end{aligned}$$

Using Polar decomposition \exists isometry $S \in \mathcal{L}(V)$ s.t. $T = S\sqrt{T^*T}$.

$$S\sqrt{T^*T} \varphi = \langle \varphi, e_1 \rangle s_1 e_1 + \dots + \langle \varphi, e_n \rangle s_n e_n$$

S is an isometry \Rightarrow s_1, \dots, s_n is an orthonormal basis. set $Se_i = f_i$

$$T\varphi = \sqrt{T^*T} \varphi = \langle \varphi, e_1 \rangle s_1 f_1 + \dots + \langle \varphi, e_n \rangle s_n f_n.$$

T is diagonalizable $\Rightarrow \exists$ a basis e_1, \dots, e_n s.t. $\mu(T, e_1, \dots, e_n)$ is diagonal.

Now let $T \in L(V)$ then by the singular value decomposition.

$\mu(T; e_1, \dots, e_n; f_1, \dots, f_n)$ diagonal!!

$$\varphi = \sum c_i e_i$$

$$T\varphi = \sum d_i f_i$$

$$T(\varphi) = \sum d_i f_i$$

$$\simeq \begin{bmatrix} s_1 & & \\ & \ddots & \\ 0 & & s_n \end{bmatrix}$$

PD: $T \in L(V) \Rightarrow T = S\sqrt{T^*T}$ where S is an isometry

SVD: $T \in L(V)$, s_1, \dots, s_n are eigenvalues of $\sqrt{T^*T}$ $\Rightarrow \exists$ orthonormal basis e_1, \dots, e_n & f_1, \dots, f_n st.

(singular values of T)

$$Te_i = s_1 \langle v, e_1 \rangle f_1 + \dots + s_n \langle v, e_n \rangle f_n$$

T is diagonalizable if \exists basis e_1, \dots, e_n st $\mu(T; e_1, \dots, e_n; f_1, \dots, f_n) = \text{diagonal}$

$$\mu(T; e_1, \dots, e_n; f_1, \dots, f_n) = \begin{bmatrix} s_1 & & & \\ & \ddots & & \\ & & \ddots & \\ & & & s_n \end{bmatrix}$$

Recall that T^*T is a positive operator

Theorem: $T \in L(V)$. Then the singular values of T are nonnegative square roots of the eigenvalues of T^*T & each eigenvalue is repeated $\dim E(\lambda, T^*T)$ times.

§ Generalized Eigenvectors & Nilpotent Operators.

Lemma: $T \in L(V)$. Then

$$\{0\} = \text{null } T^0 \subseteq \text{null } T^1 \subseteq \text{null } T^2 \subseteq \dots$$

Proof: k is nonnegative integer. $v \in \text{null } T^k \stackrel{?}{\subseteq} \text{null } T^{k+1}$.

$$\begin{array}{c} \downarrow \\ T^K v = 0 \\ \downarrow \\ T^{K+1} v = T(T^K v) = T(0) = 0 \end{array}$$

Lemma: $T \in L(V)$. Suppose m is positive integer. s.t. $\text{null } T^m = \text{null } T^{m+1}$

$$\text{Then } \text{null } T^m = \text{null } T^{m+1} = \text{null } T^{m+2} = \dots = \text{null } T^{m+k+1} = \dots$$

Proof: WTS $\text{null } T^{m+k} = \text{null } T^{m+k+1}$.

$$\subseteq \checkmark \quad (\text{from previous lemma})$$

$$\begin{aligned}
 v \notin \text{null } T^{m+k+1} &\Rightarrow T^{m+k+1}(v) = 0 \\
 &\quad || \\
 T^{m+1}(T^k v) &= 0. \\
 &\quad \downarrow \\
 T^k v &\in \text{null } T^{m+1} = \text{null } T^m \Rightarrow T^m(T^k v) = 0. \\
 &\quad || \\
 T^{m+k} v &= 0 \\
 v &\in \text{null } T^{m+k}.
 \end{aligned}$$

$\text{null } T^{m+k+1} \subseteq \text{null } T^{m+k}$

Theorem: $T \in L(V)$. $n = \dim V$ Then

$$\text{null } T^n = \text{null } T^{n+1} = \text{null } T^{n+2} = \dots$$

Proof: Suppose $\text{null } T^0 \subsetneq \text{null } T^1 \subsetneq \text{null } T^2 \subsetneq \dots$ (all are subspaces of V)

$$\dim \text{null } T^0 < \dim \text{null } T^1 < \dim \text{null } T^2 < \dots$$

$$\Rightarrow \dim \text{null } T^{n+1} \geq n+1 \rightarrow \leftarrow.$$

$$\Rightarrow \text{null } T^n = \text{null } T^{n+1} = \text{null } T^{n+2} = \dots$$

Compare
with
minor
problem.

Theorem: $T \in \mathcal{L}(V)$, $n = \dim V$. Then

$$V = \text{null } T^n \oplus \text{range } T^n.$$

$$V = \text{null } T^n + \text{range } T^n$$

$$\text{null } T^n \cap \text{range } T^n = \{0\}.$$

Proof: $v \in \text{null } T^n \cap \text{range } T^n$.

$$v \in \text{null } T^n \Rightarrow T^n v = 0$$

$$v \in \text{range } T^n \Rightarrow \exists u \in V \text{ s.t. } T^n u = v$$

$$T^n(T^n u) = T^{2n} u = 0$$

$$\Rightarrow u \in \text{null } T^{2n} = \text{null } T^n \text{ (from previous result)}$$

$$\Rightarrow T^n u = 0$$

//

$$0 = \text{null } T^n \cap \text{range } T^n$$

$$\dim(\text{null } T^n \oplus \text{range } T^n) = \dim \text{null } T^n + \dim \text{range } T^n = \dim V. \quad T^n \in \mathcal{L}(V).$$

$$\Rightarrow \text{null } T^n \oplus \text{range } T^n = V.$$

Ex. $T \in \mathcal{L}(F^3)$

$$T \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} z_2 \\ 0 \\ -z_3 \end{bmatrix}$$

$\text{null } T + \text{range } T$ is not a direct sum.

$$\text{null } T = \left\{ \begin{bmatrix} z_1 \\ 0 \\ 0 \end{bmatrix} \right\}$$

$$\text{range } T = \left\{ \begin{bmatrix} z_1 \\ 0 \\ z_3 \end{bmatrix} \right\}.$$

$$T^3 \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 12z_3 \end{bmatrix}$$

$$\text{null } T^3 = \left\{ \begin{bmatrix} z_1 \\ z_2 \\ 0 \end{bmatrix} \right\}$$

$$\text{range } T^3 = \left\{ \begin{bmatrix} 0 \\ 0 \\ z_3 \end{bmatrix} \right\}$$

$$V = \text{null } T^3 \oplus \text{range } T^3.$$

T is diagonalizable $\Leftrightarrow \exists$ basis v_1, \dots, v_n $M(T; v_1, \dots, v_n)$ is diagonal
 $\Leftrightarrow V = E(\lambda_1, T) \oplus \dots \oplus E(\lambda_m, T)$

Def: $T \in L(V)$. λ is an eigenvalue of T
 $v \in V$ is called a generalized eigenvector of T correspondingly to λ , if $v \neq 0$ &

$$(T - \lambda I)^j v = 0 \quad \text{for some positive integer } j.$$

Def: $T \in L(V)$. $\lambda \in \mathbb{F}$. The generalized eigenspace of T correspondingly to λ , denoted by $G_\lambda(T)$
is the set of all generalized eigenvectors of T correspondingly to λ , along with 0 vector.

Every eigenvector is a generalized eigenvector

$$\Rightarrow E(\lambda, T) \subseteq G_\lambda(T).$$

$$G_\lambda(T) = \bigcup_{j \geq 0} \text{null}(T - \lambda I)^j$$

$$\text{null}(T - \lambda I) \subseteq \text{null}(T - \lambda I)^2 \subseteq \text{null}(T - \lambda I)^3 \subseteq \dots$$

$n = \dim V$

$$\text{null}(T - \lambda I)^n = \text{null}(T - \lambda I)^{n+1} = \dots$$

Theorem: $T \in \mathcal{L}(V)$ $\lambda \in \mathbb{F}$, $G_T(\lambda, T) = \text{null}(T - \lambda I)^{\dim V}$

Proof: $v \in \text{null}(T - \lambda I)^n$ $n = \dim V$.

$$\Rightarrow v \in G_T(\lambda, T)$$

Conversely if $v \in G_T(\lambda, T)$ $\Rightarrow v \in \text{null}(T - \lambda I)^j$ for some j
 $\subseteq \text{null}(T - \lambda I)^n$ $n = \dim V$.

