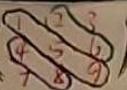


1. let $A \in M_n(F)$ 

- main diagonal $\{a_{ii}\}$, main sub-diagonal $\{a_{i,i-1} | i=2, \dots, n\}$, main super-diagonal $\{a_{i+1,i} | i=2, \dots, n\}$

Date trace ! = $\sum a_{ii}$

let $A, B \in M_n(F)$, $r, s \in F$, $\text{tr}(rA+sB) = r\text{tr}(A)+s\text{tr}(B)$ (linearity)

(commutative) $\text{tr}(AB) = \text{tr}(BA)$ $\sum_{i,j} (AB)_{ij} = \sum_i \sum_k A_{ik} B_{kj} = \sum_k \sum_i B_{ki} A_{ik} = \sum_k (BA)_{kk}$ OR by [15] P. 6 to [24] II. 1

$\text{tr}(ABC) = \text{tr}(BCA)$ (cyclic) $\sum_t (ABC)_{kt} = \sum_t \sum_k A_{tk} (BC)_{kt} = \sum_{t,k} A_{tk} \sum_{j,t} B_{kj} C_{jt}$
 $\neg \text{tr}(AB) \neq \text{tr}(A) + \text{tr}(B)$ $= \sum_t \sum_j B_{kj} C_{jt} A_{tk} = \sum_k \sum_t B_{kj} C_{jt} A_{tk} = \sum_k (BCA)_{kk}$

$\text{tr}(A^T) = \text{tr}(A)$

$\text{tr}(A^T B) = \sum_{i,j} A_{ij} B_{ji}$, $\text{tr}(u \otimes v) = u^T v$ $\text{tr}(\underbrace{u \otimes v}_{=: u \cdot v})$ EV

2. 对 Vec Space V 的 nonempty subset S . $\text{span}(S) \triangleq L.$ comb. of vectors in S , $\text{span}(\emptyset) = \{0\}$

Affirmation: $\text{span}(S)$ 为 V 的子空间, P. 11. 例 1 说明, 由生成元的线性组合构成的集合是闭合的, 即若 $x, y \in \text{span}(S)$, 则 $x+y \in \text{span}(S)$.

T_1 若 $S \subseteq V$, $\text{span}(S) \triangleq V$ 的子空间; 反之, 若 V 的子空间 W 含有 S , 也含 $\text{span}(S)$.

\square i) 若 $S = \emptyset$, $\text{span}(S) = \emptyset$ 任何 vec space 为 trivial subspace; 若 $S \neq \emptyset$, $0 \in \text{span}(S)$, $\forall x \in V$, $x = a_1 v_1 + \dots + a_n v_n$, $a_i \in F$

ii) BP: $s_1, \dots, s_n \in W$, 对 $a_i \in F$, $a_1 s_1 + \dots + a_n s_n \in W$ 由 induction on n. 易证 $x+y, x \in \text{span}(S) \Rightarrow$

3. DEF $S \subseteq$ vector space V , S generates / spans V iff $\text{span}(S) = V$

eg. 1. $\{(1,0), (1,1), (0,1)\}$ spans \mathbb{R}^2 $\exists x \in V \forall x = (a,b) \in \mathbb{R}^2$, $\begin{cases} a = x(1,0) + y(0,1) \\ b = x(0,1) + y(1,0) \end{cases}$

T_2 : $W \subseteq V \Leftrightarrow \text{span}(W) = W$

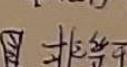
$\square \Rightarrow$, $\text{span}(W) \subseteq W$: $\forall w \in W$, $\text{span}(W) \subseteq W$; $\text{span}(W) \supseteq W$: let $w_i \in W$, $w_i = a_1 w_1 + \dots + a_m w_m \in \text{span}(W)$

$\Leftarrow \text{span}(\emptyset) = \{0\}$, $\therefore W$ nonempty: $\forall w_1, w_2 \in W$, $w_1 + w_2 \in W$ (L. comb. $\neq \emptyset$); scalar multiplication trivial

4. P_i) (span 传递) $S_1, S_2 \subseteq V$ 的子集, $S_1 \subseteq S_2 \Rightarrow \text{span}(S_1) \subseteq \text{span}(S_2)$; 若 $\text{span}(S_1) = V$, $\text{span}(S_2) = V$

i) $S_1, S_2 \subseteq V$, $\text{span}(S_1 \cup S_2) = \text{span}(S_1) + \text{span}(S_2)$

$\square \Leftarrow$: $e \in \text{LHS} \Rightarrow e = \sum_{i=1}^m a_i s_i + \sum_{j=1}^n b_j t_j \in \text{span}(S_1) + \text{span}(S_2) \Rightarrow e \in \text{RHS}$

iii) $S_1, S_2 \subseteq V$, $\text{span}(S_1 \cap S_2) \subseteq \text{span}(S_1) \cap \text{span}(S_2)$ 

$\square e \in \text{LHS} \Rightarrow e = \sum a_i s_i \in \text{span}(S_1) \cap \text{span}(S_2) \Rightarrow e \in \text{span}(S_1) \cap \text{span}(S_2) \Rightarrow e \in \text{RHS}$

\checkmark : $S_1 = \{(1,0)(1,1)\}, S_2 = \{(0,1)(1,0)\}$, LHS = $\text{span}(\emptyset) = \{(0,0)\}$, RHS = $\text{span}(\emptyset) = \{(0,0)\}$

iv) 对 $S \subseteq V$, 若 S L. Ind. 2) $\forall s \in \text{span}(S)$, s 由 S unique 表示

\square let $s = \sum a_i s_i$, $a_i, s_i \in F$, $\sum a_i s_i = 0$, 由 L. Ind. $a_i \cdot b_i = 0, \forall i$
 $\therefore a_i = b_i$

v) (若 $\exists n$) $\forall s \in \text{span}(v_1 \dots v_n)$ 存在 $\lambda_1, \dots, \lambda_n \in F$ 使得 $s = \lambda_1 v_1 + \dots + \lambda_n v_n$