Ex: $T \in \mathcal{L}(\mathbb{C}^3)$

$$T \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} \bar{z}_2 \\ 0 \\ \bar{z}_3 \end{bmatrix}$$

Eigenvalues are 0 & 5.

$$E(0, T) = \left\{ \begin{bmatrix} z_1 \\ 0 \\ 0 \end{bmatrix} \right\} \quad V \neq E(0, T) \oplus E(5, T)$$

$$E(5, T) = \left\{ \begin{bmatrix} 0 \\ 0 \\ z_3 \end{bmatrix} \right\}$$

$$T^3 \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 125z_3 \end{bmatrix}.$$

$$G_T(0, T) = \left\{ \begin{bmatrix} z_1 \\ z_2 \\ 0 \end{bmatrix} \right\} \quad G_T(5, T) = \left\{ \begin{bmatrix} 0 \\ 0 \\ z_3 \end{bmatrix} \right\}$$

$$V = G_T(0, T) \oplus G_T(5, T)$$

Theorem: $T \in L(V)$. $\lambda_1, \dots, \lambda_m$ are distinct eigenvalues of T .

v_1, \dots, v_m are generalized eigenvectors of T

Then v_1, \dots, v_m are linearly independent

Generalized Eigenvector: $v \in V$, $T \in \mathcal{L}(V)$, λ is an eigenvalue of T , $(T - \lambda I)^j v = 0$ for some j
 $\Leftrightarrow v \in \text{null}((T - \lambda I)^j)$ for some j

Generalized Eigenspace: $G_r(\lambda, T) = \text{collection all generalized eigenvectors} = \bigcup_j \text{null}((T - \lambda I)^j)$
 $G_r(\lambda, T) = \text{null}((T - \lambda I)^n)$ $n = \dim V$

$$E(\lambda, T) = \text{null}(T - \lambda I) \subseteq G_r(\lambda, T)$$

$$T \in \mathcal{L}(V)$$

$$E(\lambda_1, T) \oplus \dots \oplus E(\lambda_m, T) \leq V$$

$$G_r(\lambda_1, T) \oplus \dots \oplus G_r(\lambda_m, T) = V$$

Theorem: $T \in \mathcal{L}(V)$. $\lambda_1, \dots, \lambda_m$ are distinct eigenvalues of T .

v_1, \dots, v_m are generalized eigenvectors of T

$\Rightarrow v_1, \dots, v_m$ are linearly independent

Proof: $a_1 v_1 + \dots + a_m v_m = 0$ \circledast

v_i 's are generalized eigenvectors.

$v_i \in \text{null } (T - \lambda_i I)^n \quad G_i(\lambda_i, T)$

$n = \dim V$.

v_i is a generalized eigenvector $\Rightarrow \exists$ some k s.t. $(T - \lambda I)^k v_i = 0$

\Rightarrow choose k to be the largest integer s.t. $(T - \lambda I)^{k-1} v_i \neq 0$
 $\Rightarrow (T - \lambda I)^k v_i = 0$

$$w = (T - \lambda I)^{k-1} v_i$$

$$(T - \lambda I) w = (T - \lambda I) (T - \lambda I)^{k-1} v_i = (T - \lambda I)^k v_i = 0$$

$$\Rightarrow T w = \lambda w$$

$$(T - \lambda I) w = T w - \lambda w = \lambda_1 w - \lambda w = (\lambda_1 - \lambda) w$$

Apply $(T - \lambda_1 I)^{k-1}$, $(T - \lambda_2 I)^n$, ..., $(T - \lambda_m I)^n$ to \circledast .

$$a_1 (T - \lambda_1 I)^{k-1} (T - \lambda_2 I)^n \dots (T - \lambda_m I)^n v_i = 0$$

v_i 's are generalized eigenvectors $v_i \in \text{null } (T - \lambda_i I)^n$.

$$a_1 (T - \lambda_2 I)^n \dots (T - \lambda_m I)^n w = 0$$

$$a_1 (\lambda_1 - \lambda_2)^\alpha \cdots (\lambda_1 - \lambda_m)^\alpha \omega = 0$$

λ 's are distinct $\Rightarrow \lambda_1 - \lambda_2 \neq 0 \cdots (\lambda_1 - \lambda_m) \neq 0 \quad \omega \neq 0 \quad \Rightarrow \alpha_1 = 0$.

Keep repeating this process to see that $a_2 = a_3 = \dots = a_m = 0$.

$\Rightarrow v_1, \dots, v_m$ are linearly independent.

§ Nilpotent Operator.

Def: An operator $T \in \mathcal{L}(V)$ is nilpotent if $T^j = 0$ for some j .

Ex: ① $V = P_k(\mathbb{R})$ $D \in \mathcal{L}(V)$ is the differential operator $D(f) = f' \quad D^5 = 0$

② $V = \mathbb{R}^4$ $T \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} z_3 \\ z_4 \\ 0 \\ 0 \end{bmatrix}$ $T^2 = 0$

Theorem: $N \in L(V)$, Then $N^n = 0$ $n = \dim V$.

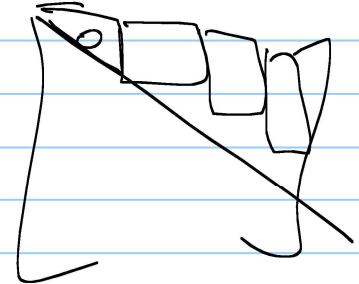
$$G_r(\lambda, T) = \text{null } (T - \lambda I)^n$$

Proof: N is nilpotent $\Rightarrow G_r(0, N) = V \Rightarrow \text{null } N^n = V$.

$$\text{null } (N - 0I)^n = \text{null } N^n$$

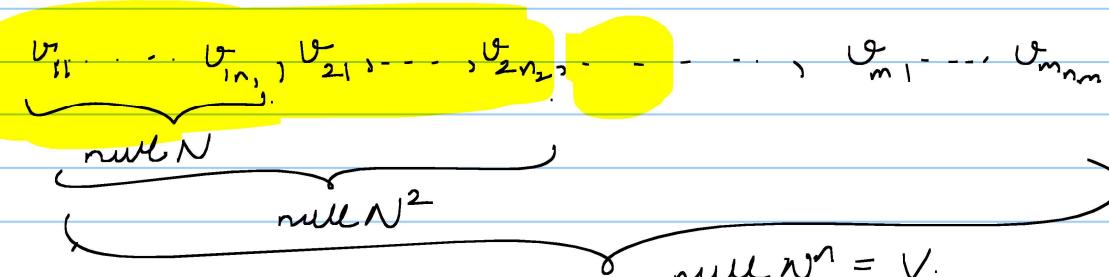
Theorem: N is an nilpotent operator on V . Then \exists a basis of V with respect to which the matrix of N has the form

$$\begin{bmatrix} 0 & * \\ 0 & \ddots \end{bmatrix}$$



Proof: $V = \text{null } N^n$ for $n = \dim V$.

Basis for V is $v_{11}, \dots, v_{1n_1}, v_{21}, \dots, v_{2n_2}, \dots, v_m, \dots, v_{mn_m} = \mathcal{B}$



$$\begin{aligned}
 N(\omega_{ii}) &= \text{null } N^i \\
 \text{null } N^i &= \text{null } N^{i+1} \\
 \text{null } N^{i+1} &= \text{null } N^{i+2} \\
 \vdots &\quad \vdots \\
 \text{null } N^{n-1} &= \text{null } N^n \\
 \text{null } N^n &= \text{null } N^{n+1} \\
 \text{null } N^{n+1} &= \text{null } N^{n+2} \\
 \vdots &\quad \vdots \\
 \text{null } N^{m-1} &= \text{null } N^m \\
 \text{null } N^m &= \text{null } N^{m+1} \\
 \vdots &\quad \vdots \\
 \text{null } N^{N-1} &= \text{null } N^N
 \end{aligned}$$

$$M(T; B) = \begin{pmatrix} 0 & 0 & a_{11} & a_{12} & \cdots \\ 0 & 0 & a_{21} & a_{22} & \cdots \\ \vdots & \vdots & \ddots & \ddots & \ddots \\ 0 & 0 & a_{n1} & a_{n2} & \cdots \\ 0 & 0 & 0 & 0 & \ddots \end{pmatrix}$$

$\underbrace{\qquad\qquad}_{n_1}$ $\underbrace{\qquad\qquad}_{n_2}$

$$\begin{aligned}
 \omega &\in \text{null } N^i \\
 N(\omega) &\in \text{null } N^{i-1}
 \end{aligned}$$

$$\begin{aligned}
 x^{(px)} &= x^3 + 3x^2 + x \\
 p(x)x &= \\
 p(x) &= x^2 + 2x + 1 \\
 p(T) &= T^2 + 2T + 1
 \end{aligned}$$

§ Decomposition of Operator

Theorem: $T \in L(V)$, $p \in P(F)$. Then null $p(T)$ & range $p(T)$ are invariant under T .

Proof: $v \in \text{null}(p(T))$ $p(T)v = 0$

$$p(T)(Tv) = (p(T) \cdot T)v = (T \cdot p(T))v = T(p(T)v) = T(0) = 0$$

$$v \in \text{range}(P(T)) = \{v = P(T)u\}$$

$$Tv = T(P(T)u) = P(T)(Tu). \Rightarrow T \text{ is range } P(T).$$

Theorem: V is a complex vector space. $T \in L(V)$. Let $\lambda_1, \dots, \lambda_m$ are distinct eigenvalues of T . Then

$$\textcircled{1} \quad V = G_1(\lambda_1, T) \oplus \dots \oplus G_1(\lambda_m, T)$$

\textcircled{2} each $G_1(\lambda_i, T)$ are invariant under T

\textcircled{3} each $(T - \lambda_j I)|_{G_1(\lambda_j, T)}$ is nilpotent.

Proof: \textcircled{2} $G_1(\lambda_i, T) = \text{null } (T - \lambda_i I)^n$ $n = \dim V$. invariant under T by previous theorem.

\textcircled{3} $G_1(\lambda_j, T) = \text{null } (T - \lambda_j I)^n$ $n = \dim V$.

$$\forall v \in G_1(\lambda_j, T) \Rightarrow (T - \lambda_j I)^n v = 0 \Rightarrow (T - \lambda_j I)^n = 0 \text{ on } G_1(\lambda_j, T)$$

$$\Rightarrow (T - \lambda_j I)|_{G_1(\lambda_j, T)} \text{ is nilpotent.}$$

(1) Induct on $n = \dim V$.

If $n=1$, T is already diagonalizable $\Rightarrow V = E(\lambda_1, T) = G_1(\lambda_1, T)$ ✓

Suppose $n > 1$. & assume by induction hypothesis that it is true for vector space of dimension $< n$.

V is a complex vector space $\Rightarrow \exists \lambda_1$, eigenvalue of T :

$$(T - \lambda_1 I) \in L(T)$$

$$V = \text{null } (T - \lambda_1 I)^n \oplus \text{range } (T - \lambda_1 I)^n$$

$$G_1(\lambda_1, T)$$

$\dim U < \dim V$ By induction hypothesis on $T|_U$

$$U = G_1(\lambda_2, T|_U) \oplus \dots \oplus G_r(\lambda_m, T|_U)$$



$$E(\lambda, T) = \{v : T_v = \lambda v\} = \text{null}(T - \lambda I)$$

$$G_i(\lambda, T) = \{v : (T - \lambda I)^i v = 0 \text{ for some } i\} = \text{null}(T - \lambda I)^{\dim V}$$

$$V = E(\lambda_1, T) \oplus \dots \oplus E(\lambda_n, T)$$

$P(T) \subset P(T)$

null $P(T)$, range $P(T)$
invariant under

Theorem: V is a complex vector space. $T \in L(V)$. $\lambda_1, \dots, \lambda_m$ are eigenvalues

$$\textcircled{1} \quad V = G_1(\lambda_1, T) \oplus \dots \oplus G_1(\lambda_m, T)$$

\textcircled{2} $G_i(\lambda_j, T)$ is invariant under T

\textcircled{3} each $(T - \lambda_j I)|_{G_i(\lambda_j, T)}$ is nilpotent

on $G_i(\lambda_j, T)$

$$(T - \lambda_j I)^{\dim V} v = 0$$

$v \in G_i(\lambda_j, T)$

$$\Rightarrow (T - \lambda_j I)^{\dim V} = 0$$

$\Rightarrow T - \lambda_j I$ is nilpotent.

Proof: \textcircled{1} Induction $\dim V = n$.

$$n=1 \checkmark$$

Suppose result is true for all vector spaces of dim $< n$.

V is complex vector space $\Rightarrow \exists \lambda_1 \in \mathbb{C}$ eigenvalue

$$V = \underbrace{G(\lambda_1, T)}_{\text{null } (T - \lambda I)^n} \oplus \underbrace{U}_{\text{range } (T - \lambda I)^n} \quad (\text{result: } V = \text{null } T^n \oplus \text{range } T^n)$$

$$\dim U = \dim \text{range } (T - \lambda I)^n < \dim V$$

→ By induction hypothesis $\lambda_2, \dots, \lambda_m$ s.t.

applied to $T|_U \in \mathcal{L}(U)$

$$U = G(\lambda_2, T|_U) \oplus \dots \oplus G(\lambda_m, T|_U)$$

$$G(\lambda_j, T|_U) = \left\{ u \in U : (T - \lambda_j I)^k u = 0 \text{ for some } k \right\}$$

$$G(\lambda_j, T) = \left\{ v \in V : (T - \lambda_j I)^k v = 0 \text{ for some } k \right\}$$

Any generalized eigenvector of $T|_U$ is also a generalized eigenvector of T

$$v \in G(\lambda_j, T) \quad v = v_1 + u \quad \text{where } v_1 \in G(\lambda_j, T) \Rightarrow u \in U$$

$$\Rightarrow u = u_1 + \dots + u_m \quad \text{where } u_i \in G(\lambda_i, T|_U)$$

$$v = v_1 + v_2 + \dots + v_m$$

Generalized eigenvectors corresponding to distinct eigenvalues are linearly independent

all $v_i = 0$ except possibly $v_j \Rightarrow v = v_j \in G(\lambda_j, T|_U)$.

$$G(\lambda_j, T) = G(\lambda_j, T|_U)$$

$$\Rightarrow V = G(\lambda_1, T) \oplus \cdots \oplus G(\lambda_m, T)$$

Theorem: Suppose V is a complex vector space. $T \in L(V)$. Then \exists a basis of V consisting of generalized eigenvectors of T .

Proof: ✓

Def: $T \in L(V)$. The multiplicity of an eigenvalue λ of T is the dim of $G(\lambda, T)$.

Algebraic multiplicity = $\dim G(\lambda, T)$.

Geometric multiplicity = $\dim E(\lambda, T)$

Theorem: Suppose V is a complex vector space $T \in L(V)$. Then the sum of the ^{alg} multiplicities of all eigenvalues of T equals $\dim V$.

Def: Block diagonal matrix.

A block diagonal matrix is a square matrix of the form

$$\begin{bmatrix} A_1 & & \\ & \ddots & \\ 0 & & A_m \end{bmatrix}$$

where A_1, \dots, A_m are square matrices

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

Theorem: V is a complex vector space $T \in L(V)$. Let $\lambda_1, \dots, \lambda_m$ be distinct eigenvalues of T with multiplicities d_1, \dots, d_m . Then \exists a basis of V with respect to which T has a block diagonal form

$$\begin{bmatrix} A_1 & & \\ & \ddots & \\ 0 & & A_m \end{bmatrix} \text{ where } A_j \text{ is } d_j \times d_j \text{ matrix} \quad \begin{bmatrix} \lambda_{1j} & * & & \\ 0 & \ddots & & \\ & & \ddots & \\ & & & \lambda_{mj} \end{bmatrix}$$

Proof:

$$V = G_r(\lambda_1, T) \oplus \dots \oplus G_r(\lambda_m, T)$$

$G_r(\lambda_i, T)$ are T -invariant

$$T \mid_{G_r(\lambda_i, T)}$$

$$(T - \lambda_i I) \mid_{G_r(\lambda_i, T)} \text{ is nilpotent.}$$

The matrix of nilpotent operator are

$$\begin{bmatrix} 0 & * \\ 0 & 0 \end{bmatrix}$$

Uppertriangular with diagonal entries 0!

reflected
 $\rightarrow G_r(\lambda_j, T)$

$$\left\{ M(T - \lambda_j T) = \begin{bmatrix} 0 & * \\ 0 & 0 \end{bmatrix} \right.$$

$$\left. M(T) = \begin{bmatrix} 0 & * \\ 0 & 0 \end{bmatrix} + \lambda_j I = \begin{bmatrix} 0 & * \\ 0 & \lambda_j \end{bmatrix} \right.$$

$$v^1 \in V_2 \\ T(v^1) \in V_2$$

$$\begin{matrix} v^1 \\ T(v^1) \\ T(v^2) \end{matrix} \in V_1$$

$$V = U_1 \oplus U_2 \\ T \in L(V) \\ U_1, U_2 \text{ are } T\text{-invariant}$$

$$M(T) = \begin{bmatrix} * & * & 0 \\ * & * & 0 \\ 0 & 0 & * \end{bmatrix}$$

$$V = V_1 \oplus V_2$$

$\{v_1, v_2\}$
 $\{v_3, v_4\}$
 $\{v_5, v_6\}$
 $\{v_7, v_8\}$
 $\{v_9, v_{10}\}$
 $\{v_{11}, v_{12}\}$
 $\{v_{13}, v_{14}\}$
 $\{v_{15}, v_{16}\}$
 $\{v_{17}, v_{18}\}$
 $\{v_{19}, v_{20}\}$

T is T -invariant
 $\mu(T) = \begin{pmatrix} \mu(T|V_1) & 0 \\ 0 & \mu(T|V_2) \end{pmatrix}$

$$\mu(T) = \begin{bmatrix} \mu(T|_{G_1(\lambda_1, T)}) \\ \mu(T|_{G_2(\lambda_2, T)}) \\ \vdots \\ \mu(T|_{G_n(\lambda_n, T)}) \end{bmatrix}$$

$$A_m = \begin{bmatrix} A_1 & & & \\ & \ddots & & \\ & & A_m & \\ & & & \ddots \end{bmatrix}$$

$$\text{Ex: } T \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} 6z_1 + 3z_2 + 4z_3 \\ 6z_2 + 2z_3 \\ 7z_3 \end{bmatrix}$$

$$M(T) = \begin{bmatrix} 6 & 3 & 4 \\ 0 & 6 & 2 \\ 0 & 0 & 7 \end{bmatrix}$$

$$G_1(6, T) = \text{Span} \left\{ \begin{bmatrix} v_1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} v_2 \\ 0 \\ 3 \end{bmatrix} \right\}$$

$$G_2(T, T) = \text{Span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} \right\}$$

$$M(T, v_1, v_2, v_3) = \begin{bmatrix} 6 & 3 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 7 \end{bmatrix}$$

Def: V is a complex vector space. $T \in L(V)$

$\lambda_1, \dots, \lambda_m$ are eigenvalues of T

d_1, \dots, d_m are alg. multiplicities of T .