证 1. $x^T y \cdot y^T x = \text{tr}(x x^T y y^T) = \sum_{i,j} x_i x_j \cdot y_i y_j$

\square 令 $u = x$, $v = y^T x$ 由 1. $u^T v = \text{tr}(x x^T y y^T) = \sum_{i,j} (x x^T)_{ij} (y y^T)_{ij}$

Date.

$$\text{例 1} \det \begin{pmatrix} 1 & 2 & 3 & \cdots & n \\ 2 & 3 & 4 & \cdots & n-1 \\ 3 & 4 & 5 & \cdots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ n & n-1 & n-2 & \cdots & n-n \end{pmatrix},$$

□ note 各行之和相等，各列全加于 col 1, 提出 col 1: $= \sum_{i=1}^n i \begin{vmatrix} 1 & 2 & 3 & \cdots & n \\ 2 & 3 & 4 & \cdots & n-1 \\ 3 & 4 & 5 & \cdots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ n & n-1 & n-2 & \cdots & n-n \end{vmatrix}$ note 第 1 行扣差 1, 依上 -19
 $\times (-1)$ 因为第 1 行

$$= \sum_i \begin{vmatrix} 1 & \cdots & n \\ 0 & 1 & \cdots & n-1 \\ 0 & 1 & \cdots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 1 & \cdots & 1 \end{vmatrix} \quad \text{col 1 展开, } = \sum_i \begin{vmatrix} 1 & \cdots & n-1 \\ 0 & 1 & \cdots & 1 \\ 0 & 1 & \cdots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 1 & \cdots & 1 \end{vmatrix} \quad \left\{ \begin{array}{l} n-1 \text{ 行各数之和} = -1 \\ \text{倒数一行减去, 其余全加于 col 1} \end{array} \right. = \sum_i \begin{vmatrix} 1 & \cdots & n-1 \\ -1 & 1 & \cdots & 1 \\ -1 & 1 & \cdots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ -1 & 1 & \cdots & 1 \end{vmatrix}$$

$$\text{用第 } -1 \text{ 行 } -1 \text{ 倍加其全行} \sum_i \begin{vmatrix} 1 & \cdots & n-1 \\ -1 & 1 & \cdots & 1 \\ -1 & 1 & \cdots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ -1 & 1 & \cdots & 1 \end{vmatrix} = \sum_i (-1)^{n-1+i} \begin{vmatrix} 0 & \cdots & 1 \\ -1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ -1 & \cdots & 0 \end{vmatrix} = \frac{n(n+1)}{2} \cdot (-1)^{\frac{n+1}{2}} \cdot \begin{bmatrix} 1 & \cdots & \frac{(n+2)(n+1)}{2} \\ 0 & \cdots & (-1)^{\frac{n+1}{2}} \end{bmatrix}$$

$$\text{例 2} \begin{vmatrix} a_1 & a_2 & \cdots & a_n \\ b_1 & 0 & & \\ b_2 & & 0 & \\ \vdots & & \vdots & \ddots \\ b_n & & 0 & 1 \end{vmatrix} \quad \text{逐列提 } b_1 \quad \begin{vmatrix} * & a_2 & \cdots & a_n \\ 0 & 1 & & \\ 0 & & \ddots & \\ 0 & & & 1 \end{vmatrix} \quad * = a_1 - a_2 \cdot b_2 - \cdots - b_n \cdot a_1 \\ = a_1 - \sum_{i=2}^n a_i \cdot b_i$$

对各列去掉 col 1

$$\text{例 3} \begin{vmatrix} a_1+b_1 & a_1+b_2 & \cdots & a_1+b_n \\ a_2+b_1 & a_2+b_2 & \cdots & a_2+b_n \\ \vdots & \vdots & \ddots & \vdots \\ a_n+b_1 & a_n+b_2 & \cdots & a_n+b_n \end{vmatrix} = \begin{vmatrix} a_1 & a_1+b_2 & \cdots & a_1+b_n \\ a_2 & b_1 & a_2+b_2 & \cdots & a_2+b_n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_n & b_1 & a_n+b_2 & \cdots & a_n+b_n \end{vmatrix} + \begin{vmatrix} b_1 & a_1+b_2 & \cdots & a_1+b_n \\ b_2 & b_1 & a_2+b_2 & \cdots & a_2+b_n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ b_n & b_1 & a_n+b_2 & \cdots & a_n+b_n \end{math}$$

$$\text{例 4} \begin{vmatrix} a_1 & b_1 & 0 & \cdots & 0 \\ c_1 & a_2 & b_2 & & \\ 0 & c_2 & \ddots & & \\ & & \ddots & & b_{n-1} \end{vmatrix} = \begin{vmatrix} a_1 & b_1 & 0 & \cdots & 0 \\ c_1 & a_2 & b_2 & & \\ 0 & c_2 & \ddots & & \\ & & \ddots & & c_{n-1} \\ 0 & \cdots & c_{n-1} & a_n & \end{vmatrix} \quad \text{对第 } n \text{ 行 } \begin{array}{l} \text{全其 } D_n, \text{ 对 last row expand, } = a_n D_{n-1} - c_{n-1} \end{array} \quad \begin{vmatrix} a_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & b_{n-1} \end{vmatrix} = 0 + 0 = 0$$

$$= a_n D_{n-1} - (c_{n-1} \cdot b_{n-1}) \cdot D_{n-2}$$

$$\text{例 5} \begin{vmatrix} x & y & \cdots & y \\ z & \cdots & \cdots & y \\ \vdots & \ddots & \ddots & y \\ z & \cdots & z & x \end{vmatrix} = \begin{vmatrix} x & \cdots & 0 \\ z & \cdots & \cdots \\ \vdots & \ddots & \cdots \\ z & \cdots & z-x \end{vmatrix} + \begin{vmatrix} x & \cdots & y \\ z & \cdots & y \\ \vdots & \ddots & y \\ z & \cdots & z-x \end{vmatrix} \quad \text{last col } \begin{array}{l} \text{对 } (x-y) \text{ 提 } \\ \text{split } \end{array} \quad \text{对 } x-1 \text{ 加此行}$$

$$= (x-y) D_{n-1} - (-1)^{n-1} y (z-x)^{n-1} \text{ expand last col}$$

$$\text{例 6} \neq k \text{ st. } \begin{vmatrix} b_1+c_1, b_2+c_2, b_3+c_3 \\ a_1+c_1, a_2+c_2, a_3+c_3 \\ a_1+b_1, a_2+b_2, a_3+b_3 \end{vmatrix} = k \begin{vmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{vmatrix} \quad \text{提出 r1 的 2 乘于 r2, r3 去 a}$$

$$\square = - \begin{vmatrix} -(b_1+c_1) \cdots -(b_3+c_3) \\ \cdots \\ \cdots \end{vmatrix} = - \begin{vmatrix} 2a_1 \cdots 2a_3 \\ \cdots \\ \cdots \end{vmatrix} \quad \text{r2, r3 加于 r1} = -2 \begin{vmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{vmatrix} = 2 \begin{vmatrix} a_1 & \cdots & \\ b_1 & \cdots & \\ c_1 & \cdots & \end{vmatrix}$$

$$\text{例 7} \quad \boxed{1 \cdots 1 \cdots 1 \cdots 1}$$

□

$$= \begin{vmatrix} a_1 & \cdots & \\ a_2 & \cdots & \\ \vdots & \ddots & \end{vmatrix}$$

$$= 0 + 0 = 0$$

determinant 1
minor

1. 令 $A_{n \times n}$, DEF $\tilde{A}_{ij} \in (n-1) \times (n-1)$, \tilde{A}_{ij} 为 A 中除去第 i 行第 j 列.

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recursively def $\det(A) = \sum_{j=1}^n (-1)^{1+j} A_{1,j} \cdot \det(\tilde{A}_{1,j})$, cofactor expansion (1st row)
 $\Rightarrow A \in \mathbb{R}^{2 \times 2}, \det\begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$

P1 i) ($\det \rightarrow$ linear function of row)

$$\det\begin{pmatrix} -a_1 \\ -u+kv \\ 0_1 \end{pmatrix} = \det\begin{pmatrix} -a_1 \\ -u \\ 0_1 \end{pmatrix} + k \det\begin{pmatrix} -a_1 \\ -v \\ 0_1 \end{pmatrix} \leftarrow r \text{ row}$$

\square induction on n . 对 $n-1$ 成立, 对 n , $A_{r,r} \in (b_1+kC_1, \dots) = u+kv$, 对 $r=1, \text{ trivial. } \forall r>1$,
 考虑 $\tilde{A}_{1,j}$, $\tilde{A}_{1,j}$ 由 $(b_1+kC_1, \dots, b_{j-1}+kC_{j-1}, b_{j+1}+kC_{j+1}, \dots)$, 由 $\tilde{B}_{1,j}$ 为 $r-1$ 行 + $\tilde{C}_{1,j}$ 为 $r-1$ 行, 且其余

row $\tilde{A}_{1,j}$ 同 $\tilde{B}_{1,j}$ 同 $\tilde{C}_{1,j}$. 因此 $\det(\tilde{A}_{1,j}) = \det(\tilde{B}_{1,j}) + k \det(\tilde{C}_{1,j})$

$$\det(A) = \sum_{j=1}^n (-1)^{1+j} A_{1,j} \det(\tilde{A}_{1,j}) = \sum_{j=1}^n (-1)^{1+j} A_{1,j} \cdot \det(\tilde{B}_{1,j}) + k \sum_{j=1}^n (-1)^{1+j} A_{1,j} \det(\tilde{C}_{1,j}) = \det B + k \det C$$

$$(i) \text{ if all } 0, \det = 0 \quad \square \text{ by i) } \det\begin{pmatrix} u \\ 0 \\ 0 \end{pmatrix} = \det\begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} = 0. \det\begin{pmatrix} -a_1 \\ -v \\ 0_1 \end{pmatrix} = 0 \quad (\text{同理})$$

$$(ii) A \text{ row } i = e_k = (0 \dots \underset{i}{1} \dots 0), \det + A = (-1)^{i+k} \det(\tilde{A}_{i,k})$$

\square 归纳 n . 若 $n-1$ 成立, 对 n , 若 $i=1$, 由 i) 及 $\tilde{A}_{1,j}$; 若 $i>1$, 考虑 $\forall j$, $\tilde{A}_{1,j}$ 为 $i-1$ 行 $\left\{ \begin{array}{l} \text{shift up } \begin{matrix} 1 \\ \vdots \\ i-1 \\ \vdots \\ k \end{matrix} \\ \text{row } i-1 \\ \text{row } i \\ \text{row } k \end{array} \right\}$

$\square \tilde{A}_{1,j} \in \mathbb{R}^{(n-1) \times (n-1)}$, induction hypothesis, $\det(\tilde{A}_{1,j}) = \begin{cases} (-1)^{1+j+k-1} \det(C_{ij}) & j < k \\ 0 & j=k \text{ 行全 0} \\ (-1)^{i+k} \det(C_{ij}) & j > k \text{ 其他不变} \end{cases}$

$$\begin{aligned} \therefore \det A &= \sum_{j=1}^k (-1)^{1+j} A_{1,j} \det(\tilde{A}_{1,j}) = \sum_{j=1}^k (-1)^{1+j} A_{1,j} (-1)^{i+k} \det(C_{ij}) + \sum_{j=k+1}^n (-1)^{1+j} (-1)^{i+k} A_{1,j} \det(C_{ij}) \\ &= (-1)^{i+k} \left[\sum_{j=1}^k (-1)^{1+j} A_{1,j} \det(C_{ij}) + \sum_{j=k+1}^n (-1)^{1+j} A_{1,j} \det(C_{ij}) \right] \quad \text{由 ii) } \tilde{A}_{1,j} \text{ 为 1st row cofactor expansion. } = \det \tilde{A}_{i,k} \end{aligned}$$