Then $\phi(z) = (z - \lambda_1)^{d_1} \cdots (z - \lambda_m)^{d_m}$ is called
characteristic polynomial of T

$$G_2(\lambda, T) = \text{null}(T - \lambda I)^2$$

$$E(\lambda, T) = \text{null}(T - \lambda I)^2 \subseteq \text{null}(T - \lambda I)^2$$

Theorem: V is a complex vector space. $T \in L(V)$.

(1) The characteristic polynomial of T has degree $\dim V$.

(2) The zeros of the characteristic polynomial of T are the eigenvalues of T .

Characteristic polynomial: $T \in L(V)$, $\lambda_1, \dots, \lambda_m$ are eigenvalues
 d_1, \dots, d_m are algebraic multiplicities

$$\rho(z) = (z - \lambda_1)^{d_1} \cdots (z - \lambda_m)^{d_m}$$

- ① $\deg \rho = \dim V$
- ② Zeros of $\rho(z)$ are eigenvalues

Theorem (Cayley-Hamilton Theorem): Suppose V is a complex vector space, $T \in L(V)$. Let $q_V(z)$ be the characteristic polynomial, then $q_V(T) = 0$

Proof: Let $\lambda_1, \dots, \lambda_m$ be eigenvalues of T
 d_1, \dots, d_m be algebraic multiplicities of T

$$V = G_1(\lambda_1, T) \oplus \cdots \oplus G_m(\lambda_m, T)$$

$$q_V(T) = 0.$$

$$(T - \lambda_1)^{d_1} q_V(T) v = 0 \quad \forall v \in V$$

$$v = v_1 + \cdots + v_m$$

$$(T - \lambda_i I) \Big|_{G_r(\lambda_i, T)} \text{ is nilpotent}$$

$$(T - \lambda_1)^{d_1} \cdots (T - \lambda_m)^{d_m} (v_1 + \cdots + v_m) = 0$$

$$(T - \lambda_i I)^{d_i} = 0 \text{ on } G_r(\lambda_i, T) \quad d_i = \dim G_r(\lambda_i, T)$$

$$(T - \lambda_i I)^{d_i} u = 0 \quad \forall u \in G_r(\lambda_i, T)$$

$$(T - \lambda_1)^{d_1} \cdots (T - \lambda_m)^{d_m} (v_1 + \cdots + v_m) = 0 \quad v_i \in G_r(\lambda_i, T)$$

$$(T - \lambda_1)^{d_1} \cdots (T - \lambda_m)^{d_m} v_i = 0 \quad \forall i$$

$$(T - \lambda_1)^{d_1} \cdots (T - \lambda_{i-1})^{d_{i-1}} (T - \lambda_{i+1})^{d_{i+1}} \cdots (T - \lambda_m)^{d_m} (T - \lambda_i)^{d_i} v_i = 0$$

$(T - \lambda_1)^{(T - \lambda_2)}$
 $(\lambda_1 - \lambda_2)^{(T - \lambda_3)}$
 \vdots
 $(T - \lambda_1)^{(T - \lambda_2)^{(T - \lambda_3)}}$
 \vdots
 \vdots

Def: $T \in L(V)$. A monic polynomial is a polynomial whose highest degree coefficient equals 1.

Theorem: $T \in L(V)$. Then \exists a unique monic polynomial $p(z)$ of smallest degree s.t $p(T) = 0$

Proof: $n = \dim V$. Consider the list $1, T, T^2, \dots, T^{n^2}$ is n^2+1 in number $\dim L(V) = n^2$.
 $\Rightarrow 1, T, T^2, \dots, T^{n^2}$ is linearly dependent.

$$\Rightarrow \exists a_0, \dots, a_{n^2} \text{ s.t } a_0 + a_1 T + a_2 T^2 + \dots + a_{n^2} T^{n^2} = 0 \quad \text{not all } a_i = 0.$$

Choose m to be the smallest integer s.t

$1, T, T^2, \dots, T^m$ is linearly dependent.

$\Rightarrow \exists a_0, \dots, a_m \text{ not all zero s.t}$

$$a_0 + a_1 T + \dots + a_m T^m = 0.$$

I find the monic polynomial $p(z)$ to be $\frac{a_0}{a_m} + \frac{a_1}{a_m} z + \dots + \frac{a_{m-1}}{a_m} z^{m-1} + z^m = 0$

($a_m \neq 0$ as m is smallest s.t
 $1, T, \dots, T^m$ is l.d.)

Suppose $\exists q(\tau)$, a monic polynomial of degree m s.t. $q(\tau) = 0$.

Consider $p(\tau) - q(\tau)$ has degree $m-1$.

$$p(p(\tau) - q(\tau))v = 0 \quad \text{as } p(p(\tau)v) = q(\tau)v = 0$$

contradiction to minimality of m .

Def: $T \in L(V)$. Then the minimal polynomial of T is the unique monic polynomial p of smallest degree s.t. $p(T) = 0$.

$T \in L(V)$ $M(T)$ For each m s.t. $M(T^m)$

$$\text{solve } -M(T^2) = a_0 + a_1 M(T)$$

$$\text{if not solve } -M(T^3) = a_0 + a_1 M(T) + a_2 M(T^2)$$

:

& keep going on.

Theorem: $T \in L(r)$. If $f(z)$ is a polynomial. Then $g(T) = 0 \iff g$ is a polynomial multiple of the minimal polynomial

Proof: Suppose $g_T = f \cdot p$ where $p(z)$ is the minimal polynomial.

$$g_T(T) = f(T) p(T) \quad p(T) \neq 0 \quad \Rightarrow g(T) = 0.$$

Conversely suppose $g_T(z)$ is a polynomial s.t $g(T) = 0$.

By division algorithm $\frac{g_T(T)}{p(T)} = q(T) + r(T)$ where $\deg r(T) < \deg p(T)$

$$\Rightarrow r(T) = 0 \quad \& \quad \deg r(T) < \deg p(T)$$

$$\Rightarrow r \equiv 0$$

$$\Rightarrow g(T) = f(T) p(T)$$

Theorem: Suppose V is a complex vector space. $T \in L(V)$. Then the characteristic polynomial of T is a polynomial multiple of the minimal polynomial.

$\xrightarrow{\text{R} = R^T}$ Theorem: $T \in L(V)$. Then the zeros of minimal polynomial of T are precisely the eigenvalues of T .

Proof: $p(z) = a_0 + a_1 z + \dots + a_{m-1} z^{m-1} + z^m$

λ is a zero for $p(z)$

$$\Rightarrow p(z) = (z - \lambda) q(z)$$

$$p(T) = 0 \Rightarrow p(T) = (T - \lambda) q(T)$$

$$\forall v \in V \quad 0 = p(T)v = (T - \lambda) \underbrace{q(T)v}_{\neq 0} \quad \deg q(T) < \deg p(T)$$

$$= q(T)(T - \lambda)v$$

$$\deg q(T) < \deg p(T) \Rightarrow q(T) \neq 0 \Rightarrow q(T)v \neq 0 \text{ for some } v \in V.$$

\Rightarrow For the v , $D = q_1(T)(T-\lambda)v$.

$$\Rightarrow T v = \lambda v$$

$\Rightarrow \lambda$ is an eigenvalue & v is its eigenvector.

Suppose $\lambda \in F$ is an eigenvalue of T . $\Rightarrow T v = \lambda v$ for some v .

$$T^i v = \lambda^i v \quad \forall i$$

$$\begin{aligned} D = P(T)v &= (T^m + a_{m-1}T^{m-1} + \dots + a_1T + a_0)v \\ &= (\lambda^m + a_{m-1}\lambda^{m-1} + \dots + a_1\lambda + a_0)v \end{aligned}$$

$$\Rightarrow \lambda^m + a_{m-1}\lambda^{m-1} + \dots + a_1\lambda + a_0 = 0$$

$\stackrel{\text{def}}{=} p(\lambda) \Rightarrow \lambda$ is a root of the minimal polynomial $p(z)$.

Ex: $V = \mathbb{C}^3$ $T \in \mathcal{L}(V)$. Suppose $\det(T - \lambda I) = 0$, solve to get eigenvalues of T : $\lambda_1, \lambda_2, \lambda_3$.

min poly $(T-\lambda_1)(T-\lambda_2)(T-\lambda_3)$

① characteristic poly is $(T-\lambda_1)(T-\lambda_2)(T-\lambda_3)$

$\lambda_1 = \lambda_2$ $(T-\lambda_1)^2 (T-\lambda_3)$

$\lambda = \lambda_1 = \lambda_3$ $(T-\lambda_1)^3$

$(T-T)^3 (T-T)^0$

Ex: $V = \mathbb{R}^3$ $\gamma_T(T) = 0$ $\gamma_T(z) = (z^2 + 1)(z - 1)$

$\gamma_T(z) = z^2 - 1$

Ex: $V = \mathbb{C}^5$

§ Jordan Canonical Form

Ex: $N \in \mathbb{R}^4$. $N \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix} = \begin{bmatrix} 0 \\ z_1 \\ z_2 \\ z_3 \end{bmatrix}$ $N^4 = 0$.

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

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

$$N^2 v = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

$$N^3 v = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$\{v, Nv, N^2v, N^3v\}$ form a basis for V .

$$Nv = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} = 0v + 1Nv + 0N^2v + 0N^3v$$

$$\mu(N) = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}^T$$

Ex: $N \in \mathcal{L}(C^b)$

$$N \begin{bmatrix} z_1 \\ \vdots \\ z_6 \end{bmatrix} = \begin{bmatrix} 0 \\ z_1 \\ z_2 \\ 0 \\ z_4 \\ z_5 \end{bmatrix}$$

$$N^6 = 0$$

\exists no v s.t. $v, Nv, N^2v, N^3v, \dots, N^5v$ is a basis for V .

§ Jordan Form

Ex: $N \in \mathcal{L}(C\mathbb{P}^4)$

$$N \begin{pmatrix} \bar{z}_1 \\ z_2 \\ z_3 \\ z_4 \end{pmatrix} = \begin{pmatrix} 0 \\ z_1 \\ z_2 \\ \bar{z}_3 \end{pmatrix}$$

$$N^4 = 0$$

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

$$N\varphi = \begin{bmatrix} 0 & 1 \\ 1 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \varphi$$

$$N^2\varphi = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 1 & 0 \\ 0 & 0 \end{bmatrix} \varphi$$

$$N^3\varphi = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ 1 & 0 \end{bmatrix} \varphi$$

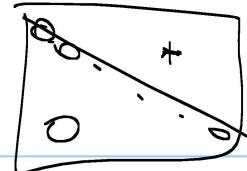
$$\mathcal{B} = [N^3\varphi, N^2\varphi, N\varphi, \varphi]$$

$$\mathcal{M}(N, \mathcal{B}) = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Ex: $N \in \mathcal{L}(\mathbb{F}^6)$

$$N \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \\ z_5 \\ z_6 \end{bmatrix} = \begin{bmatrix} 0 \\ z_1 \\ z_2 \\ 0 \\ z_4 \\ 0 \end{bmatrix}$$

$\rightarrow N \in \mathcal{L}(T)$



$\rightarrow V = G_1(\lambda_1, T) \oplus \dots \oplus G_m(\lambda_m, T)$

There is no v s.t. $N^5v, N^4v, N^3v, N^2v, v$ is a basis for V !!!

choose $v_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}, v_3 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$

$B \{ N^2v_1, Nv_1, v_1, Nv_2, v_2, v_3 \}$ form a basis for V !

$$M(T, B) =$$

$$\left[\begin{array}{ccc|c} 0 & 1 & 0 & \\ 0 & 0 & 1 & \\ 0 & 0 & 0 & \\ \hline 0 & 0 & 0 & \\ 0 & 1 & 0 & \\ 0 & 0 & 0 & \end{array} \right]$$

Theorem: $N \in L(V)$ is nilpotent. Then $\exists v_1, \dots, v_n$ of V & non-negative integers

m_1, \dots, m_n s.t.

(1)

$N^{m_1}v_1, N^{m_1-1}v_1, \dots, N^{m_1}v_1, N^{m_2}v_2, \dots, N^{m_2}v_2, \dots, N^{m_n}v_n, \dots, N^{m_n}v_n$ forms a basis of V .

(2)

$$N^{m_1+1}v_1 = \dots = N^{m_n+1}v_n = 0$$

Proof: By induction on $\dim V$.

If $\dim V=1 \Rightarrow N=0 \Rightarrow$ choose any basis for V & it will satisfy the conclusion.

Assume by induction hypothesis the result is true for vector spaces of $\dim < n$.

N is nilpotent $\Rightarrow N$ is not injective $\Rightarrow \ker N \neq 0 \Rightarrow \dim \ker N \geq 0$.

$$\Rightarrow \dim V = \dim \ker N + \dim \text{range } N \quad \text{rank-nullity}$$

$$\Rightarrow \dim \text{range } N < \dim V.$$

$\text{range } N$ is invariant subspace of V & $\dim \text{range } N < \dim V$.

By induction hypothesis $\exists v_1, \dots, v_n \in \text{range } N$ s.t. $m_1, \dots, m_n \geq 0$ s.t.

(*) $N^{m_1}v_1, \dots, N^{m_n}v_n$ form a basis for range N .
 $\& N^{m_1+1}v_1 = N^{m_2+1}v_2 = \dots = N^{m_n+1}v_n = 0$.

$v_1, \dots, v_n \in \text{range } N \Rightarrow \exists u_i \in V$ s.t. $Nu_i = v_i \quad 1 \leq i \leq n$.

(**)

$$N^{k+1}u_i = N^k v_i$$

Claim: $N^{m_1+1}u_1, \dots, N^{m_n+1}u_n$ is l.i. in V .

Suppose

$$\alpha_{1m_1+1}N^{m_1+1}u_1 + \dots + \alpha_{11}u_1 + \alpha_{2m_2+1}N^{m_2+1}u_2 + \dots + \alpha_{21}u_2 + \dots + \alpha_{nm_n+1}N^{m_n+1}u_n + \dots + \alpha_{n1}u_n = 0$$

Applying N to above -

$$\alpha_{1m_1+1}\cancel{N^{m_1+2}u_1} + \dots + \alpha_{11}Nu_1 + \alpha_{2m_2+1}\cancel{N^{m_2+2}u_2} + \dots + \alpha_{21}Nu_2 + \dots + \alpha_{nm_n+1}\cancel{N^{m_n+2}u_n} + \dots + \alpha_{n1}Nu_n = 0$$

$$\Rightarrow \alpha_{1m_1+1}N^{m_1+1}u_1 + \dots + \alpha_{11}Nu_1 + \alpha_{2m_2+1}N^{m_2+1}u_2 + \dots + \alpha_{21}Nu_2 + \dots + \alpha_{nm_n+1}N^{m_n+1}u_n + \dots + \alpha_{n1}Nu_n = 0$$

$$\Rightarrow \alpha_{1m_1}, \alpha_{1m_1+1}, \dots = \alpha_1 \alpha_{2m_2} = \dots = \alpha_{21} = \dots = \alpha_{nm_n} = \dots = \alpha_{n1} = 0.$$

$$\Rightarrow \alpha_{1m_1+1} N^{m_1+1} u_1 + \dots + \alpha_{nm_n+1} N^{m_n+1} u_n = 0.$$

$$\alpha_{1m_1+1} N^{m_1+1} u_1 + \dots + \alpha_{nm_n+1} N^{m_n+1} u_n = 0.$$

Extend \oplus to a basis $\Rightarrow x_{i,j} = 0$ by \star

by \star
from $Nw_i = 0$

$N^{m_1} u_1, \dots, u_1, N^{m_2} u_2, \dots, u_2, \dots, N^{m_n} u_n, \dots, u_n, w_1, \dots, w_p$ of V .

$Nw_j \in \text{range } N \Rightarrow Nw_j$ is written as a linear combination of vectors \oplus

$\Rightarrow \exists$ some $x_j \in \text{range } N$ s.t. $N x_j = N w_j$

$$\Rightarrow N(w_j - x_j) = 0$$

Set $w_j - x_j = u_{n+j}$

$N^{m_1} u_1, \dots, u_1, N^{m_2} u_2, \dots, u_2, \dots, N^{m_n} u_n, \dots, u_n, u_{n+1}, \dots, u_{n+p}$

is linearly independent and satisfies \oplus

$w_1 - x_1, \dots, w_p - x_p$

Def: Jordan Basis

$T \in \mathcal{L}(V)$. A basis of V is called a Jordan basis for T if w.r.t this basis the matrix of T is a block diagonal form

$$\begin{bmatrix} A_1 & & & \\ & \ddots & & 0 \\ & & \ddots & \\ 0 & & & A_p \end{bmatrix}$$