iv) (Laplace Expansion) cofactor expansion 为 A 的 i 行展开

$$\square \text{ 对 } \forall i, \det A = \sum_{j=1}^n (-1)^{1+j} A_{ij} \det(\tilde{A}_{ij}), i=1 \text{ 时 } \det A = A_{1,1} \cdot e_1 + A_{1,2} \cdot e_2 + \dots + A_{1,n} \cdot e_n \stackrel{(1|1)}{=} 1 \cdot (1|00) + 3 \cdot (0|10) + 2 \cdot (0|01)$$

$$\text{由 P1 i), } \det A = A_{1,1} \det\begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix} + \dots + A_{1,n} \det\begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix} = \sum_j A_{1,j} \det\begin{pmatrix} 0 \\ \vdots \\ 0 \\ e_j \\ 0 \end{pmatrix} = \sum_j A_{1,j} (-1)^{1+j} \det \tilde{A}_{1,j} \text{ 由 iii) }$$

v) 二行相加, $\det = 0$

\square induction n . 若 $n-1$ 成立, 对 n , 若 n 行同 i 行 $\neq s$, $\det A = \sum_{j=1}^n (-1)^{1+j} A_{ij} \det \tilde{A}_{ij}$

$$\begin{aligned} \text{vi) } \text{对 } \forall i \text{ 行相加, } 1) \text{ row scaling} &\quad \det A \rightarrow \alpha \det A \\ 2) \text{ row exchange} &\quad \det A \rightarrow -\det A \\ 3) \text{ row adjustment} &\quad \det A \rightarrow \det A \quad \text{对 } i \text{ 行} \rightarrow \begin{array}{c} \text{接 } G \\ = 0, v \end{array} \\ \square 1) \text{ 由 i) } 2) \text{ 对 } i, j \text{ 行. } & \quad \det\begin{pmatrix} -a_1 \\ -a_1 + a_2 \\ -a_1 + a_2 + a_3 \\ \vdots \\ -a_n \end{pmatrix} = \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 + a_2 \\ \vdots \\ -a_n \end{pmatrix} + \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 + a_2 \\ \vdots \\ -a_n \end{pmatrix} = \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 \\ \vdots \\ -a_n \end{pmatrix} + \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 \\ \vdots \\ -a_n \end{pmatrix} \\ & \quad + \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 \\ \vdots \\ -a_n \end{pmatrix} + \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 \\ \vdots \\ -a_n \end{pmatrix} \\ & \quad \text{由 } 3) \det\begin{pmatrix} -a_1 \\ -a_1 + a_2 \\ -a_1 + a_2 + a_3 \\ \vdots \\ -a_n \end{pmatrix} = \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 + a_2 \\ \vdots \\ -a_n \end{pmatrix} + \det\begin{pmatrix} -a_1 \\ -a_1 \\ -a_1 \\ \vdots \\ -a_n \end{pmatrix} = 0 \end{aligned}$$

vii) 若 i 行全 0, $\det = 0$ \square 由 i) 和 v)

$$viii) \det(ka) = k^n \det(a) \quad \square \det\begin{pmatrix} -ka_1 \\ -ka_2 \\ \vdots \\ -ka_n \end{pmatrix} = k \det\begin{pmatrix} -a_1 \\ -a_2 \\ \vdots \\ -a_n \end{pmatrix} = \dots = k^n \det\begin{pmatrix} -a_1 \\ -a_2 \\ \vdots \\ -a_n \end{pmatrix}$$

$$\therefore \text{若 } n=\text{even, } \det(-A) = \det A \quad \text{或 } \det(kA) = \det kI \times \det A = k^n \det A$$

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Ex 1 $A, B \in M_n(\mathbb{C})$, B invertible, $\exists k \in \mathbb{C}$ st. $A+kB$ not invertible

$$\square \det(A+kB) = \det B \cdot \det(B^{-1}A + kI_n) \quad \text{by fundamental thm of algebra} \quad \exists \text{ zero(s) } \Rightarrow \text{ polynomial} = 0$$

$\xrightarrow{k \neq -\lambda}$ $\xrightarrow{\text{fundamental thm of algebra}}$ $\xrightarrow{\exists \text{ zero(s)}}$ $\xrightarrow{\text{polynomial} = 0}$

def 2

$$(\det A = 0 \Leftrightarrow \text{rank } A < n, \text{ or } \det I = 1 = \det A \cdot \det B = 0)$$

文字为 bi-condition, $\det A \neq 0 \Leftrightarrow \text{rank } A = n$, $\exists E$

1 T₁ 若 $A_{m \times n}$, $\text{rank } A < n$, $\det A = 0$

23. P₂₃ \square rows L. dep. $\therefore \exists \text{ row } a_r = \sum_{i \neq r} k_i a_i$ #2, $\det \begin{pmatrix} -a_1 \\ -a_2 \\ \vdots \\ -a_r \\ -a_{n+1} \end{pmatrix} = \det \begin{pmatrix} -a_1 \\ -k_1 a_1 + k_2 a_2 + \dots + k_n a_n \\ -a_{n+1} \end{pmatrix} = \det \begin{pmatrix} -a_1 \\ -0 \\ -a_{n+1} \end{pmatrix} = 0$

2 T₂ $\forall A, B \in M_n(\mathbb{R})$, $\det AB = \det A \cdot \det B$

23. P₂₃ \square lemma: $\forall \text{ elementary } E$, $\det EB = \det E \cdot \det B$. 24. P₂₄ \square $\text{rank } A = n$, $A \in M_n(\mathbb{R}) \therefore A = E_m \cdots E_1$ lemma

25. P₂₅ \triangle 若 E type I), LHS = $\det(B \text{ row scale, } \lambda \cdot \det B)$...

$$\begin{aligned} \text{RHS} &= \lambda \cdot \det B \\ \text{若 } \text{rank } A < n, \det A = 0, \therefore \text{RHS} = 0. \text{ To } \Rightarrow \text{AB}, \text{ rank } AB \leq \text{rank } A < n, \therefore \text{LHS} = 0 \end{aligned}$$

$$\begin{aligned} \det AB &= \det(E_m \cdots E_1 \cdot B) = \det(E_m) \cdot \det(E_{m-1} \cdots E_1 B) \\ &= \cdots = \det(E_m) \cdots \det(E_1) \det B = \det(E_m \cdots E_1) \det B \\ \text{lemma } E \text{ 不是 } R_S, &= \det A \cdot \det B \end{aligned}$$

3 T₃ $\det A^T = \det A$

23. P₂₃ \square $\text{rank } A < n$, $\text{rank } A^T = \text{rank } A$, $\therefore \det A \neq 0$; A invertible, $A = E_m \cdots E_1$, $\therefore \det A^T = \det E_1 \cdots \det E_m^T$

$$\therefore \det A^T = \det(E_1^T \cdots E_m^T) = \det(E_1^T) \cdots \det(E_m^T) = \det(E_1) \cdots \det(E_m) = \det(E_m \cdots E_1) = \det A$$

24. P₂₄ \triangle (使用 $|A| = \sum (-1)^i \text{Laplace expansion by } i \text{ th col/row, 增加 } P_S, P_I, N$)

4 T₄ $\det A = \prod_{i=1}^n \lambda_i$ λ_i eigenval

23. P₂₃ \square characteristic func $f(\lambda) = (\lambda - \lambda_1)(\lambda - \lambda_2) \cdots (\lambda - \lambda_n) = -(\lambda_1 - \lambda)(\lambda_2 - \lambda) \cdots (\lambda_n - \lambda) = (\lambda - \lambda_1) \cdots (\lambda - \lambda_n)$

因入为变量, 令 $\lambda = 0$, 则 $f(0) = \prod_{i=1}^n \lambda_i = \det(A - 0 \cdot I)$ (NOTE: PP 24. T₄ in NOTE, $R_S = \prod_{i=1}^n \lambda_i$)

UR 若 \exists eig decomposition $A = Q \Lambda Q^{-1}$, $\det A = \det Q \det \Lambda \det Q^{-1} = \det \Lambda = \prod_{i=1}^n \lambda_i$

$$\begin{array}{c|ccccc} 1 & y & y & -y & y & \\ \hline 2 & x & 1 & -y & y & \\ 2 & z & x & -y & y & \\ 2 & z & z & x & y & \\ 2 & z & z & -x & x & \\ \hline \end{array} \quad \begin{array}{c|ccccc} 1 & y & -y & -y & y & \\ \hline 1 & y & -y & -y & y & \\ 1 & z & x & -y & y & \\ 1 & z & z & x & y & \\ 1 & z & z & -x & x & \\ \hline \end{array}$$

$$= (\lambda_1 D_{11} + (\lambda_2)^2 y_1 z_1) + (\lambda_2 D_{22} + (\lambda_1)^2 y_2 z_2) + \cdots + (\lambda_n D_{nn} + (\lambda_{n-1})^2 y_n z_n)$$

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Vector Space

向量空间 (Vector Space) 定义和性质
 n -dim 向量空间

Let K field, $(\text{Vector Space definition}, K^n = \{(a_1; \dots; a_n) \mid a_i \in K\})$

V over K satisfy below $\Rightarrow V$ vector space

$$+ : V \times V \rightarrow V, \quad \cdot : K \times V \rightarrow V \quad (\text{operation is closed})$$

$$(V, +) \quad \text{abelian group} \quad (A1) \quad \forall x, y \in V, x+y=y+x \quad \text{commutative}$$

$$(A2) \quad \forall x, y, z \in V, (x+y)+z=x+(y+z) \quad \text{associative}$$

$$(A3) \quad \exists ! \underline{0} \in V \quad \forall x \in V, x+0=x \quad \text{identity} \quad ((0, \dots, 0) = \underline{0})$$

$$(A4) \quad \forall x \in V, \exists ! y \in V, x+y=0 \text{ inverse} \quad (\text{let } x=(x_1; \dots; x_n) \in K^n, y=-x=(-x_1; \dots; -x_n))$$

$$\text{monoid } (M1) \quad \forall a, b \in K, \forall x \in V, (ab)x=a(bx) \quad \text{associative}$$

$$(M2) \quad \exists ! \underline{1} \in K, \forall x \in V, 1 \cdot x=x \quad \text{identity}$$

$$(D1) \quad \forall x, y \in V, \forall c \in K, c(x+y)=cx+cy$$

$$(D2) \quad \forall x \in V, \forall c, d \in K, (c+d)x=cx+dx$$

! 代入 $ax=bx \Rightarrow a=b$ ($x \neq 0$) 或 $a \cdot x = a \cdot y \Rightarrow x=y$ ($a \neq 0$)

$$1. \text{ i) cancel law: } \forall y, z \in V, x+z=y+z \Rightarrow x=y$$

$$\text{ii) } A3, A4, M2 \text{ not unique} \quad \text{to } A3 \quad a+0'=0' \text{ (0' unique)} \quad \therefore 0=0'$$

$$\text{iii) } \underline{0}x=\underline{0} \quad \forall x \in V \quad 0x=(0+0)x=0x+0x, \therefore 0x=0 \quad (\text{i.i) cancel law})$$

$$\text{iv) } x0=\underline{0} \quad \forall x \in K \quad x0=x(0+0)=x0+x0 \quad P_2 \\ P_1$$

注意 set!!