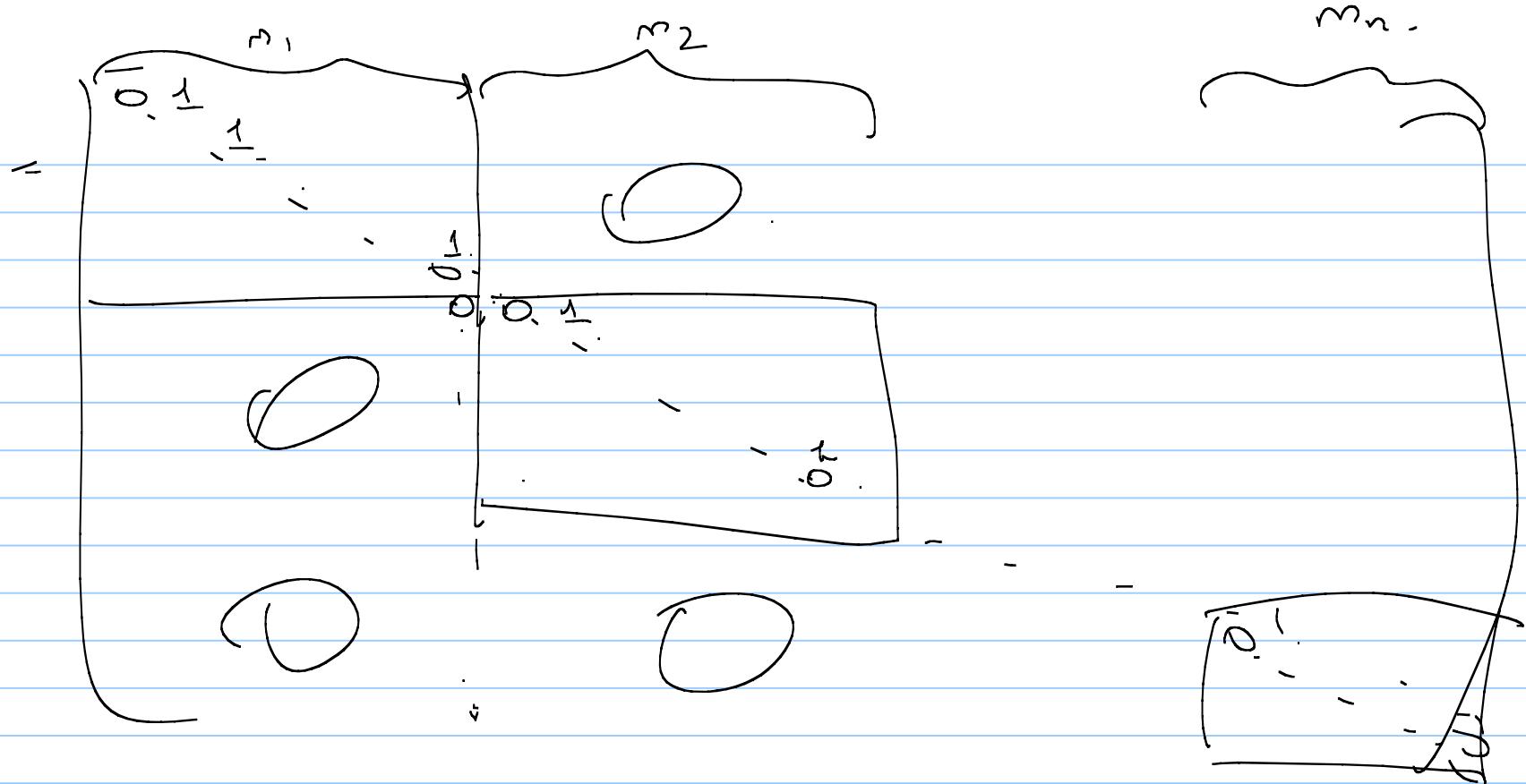
$$A_i = \begin{bmatrix} \lambda_i & 1 & & & 0 \\ 0 & \lambda_i & \ddots & & \\ & & \ddots & \ddots & \\ & & & 0 & \lambda_i \end{bmatrix}$$

Theorem: V is a complex vector space. $T \in \mathcal{L}(V)$. then \exists $\xrightarrow{\text{Jordan}}$ basis for T .

Proof: $N \in \mathcal{L}(V)$ be nilpotent.

$\Rightarrow \exists v_1, \dots, v_n \in V$ s.t. $N^{m_1} v_1, \dots, N^{m_2} v_2, \dots, N^{m_n} v_n$ form basis for V

$$\mu(N) = \begin{bmatrix} 0 & 1 & 0 & & & \\ 0 & 0 & 1 & & & \\ \vdots & \vdots & \ddots & \ddots & & \\ 0 & 0 & 0 & 0 & 1 & \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$



$$V = G(\lambda_1, T) \oplus \dots \oplus G(\lambda_m, T)$$

$G(\lambda_i, T)$ are invariant

$$(T - \lambda_i) \Big|_{G(\lambda_i, T)} \text{ is nilpotent.}$$

From what we just proved for nilpotent operators

\Rightarrow there is a basis w.r.t which

$\mu(T - \lambda_i)$ is a block diagonal matrix with 0's on diagonal

w.r.t to this basis

$\mu(T \Big|_{G(\lambda_i, T)})$ is a block diagonal matrix with λ_i on diagonal

$$\begin{bmatrix} \bar{A}_1 & & \\ & \ddots & \\ & & \bar{A}_p \end{bmatrix}.$$

$$\bar{A}_i = \begin{bmatrix} \lambda_i^1 & & & \\ & \ddots & & 0 \\ & & \ddots & \\ 0 & & & \lambda_i^1 \end{bmatrix}$$

Put such basis of $G_r(\lambda, \tau)$ together to form a basis for V . w.r.t which the
 $\mu(\tau)$ has the required form.

$$\textcircled{1} \quad V = G(\lambda_1, T) \oplus \dots \oplus G(\lambda_m, T) + \text{Ed}(V) \quad V \text{ is a complex vector space}$$

\textcircled{2} $G(\lambda_i, T)$ are T -invariant

\textcircled{3} $(T - \lambda_i I) \Big|_{G(\lambda_i, T)}$ are nilpotent $\xrightarrow{\textcircled{4}} \exists$ basis $N_{v_1}^{m_1}, \dots, N_{v_1, v_2}^{m_2}, \dots, N_{v_1, \dots, v_n}^{m_n}$ s.t.

$$\begin{bmatrix} 0 & 1 & & \\ & 0 & \ddots & \\ & & \ddots & 1 \\ & & & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & & & \\ & \ddots & & \\ & & 0 & \\ & & & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & & & \\ & \ddots & & \\ & & 0 & \\ & & & 0 \end{bmatrix}$$

$$N = T - \lambda_i I.$$

$$N^{m_i+1} v_i = 0.$$

\textcircled{5} Jordan form
 $T \in \mathcal{L}(V)$.

$$\begin{pmatrix} \lambda_1 & & & \\ & \ddots & & \\ & & \lambda_1 & \\ & & & \ddots & \lambda_m & \\ & & & & \ddots & \lambda_m \end{pmatrix}$$

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

$$\underline{\text{Ex}}: \quad A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 3 \end{bmatrix} \quad T: \mathbb{C}^3 \rightarrow \mathbb{C}^3 \quad \lambda = 1: \quad \begin{aligned} \text{null}(T - \lambda I) \\ \text{null}(T - \lambda I)^2 = 0 \\ \text{null}(T - \lambda I)^3 = 0 \end{aligned}$$

$$\det A - \lambda I_3 = (\lambda - 1)^2 (\lambda - 3) = 0 \Rightarrow \lambda = 1, 3.$$

$$u \in (\mathbb{A} - \lambda I)$$

$$\lambda = 1 : \text{null}(A - I)$$

$$\Theta_1 = \begin{bmatrix} 1 \\ 2 \\ 1/2 \end{bmatrix}$$

$$G_1(1, T) = \delta / \alpha < \left[\frac{1}{\frac{3}{2}} \right], ? \quad >$$

Compute null($A - I$)^T

$$D_2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

No more generalized eigenvalues

Can get one by solving $(A - \lambda I)^2 v_3 = 0$. $(A - \lambda I)^2 v_3 \neq 0$.

$$V = \text{Span} \left\langle \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\rangle \oplus \text{Span} \left\langle \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\rangle$$

$G_2(1, T)$ $G_3(0, T)$

$$\text{Ex: } A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & 2 \\ 0 & -1 & 1 \end{bmatrix}$$

$\det(A - \lambda I)$. $\lambda = 1$ is the only eigenvalue.

$$\text{null}(A - \lambda I) \quad v_1 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix} \text{ eigenvalue.}$$

$$(A - \lambda I)^2 = \begin{bmatrix} 2 & 0 & 4 \\ 0 & 0 & 0 \\ -1 & 0 & 2 \end{bmatrix}.$$

$$\text{null}(A - \lambda I)^2 = \left\langle \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, v_2 \right\rangle.$$

$$(A - \lambda I)^3 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\text{null}(A - \lambda I)^3 = V.$$

Expand v_1, v_2 to a basis of V to get v_3 .

v_3 will be a generalized eigenvector as
 $\text{null}(A - \lambda I)^3 = V$.

$$\text{Ex: } A = \begin{bmatrix} 1 & 2 & 0 & 1 \\ 1 & 1 & 2 & 0 \\ 0 & -1 & 1 & 1 \end{bmatrix}$$

$\lambda=1$ is eigenvalue: $v_1 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$ is eigenvector. $\Rightarrow (A - I)v_1 = 0$

$$\text{Solve } (A - I)x = v_1$$

$$\text{If it has a solution } v_2, \Rightarrow (A - I)v_2 = v_1$$

$$(A - I)^2 v_2 = (A - I)v_1 = 0$$

$\Rightarrow v_2$ is a generalized eigenvector.

$$\text{Solve } (A - I)x = v_2$$

$$\text{If it has a solution } v_3,$$

$$(A - I)v_3 = v_2$$

$$(A - I)^3 v_3 = (A - I)^2 v_2 = 0$$

v_3 is a generalized eigenvector

A is a $n \times n$ matrix. ω is a generalized eigenvector of A .

$$(A - \lambda I)^j \omega = 0$$

$i < j$

$$(A - \lambda I)^{j-i} (A - \lambda I)^i \omega = 0 \Rightarrow (A - \lambda I)^i \omega \text{ is also a generalized eigenvector.}$$

If we choose j to be the smallest integer s.t. $(A - \lambda I)^j \omega = 0$

$1 \leq i < j$ $(A - \lambda I)^i \omega$ is a generalized eigenvector.