2. $f \in S$, F field, $\mathcal{F}(S, F)$ set of all $f: S \rightarrow F$. let $f, g \in \mathcal{F}(S, F)$, $c \in F$

$$f=g \text{ iff } f(s)=g(s) \quad \forall s \in S, +: (f+g)(s)=f(s)+g(s), \cdot: (cf)(s)=c f(s), 0 \cdot f(x)=0$$

3. F field, $\mathbb{P}[F]$ coeff $\in F$ has polynomial space. $P_n(x) = a_0 + a_1 x + \dots + a_n x^n \in \mathbb{P}[F]$, degree n

事实上 $\mathbb{P}_n[F] \cong \mathbb{P}[F]$ & degree at most n , 也是 vector space over F . ($\mathbb{P}_n(F)$ 是 $\mathbb{P}(F)$ 的子空间)

$$\text{且 } P_n(x) = a_0 + \dots + a_n x^n \in \mathbb{P}_n[F] \text{ if } \begin{bmatrix} a_0 \\ \vdots \\ a_n \end{bmatrix}$$

4. 最小的 vec. space over any field F is $\{0\}$, $0+0=0$

• (seq space) Seq(F) & ele in F has seq. $\{s_n\} + \{t_n\} = \{s_n + t_n\}$, $c\{s_n\} = \{c \cdot s_n\}$

• (coordinate space) $(a_1, \dots, a_n), a_i \in F : F^n$

• (Complex) \mathbb{C} over \mathbb{C} basis $\{1\}$, \mathbb{C} over \mathbb{R} basis $\{1, i\}$, \mathbb{R} over \mathbb{C} not VS. $i \cdot 1 = i \notin \mathbb{R}$

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9

1 \mathbb{R}^+ all positive reals. \mathbb{R}^+ over \mathbb{R} VS, addition: $x \oplus y = xy$

scalar multi: $\alpha \odot y = y^\alpha$

$d_m = 1$, basis: $\{1\}$ ($\forall x \in \mathbb{R}^+, x = (\log x) \odot 10 \quad 1 = e^{kx - 2} L. dep$)

Ex 1. #. 1st: $1^o V = \{(a_1, a_2) | a_1, a_2 \in F\}$, +: card \rightarrow $(a_1, a_2) + (b_1, b_2) = (a_1 + b_1, a_2 + b_2)$

$2^o V = \{(a_1, a_2) | a_1, a_2 \in \mathbb{R}\}$, +: $(a_1, a_2) + (b_1, b_2) = (a_1 + 2b_1, a_2 + 3b_2)$

$\square 1^o \times \because \boxed{0 \cdot (a_1, a_2) = (a_1, 0)}$ zero vec. unique $2^o \times \boxed{[(a_1, a_2) + (b_1, b_2)] + (c_1, c_2) \neq (a_1, a_2) + [(b_1, b_2) + (c_1, c_2)]}$

1. subspace $W \subseteq$ vector space V

subspace 不一定将 x, y 包括 $\cup S$

应该满足 vector space 的性质. P_6 为零 & 起点 (A3), (A4), 及 "+" / "-" closure 为 subspace 必具的 (原故本)

而因 5 的 3 条 ("+" / "-" closure + (A3)) 可证 (A4)

□ let x inverse of $(-1) \cdot x$, $x + (-1) \cdot x = (1 + (-1)) \cdot x = 0 \cdot x = 0$ $\forall x \in W$
field multi. identity has odd inverse $\forall x \in W$ 为 W 的子空间

NOTE, 由 $\forall x \in V, 0 \cdot x = 0$ ($P_6, 1, iii$) 故: $0 \in V$ 是在任 V 的满足 "+" closure 的 subset

1. subspace 为 "zero vector in W " 还保障了至 W 非空 ($\exists \phi$, closure trivially true)

ex 1. W 为 set of symmetric matrix in $M_{n,n}(F)$, $W \not\subseteq M_{n,n}(F)$ 为 subspace. $\square (A+B)^T = A^T + B^T = A + B \in W$

2. $F(R, R)$ 为 subspace 为 all continuous func: $R \rightarrow R$ \square sum of cont. cont.

$$M_{1,1} = \{0, 1, -1\}$$

set of $M_{n,n}(F)$ trace = 0

3. set of diagonal matrix \subseteq subspace of $M_n(F)$ / set of upper triangular matrix subspace of $M_n(F)$

2. P_1 intersection of subspaces of V . V 为 V 为 subspace \Rightarrow intersection 是 largest subset of V
□ let C 为 V 的 subspaces, $W \not\subseteq C$ 为 C 的交集 (intersection) $\cap C_i$. $0 \in W$
 $x, y, c \in W$ (x, y 在 C_i 中)

NOTE, union 为 subspace, $\Rightarrow V = \mathbb{R}^2$, $W_1 = \{[x]: x \in \mathbb{R}\}$, $W_2 = \{[y]: y \in \mathbb{R}\}$, 及 P_2

{3.1} V 为 V.S. 且 $W \subseteq V$ 为 V.S., $W \not\subseteq V$ subspace \times 除非 "+" 也是 - 一致 $\Rightarrow V = \mathbb{R}$ on field \mathbb{R}
 $W \not\subseteq \mathbb{R}^3$ 为 xy-plane, $\exists W = \{(a_1, a_2, 0)\}$, $\exists W = \mathbb{R}^2 \times$ $W \subseteq \mathbb{R}^2$, $w \in W$ 为 3 元, $\mathbb{R} \not\subseteq \mathbb{R}^2$ 只有 2 个
isomorphism

P_2 . $W_1, W_2 \not\subseteq V$ 为 subspace, $W_1 \cup W_2 \not\subseteq V$ subspace $\Leftrightarrow W_1 \subseteq W_2 \vee W_2 \subseteq W_1$

□ \Leftarrow trivial, \Rightarrow 若两等都不对, $\exists x \in W_1 - W_2, y \in W_2 - W_1$. $\exists x, y \in W_1 \cap W_2, x+y \in W_1, W_2$
对 $W_1, \dots, W_n \subseteq V$, $\exists \cup W_i \subseteq V \Leftrightarrow \exists W_k$ 为 W_i 为 n 个 all. 互不相容. $x, y \in W_1 \cap W_2$ 互不相容
($\subseteq V$)

3. 对 W_1, W_2 , $\exists W_1 + W_2 = \{(x+y) | x \in W_1, y \in W_2\}$, if $W_1 \cap W_2 = \{0\} \wedge W_1 + W_2 = V$, $\square V = W_1 \oplus W_2$

P_1 若 W_1, W_2 为 V 的 subspaces, $W_1 + W_2$ 为 subspace; 任 subspace of V 含有 both W_1, W_2 必也含有 $W_1 + W_2$