Ex: For the previous example

$$A - \lambda I = \begin{bmatrix} 0 & 2 & 0 \\ 1 & 0 & 2 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\omega_1 = \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix}$$

$$(A - \lambda I) \omega_1 = \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix}$$

$$(A - \lambda I)^2 \omega_1 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$$

$$\Rightarrow (A - \lambda I)^3 \omega_1 = 0$$
$$(A - \lambda I)^2 \omega_1 \neq 0$$

Ex: $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 2 \end{bmatrix}$

$\det(A - \lambda I) = (\lambda - 2)(\lambda - 1)^2 = 0 \Rightarrow \lambda = 1, 2.$

$\lambda = 2$: Eigenvalue $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = v_1$.

$\lambda = 1$ eigenvalue $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = v_2$

Generalized eigenvectors.

$$(A - \lambda I) \mathbf{x} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = v_2$$

Solve
to get $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = v_3$.

Q8

Ex: $A = \begin{bmatrix} 1 & -2 & 0 \\ 2 & -3 & 0 \\ -1 & 1 & -1 \end{bmatrix}$

$$\det(A - \lambda I) = (\lambda + 1)^3 = 0 \quad \lambda = -1,$$

$$E(-1, T) = \text{Span}\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}\right)$$

$$(A + I)x = v_1$$

$$(A + I)x = v_2$$

inconsistent!!

$$(A + I)^2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Expand $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ to a basis $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$.

third generalized eigenvector

$$\text{Ex: } A = \begin{bmatrix} 0 & 1 & -1 & -1 \\ 0 & 0 & 0 & 0 \\ 0 & -1 & 2 & 2 \\ 0 & 1 & -2 & -2 \end{bmatrix}.$$

The only eigenvalue is $\lambda=0$ of algebraic multiplicity 4.

$$E(0, T) = \text{Span} \left\langle \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix} \right\rangle = \text{eigen} \text{span of } \lambda=0. = \text{ml}(A)$$

$$A \underline{x} = \underline{v}_1$$

solve

$$\underline{v}_3 = \begin{bmatrix} 0 \\ -1 \\ -1 \\ 0 \end{bmatrix}.$$

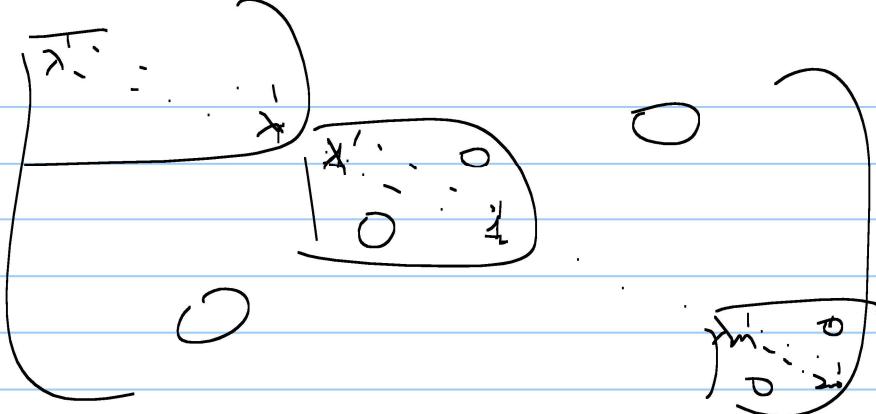
$$A \underline{x} = \underline{v}_2$$

solve.

$$\underline{v}_4 = \begin{bmatrix} 0 \\ 2 \\ 1 \\ 0 \end{bmatrix}.$$

$\underline{v}_1, \underline{v}_2, \underline{v}_3, \underline{v}_4$ form a basis for $G_2(0, T) = V$.

\exists a basis $N^{m_1}v_1, \dots, v_i, N^{n_2}v_2, \dots, v_j, \dots, N^{m_n}v_n, \dots, v_n$ form a basis for V .



$\frac{1}{2}$	$\frac{1}{2}$	$\textcircled{0}$
0	1	
2	1	
0	2	

superdiagonal
is zero!

Jordan Canonical Form

Ex: $A = \begin{bmatrix} 4 & 0 & 1 \\ 2 & 3 & 2 \\ 0 & 0 & 4 \end{bmatrix}$

eigenvalues $\lambda = 3, 5$

$$\lambda = 5 \quad E(5, T) = \text{Span} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

$$\lambda = 3 \quad E(3, T) = \text{Span} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

\exists a basis $\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ consisting of eigenvectors of A .

$$A \sim \begin{bmatrix} 5 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix} = \text{Jordan form.}$$

Diagonalizable

Ex: $A = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$\lambda=1$ is the only value

$$E(1, A) = \text{Span} \left\langle \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right\rangle$$

choose v_3 s.t. v_1, v_2, v_3 form a basis

choose $v_1 = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$ $v_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$

$(A - I)v_2 = v_2 \Rightarrow v_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$

Jordan form = $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$

$N = A - I$

$N_{v_1}^{m_1}, \dots, v_r, \dots, N_{v_n}^{m_n}$

$$(A - I)^2 = \begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}^2 = 0$$

generalized eigenvectors
 $\text{Span}\{v_1, v_2, v_3\} = G(1, 1)$

$$\underline{\text{Ex}}: A = \begin{bmatrix} -1 & -1 & 0 \\ 0 & -1 & -2 \\ 0 & 0 & -1 \end{bmatrix}$$

eigenvalue = -1.

$$E(-1, A) = \text{Span} \left\langle \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\rangle$$

$$(A + I)x = v_1 \text{ solve to get } v_2 = \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix}$$

$$G(1, T) = \text{Span} \langle v_1, v_2, v_3 \rangle$$

$$(A + i)x = v_3 \text{ solve to get } v_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Jordan form

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

§ Real vectorspace.

Def: V is a real vectorspace

$$\{(u, v); u, v \in V\}$$

① The complexification of V , denoted by V_C is $V \times V$. The elements of $V \times V$ is denoted by $u+iw$ where $u, w \in V$.

② Addition on V_C :

$$(u_1 + i(v_1)) + (u_2 + i(v_2)) = (u_1 + u_2) + i(v_1 + v_2)$$

③ ~~Complex~~ scalar multiplication on V_C

$$\underbrace{(a+ib)}_{\in \mathbb{C}} \cdot \underbrace{(u+iw)}_{u, w \in V} = (au - bw) + i(av + bu) \in V \times V$$

$a, b \in \mathbb{R}$

Remark: $V_{\mathbb{C}}$ is a complex vector space. (check)

Theorem: V is a real vector space.

① v_1, \dots, v_n is a basis for $V \Rightarrow v_1, \dots, v_n$ is a basis for $V_{\mathbb{C}}$.

② $\dim_{\mathbb{R}} V_{\mathbb{C}} = \dim_{\mathbb{R}} V$.

Proof: $u + iv \in V_{\mathbb{C}}$ $u, v \in V = \text{Span}_{\mathbb{R}}\{v_1, \dots, v_n\}$

$$u = \sum g_j v_j \quad v = \sum d_j v_j$$

$$u + iv = \sum g_j v_j + i \sum d_j v_j = \sum_j (g_j + i d_j) v_j$$

$$V_{\mathbb{C}} = \text{Span}_{\mathbb{C}}\{v_1, \dots, v_n\}$$

$$\lambda_1 v_1 + \dots + \lambda_n v_n = 0 \quad \lambda_i \in \mathbb{C}$$

$$\lambda_j = \operatorname{Re} \lambda_j + i \operatorname{Im} \lambda_j$$

$$\left(\operatorname{Re} \lambda_1 v_1 + \dots + \operatorname{Re} \lambda_n v_n \right) + i \left(\operatorname{Im} \lambda_1 v_1 + \dots + \operatorname{Im} \lambda_n v_n \right) = 0$$

\Downarrow " \Downarrow

$$\operatorname{Re} \lambda_j \in \mathbb{R} \qquad \qquad \operatorname{Im} \lambda_j \in \mathbb{R}$$

$$\Rightarrow \operatorname{Re} \lambda_j = 0$$

$$\operatorname{Im} \lambda_j = 0$$

$$\Rightarrow \lambda_j = 0$$

$\Rightarrow v_1, \dots, v_n$ is linear in $V_{\mathbb{C}}$.

Def: Complexification of $T \in \mathcal{L}(V)$, V a real vector space.

$$T_{\mathbb{C}} \in \mathcal{L}(V_{\mathbb{C}})$$

$$T_{\mathbb{C}}(u + i v) = Tu + iTv \quad u, v \in V$$

Remark: $T_{\mathbb{C}} \in \mathcal{L}(V_{\mathbb{C}})$ addition ✓

scalar mult-

$$T_{\mathbb{C}}(\lambda(u + iv)) = \lambda T_{\mathbb{C}}(u + iv) \quad \lambda \in \mathbb{C}$$

$$\lambda = a + bi$$

$$T_{\mathbb{C}}((a+bi)(u+iv)) = (a+bi)T_{\mathbb{C}}(u+iv)$$

Ex: $\mathbb{R}^n \rightarrow \mathbb{C}^n$
Complexification

$T \in L(\mathbb{R}^n)$ $M(T)$ is $n \times n$ matrix with real entries

$T_{\mathbb{C}} \in L(\mathbb{C}^n)$ $M(T_{\mathbb{C}})$ is also $n \times n$ matrix with complex entries

$L(\mathbb{C}^n)$.
 $\begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix}$.
 $\in L(\mathbb{R}^n)$

Theorem: V is a real vector space. $\text{v... } v_n$ is a basis for V . $T \in L(V)$.

$$M(T) = M(T_{\mathbb{C}})$$

Theorem: Every operator on a non-zero f.d. V 's has an invariant subspace of dim 1 or 2.