□ i) $\forall 0 + 0 \in W_1 + W_2$ ii) $\forall W$ subspace, $\exists x \in W_1, x \notin W_2 \Rightarrow W_1 + W_2 \subseteq W$
 $\forall x, y, z \in W_1 + W_2 \in W_1 + W_2 \subseteq W$

{3.2}. 证 $\mathbb{F}^n = \{(a_1, \dots, a_n) | a_n = 0\} \oplus \{(a_1, \dots, a_n) | a_1 = \dots = a_{n-1} = 0\}$

□ $W_1 \cap W_2 = \{0\}$, 而 $\{W_1, W_2\} \subseteq V$: 易证 W_1, W_2 为 \mathbb{F}^n , $\therefore W_1 + W_2$ 为 subspace, 为 subset

$V \subseteq W_1 + W_2$: let $v = (v_1, \dots, v_n) \in V$, $v = (v_1, \dots, v_{n-1}, 0) + (0, \dots, 0, v_n)$ 为 subspace, $\mathbb{R} \oplus \mathbb{R}$

{3.3} 证 当 F 为 characteristic 2, $M_n(F) = \{\text{set of skew-symmetric}\} \oplus \{\text{set of symmetric}\}$

□ $W_1 \cap W_2 = \{0\}$, 而 $M_n(F) = \{A: A \in M_n(F)\} = \{(A+A^T) + (A-A^T) | A \in M_n(\mathbb{R})\} = W_1 + W_2$
若 F characteristic 2, $F \cong \mathbb{Z}_2$, $M^T = -M^T$ ($-1 \equiv 1$)

事实上, F 为 characteristic 2 且, $M_n(F) = \{\text{set of symmetric}\} \oplus \{\text{set of } A, A_{ij} = 0 \text{ } i \neq j\}$

□ $W_1 \cap W_2 = \{0\}$, 而 $M_n(\mathbb{R}) = \{A: A \in M_n(F)\} = \{T+A-T\} = W_1 + W_2$, $T_{ij} = T_{ji} = A_{ij}$ 且 $T \in W_1$

NOTE, 由例 3, 3. 两个 W_1, W_2 不唯一:

P_2 . W_1, W_2 为 V 的 subspace, $W_1 \oplus W_2 = V \Leftrightarrow \forall v \in V, v = x_1 + x_2$ uniquely, $x_1 \in W_1, x_2 \in W_2$

□ \Rightarrow 若不唯一, $v = x_1 + x_2 = x'_1 + x'_2$ $\frac{x_1 - x'_1}{W_1} = \frac{x_2 - x'_2}{W_2} \Rightarrow W_1 \cap W_2 \neq \{0\} \therefore x_1 = x'_1$
 $\therefore \exists t \in W_1 \cap W_2 \neq \{0\}, \exists t \in W_1 \cap W_2$ 而 $t \neq 0$, $\square t = t + 0 = 0 + t$, 不唯一

VS4
4. eg 1. 对于向量空间 V over F , 令 $v \in V$, $\langle v \rangle = \{av \mid a \in F\} \subseteq V$, 为包含 v 的子空间
由 group theory, generator, 但不像是 coset, 但 H 为 G 的子群, 因此 $a \in H \Rightarrow a \in G$. 而此处 $a \in F \neq V$

eg 2. $W \leq V$, $V + W = \{v + w \mid v \in V, w \in W\}$ 为 V 的子空间. (Alg 2nd off the sequence)

由 group theory, $v + W \in V \Leftrightarrow v \in W$; $v_1 + W = v_2 + W \Leftrightarrow v_1 - v_2 \in W$

\square i) \Rightarrow $0 \in v + W$, $0 = v - v \in W$, $\therefore v \in W \Leftrightarrow v + W \in W$ closed

ii) \Rightarrow $v_1 + w_1 = v_2 + w_2$, $v_1 - v_2 = w_2 - w_1 \in W$, $\Leftrightarrow v_2 - v_1 \in W \Leftrightarrow v_1 + W = v_2 + W$ (由 $t = v_1 - v_2 \in W$, $v = t + v_2 \in v_2 + W$)

定理 coset 为 operation: $(v_1 + W) + (v_2 + W) = (v_1 + v_2) + W$ 可见 quotient space $V/W \cong V/S$.

$a(v_1 + W) = (av_1) + W$

5. P_1, P_2 的脚注: \Rightarrow 既是 S 的子空间, 且含有 all. $\exists u \in S_1 \setminus (S_2 \cup \dots \cup S_n)$

$\exists v \in (S_2 \cup \dots \cup S_n) \setminus S_1$ 且 $au + v \in S_1$, 因 $u \in S_1, -au \in S_1$, $au + v - au = v \in S_1$, 矛盾.

若 $au + v \in S_k$ 为矛盾, 由 $a \neq b \neq a$, $b + v \in S_k$, $b + (a - b)v \in S_k$ 但 $u \in S$, 矛盾.

若拓展至 infinite, 对 partial ordered set $(P(V), \subseteq)$ 的特别 (sub(V), \subseteq), 基本定理 chain

$C = \{W_1, \dots\}$ s.t. $W_1 \subseteq W_2 \subseteq \dots$, $\bigcup C$ 为 subspace of V . P_2

\square 若 $u, v \in \bigcup C$, $u \in W_i$, $v \in W_j$, $i > j$, 则 $u + v \in W_j$

6. 对 direct product 增加维度, 对 V_1, \dots, V_n over same field F ,

$\prod^n V_i = \{(v_1, \dots, v_n) \mid v_i \in V_i\}$ 为 V 的子空间 over F , addition: $(v_1, \dots, v_n) + (u_1, \dots, u_n) = (v_1 + u_1, \dots, v_n + u_n)$

乘法: $\forall v_i \in V_i, \forall v_j \in V_j$, multiplication: $(v_1, \dots, v_n) \cdot (w_1, \dots, w_n) = (v_1 \cdot w_1, \dots, v_n \cdot w_n)$

eg. $\mathbb{R}^3 = \mathbb{R} \times \mathbb{R} \times \mathbb{R}$ over \mathbb{R}

7. T_3 为 V.S. V , $W \leq V$, $\exists T \leq V$ s.t. $W \oplus T = V$

\square let F family of all sub(V) st. disjoint from W . 对 order operation \subseteq , let $C \in F$ 的 chain,

假设 $\bigcup C \leq V$, 且 $\bigcup C \cap W = \emptyset$, 由 V 为 upper bound, 由 Zorn's lemma, \exists maximal of F : M

if $W \oplus M \leq V$, 则 $\exists e \in V - (W \oplus M)$, $e \neq 0$, ???