Proof: If V is a complex vector space, use the eigenspace \checkmark

If V is a real vector space: $T \in L(V)$.

$T_{\mathbb{C}} \in L(V_{\mathbb{C}}) \Rightarrow T_{\mathbb{C}}$ has an eigenvalue $\lambda = a + bi$

$\Rightarrow \exists$ an eigenvalue λ of T (not both u, v zero) s.t. $T_C((u+v)) = \lambda(u+v)$

$$T(u) + i T(v) = (au - bv) + i(au + bv)$$

$$T(u) = au - bv$$

$$T(v) = av + bu$$

$$U = \text{Span}\langle u, v \rangle \quad T(U) \subseteq U.$$

↓
invariant subspace of dim 1 or 2.

Theorem: Via a real vector space. Then the minimal polynomial of T is the same as the min poly of T_C .

Proof: $P \in$



V is a real vector space $\Rightarrow V_{\mathbb{C}}$ is a complex vectorspace

$$V \times V = \{u + iv \mid u, v \in V\}$$

$$T \in \mathcal{L}(V) \Rightarrow T_{\mathbb{C}} \in \mathcal{L}(V_{\mathbb{C}}) \quad T_{\mathbb{C}}(u + iv) = T(u) + iT(v)$$

§ Minimum Polynomial

Theorem: V is a real vector space & $T \in \mathcal{L}(V)$. Then the minimum polynomial of $T_{\mathbb{C}}$ equals the minimum polynomial of T .

Proof: $p(x) \in \mathcal{P}(\mathbb{R})$ is a minimum polynomial of T .

$$T_{\mathbb{C}}^n(u + iv) = T_u^n + iT_v^n \quad (\text{check!})$$

$$p(T) = 0 \Rightarrow p(T_{\mathbb{C}}) = 0.$$

$$T_{\mathbb{C}}(u + iv) = T(u) + iT(v)$$

$$\begin{aligned} T_{\mathbb{C}}^2(u + iv) &= T_{\mathbb{C}}(T(u) + iT(v)) \\ &= T^2u + iT^2v \end{aligned}$$

$$P(T)(u) = 0 \quad \forall u \in V. \quad p(T_C)(u+iw) = \overset{\text{"}}{p(T)}(u) + i \overset{\text{"}}{p(T)}(w) = 0$$

Choose $q(x) \in \mathcal{P}(C)$ s.t. $q(T_C) = 0 \quad T_C \in \mathcal{L}(V_C)$

$$T_C^n + a_{n-1} T_C^{n-1} + \dots + a_0 = 0 \quad a_i \in C$$

$$a_i = \operatorname{Re}(a_i) + i \operatorname{Im}(a_i)$$

$$\left(\overset{\text{"}}{T_C^n} + \operatorname{Re}(a_n) \overset{\text{"}}{T_C^{n-1}} + \dots + \operatorname{Re}(a_0) \right) + i \left(\overset{\text{"}}{\operatorname{Im} a_n T^{n-1}} + \dots + \overset{\text{"}}{\operatorname{Im} a_0} \right) = 0$$

$$\Rightarrow T^n + \operatorname{Re}(a_n) T^{n-1} + \dots + \operatorname{Re} a_0 = 0$$

$$n \geq \deg p(x)$$

$\Rightarrow p(x)$ is the minimum polynomial for T_C .

Eigenvalues of T_C

V is a real vector space. $T \in L(V)$. $\lambda \in \mathbb{R}$. λ is an eigenvalue for $T_C \Leftrightarrow \lambda$ is an eigenvalue for T .

Proof: Suppose λ is an eigenvalue for T . $\Rightarrow \exists \overset{\circ}{v} \in V$ s.t. $T(v) = \lambda v$.

$$\in V_C$$

↓

$$T_C(v) = \lambda v.$$

↓
 λ is an eigenvalue for T_C .

Suppose $\lambda \in \mathbb{R}$, λ is an eigenvalue for $T_C \Rightarrow \exists u + iv \in V_C$ s.t.

$$T_C(u + iv) = \lambda(u + iv).$$

$$\overset{||}{T(u) + i T(v)} \quad \overset{||}{\lambda u + i \lambda v}.$$

$\lambda \in \mathbb{R}$

$$\Rightarrow Tu = \lambda u - T(v) = \lambda u.$$

Theorem: Suppose V is a real vectorspace. $T \in L(V)$. $\lambda \in \mathbb{C}$. j is an integer. $u, v \in V$. Then

$$(T_C - \lambda I)^j (u + iv) = 0 \iff (T_C - \bar{\lambda} I)^j (u - iv) = 0$$

Proof: $j=0$ ✓

$$j > 0 \text{ if } (T_C - \lambda I)^j (u + iv) = 0.$$

$$T_C(u + iv) = T_u + iT_v$$

$$(T_C - \lambda I)^{j+1} (T_C - \bar{\lambda} I)(u - iv) = 0 \quad \lambda = a + bi.$$

$$(T_C - \lambda I)^j ((T_u - au + bv) + i(T_v - av - bu)) = 0$$

$$\text{By induction } (T_C - \bar{\lambda} I)^{j+1} ((T_u - au + bv) - i(T_v - av - bu)) = 0$$

$$\Rightarrow (T_C - \bar{\lambda} I)^j (u - iv) = 0$$

$$\begin{aligned} T_C(u - iv) \\ = T_u - iT_v. \end{aligned}$$

Theorem: V is a real vector space. $T \in L(V)$. $\lambda \in \mathbb{C}$. Then λ is an eigenvalue of $T_C \Leftrightarrow \bar{\lambda}$ is an eigenvalue of $T_{\bar{C}}$

Theorem: V is a real vector space $T \in L(V)$. $\lambda \in \mathbb{C}$ is an eigenvalue of T_C . Then multiplicity of λ as an eigenvalue of C = multiplicity of $\bar{\lambda}$ as an eigenvalue of $T_{\bar{C}}$.

Theorem: Every operator on odd-dimensional real vector space has an eigenvalue.

Proof: Every odd degree polynomial has a real ~~\pm~~ root.

§ Characteristic Polynomial

Theorem: V is a real vector space. $T \in L(V)$. Then the coefficients of the characteristic polynomial of T_C are all real.

Proof: λ is a non-real eigenvalue of T_C then $\bar{\lambda}$ is also an eigenvalue for T_C .

Characteristic polynomial for $T_{\mathbb{C}}$ = $(z - \lambda_1)^{m_1} \cdots (z - \lambda_m)^{m_m} (z - \alpha_1)^{a_1} (z - \bar{\alpha}_1)^{\bar{a}_1} \cdots (z - \alpha_b)^{a_b} (z - \bar{\alpha}_b)^{\bar{a}_b}$

λ_i real.

$\alpha_i \in \mathbb{C} \setminus \mathbb{R}$.

$$(z - \alpha_1)^{a_1} (z - \bar{\alpha}_1)^{\bar{a}_1} = (z^2 - \underbrace{2 \operatorname{Re} \alpha_1 z}_{\in \mathbb{R}} + \underbrace{|\alpha_1|^2}_{\in \mathbb{R}})^{a_1} \in \mathcal{P}(\mathbb{R})$$

\Rightarrow characteristic polynomial has real coefficients.

Def: V is a real vector space. The characteristic polynomial of T is the characteristic polynomial of $T_{\mathbb{C}}$.

Theorem: V is a real vector space. $T \in L(V)$.

- (a) The coefficients for the characteristic polynomial are all real
- (b) The characteristic polynomial has degree $\dim V$ ($\dim V_{\mathbb{C}} = \dim V$)
- (c) The eigenvalues of T are precisely the real eigenvalues of $T_{\mathbb{C}}$.
" the real roots of the characteristic polynomial.

Theorem (Cayley Hamilton) : $T \in L(V)$. Let φ denote the characteristic polynomial of T . Then $\varphi(T) = 0$

Theorem : $T \in L(V)$. Then

(1) degree of minimum polynomial of T is at most $\dim V$.

(2) the characteristic poly of T is a polynomial multiple of the minimum polynomial

$$\lambda \in \mathbb{C} - V \cdot S$$

$$\text{char } (z-\lambda_1)^{n_1} \cdots (z-\lambda_m)^{n_m}$$

λ_i are eigenvalues

$$V \in (\mathbb{R} - V \cdot S)$$

$$(z-\lambda_1)^{n_1} \cdots (z-\lambda_m)^{n_m}$$

λ_1 is an eigenvalue

Spectral Theorem for normal operators on real vector spaces.

Theorem: V is a real vector space. $T \in L(V)$. Then the following are equivalent

(1) T is normal

(2) \exists an orthonormal basis for V w.r.t which T has a block diagonal matrix s.t each block is a 1×1 matrix or a 2×2 matrix of the form

$$\begin{bmatrix} a & -b \\ b & a \end{bmatrix} \quad b > 0$$