(由 P_2 为 chain)

NOTE, 假设了 basis, 可用 basis 证明: 易证当 $W \oplus T \leq V$, $\beta_W \cap \beta_T = \emptyset$, $V = W \oplus T \Leftrightarrow \beta_W \cup \beta_T = \beta_V$

由 T 为 W 的 basis β_W , def $T = \text{span}(\beta_V - \beta_W)$ 由 lemma, $W \oplus T = V$

In fact T 为一组, $\mathbb{R}^2 = V$, $W = \{(a_1, 0) \mid a_1 \in \mathbb{R}\}$, $T \cong \{(0, a_2) \mid a_2 \in \mathbb{R}\}$ 由 $T \cong \{(a_2, a_2) \mid a_2 \in \mathbb{R}\}$

13. 1. $W_1 = \{\text{set of odd func}\}, W_2 = \{\text{set of even func}\}$ $C(\mathbb{R}) = W_1 \oplus W_2$ 序数有限

\square 对 $\forall f \in C(\mathbb{R})$ $f(x) = \frac{1}{2}(f(x) - f(-x)) + \frac{1}{2}(f(x) + f(-x))$

odd, $g(x) = -g(-x)$ even, $g(-x) = g(x)$

例 1. 对 \$A_0\$ eg 1.3, \$\dim\$ & symmetric & skew-symmetric

\square diagonal: \$\{E_{ii}\}\$, \$\dim = n\$, trace = 0: $\begin{bmatrix} a_{11} & & \dots & a_{nn} \\ & a_{22} & & \\ & & \ddots & \\ & & & a_{nn} \end{bmatrix} = \{E_{ij} | i=j\} \cup \{\begin{bmatrix} -1 & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{bmatrix}, \begin{bmatrix} 0 & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 & & & \\ & 0 & & \\ & & \ddots & \\ & & & 0 \end{bmatrix}\}$

\square upper triangle: \$\{E_{ij} | i < j\}\$, \$\dim = \frac{(n+1)n}{2}

\square symmetric: $\{E_{ij} | i \leq j\}$, \$\dim = \frac{(n+1)n}{2}

\square skew-symmetric: $\{E_{ij} | i > j\}$, \$\dim = \frac{(n+1)n}{2}

1) T_1 \$W_1, W_2 \subseteq V\$, \$\dim V\$ finite. \$W_1 \subseteq W_2 \Leftrightarrow \dim(W_1 \cap W_2) = \dim(W_1)\$

$\square \Rightarrow W_1 \cap W_2 = W_1 \Leftrightarrow \text{let } \dim W_1 = n, \beta \text{ basis for } W_1 \cap W_2 \text{ note } W_1 \cap W_2 \subseteq W_1$

\$P \subseteq W_1\$, 且因 \$\dim W_1 = n\$, \$\beta \not\subseteq n+1\$. L.Ind. \$\dots \beta \not\subseteq W_1\$ basis. 由定理 \$W_1 \cap W_2 = W_1 \Leftrightarrow W_1 \subseteq W_2\$

OR: \$W_1 \cap W_2 \subseteq W_1\$ (由定理 \$W_1 \cap W_2 = W_1 \Leftrightarrow W_1 \cap W_2 \subseteq W_1\$) (\$W_1\$ 为子空间的向量空间 \$W_1 \cap W_2\$)

2) T_{2a} \$W_1, W_2 \subseteq V\$ 且 finite dim. \$W_1 + W_2\$ to \$\dim(W_1 + W_2) = \dim W_1 + \dim W_2 - \dim(W_1 \cap W_2)\$

\square let \$\{u_1, \dots, u_m\}\$ basis for \$W_1 \cap W_2\$, since \$W_1 \cap W_2 \subseteq W_1, W_2\$, extend to \$\{u_1, \dots, u_k, v_1, \dots, v_m\}\$ basis for \$W_1\$
\$\{u_1, \dots, u_k, w_1, \dots, w_n\}\$ basis for \$W_2\$

WTS \$\{u_1, \dots, u_k, v_1, \dots, v_m, w_1, \dots, w_n\}\$ basis for \$W_1 + W_2\$ LHS \$\in W_1\$ (由 \$u_i \in W_1\$), RHS \$\notin \in W_2\$, \$\therefore\$ 1) 1. 2. 3. 4.

i) L.Ind. let \$a_i, b_i, c_i \in F\$, \$\sum a_i u_i + \sum b_i v_i = -\sum c_i w_i \in W_1 \cap W_2\$
\$\therefore -\sum c_i w_i \in \{u_i\}\$ 表示: \$-\sum c_i w_i = t_i u_i, t_i \in F \Leftrightarrow w_i \in \{u_i\}\$ 由 \$w_i \in W_2\$ basis. L.Ind. \$c_i = t_i = 0\$
由理 \$\sum a_i u_i + \sum c_i w_i = -\sum b_i v_i, b_i = 0, \therefore \sum a_i u_i + 0 v_i = 0\$, 由 \$W_1 \cap W_2\$ basis, \$a_i = 0\$

ii) span: \$x \in W_1 + W_2 \Leftrightarrow (\sum a_i u_i + \sum b_i v_i) + (\sum c_i w_i + \sum d_i w_i)

3) T_{2b} \$V = W_1 + W_2\$, by \$V = W_1 \oplus W_2 \Leftrightarrow \dim V = \dim W_1 + \dim W_2\$

$\square W_1 \cap W_2 = \emptyset \Leftrightarrow \dim(W_1 \cap W_2) = 0 \Leftrightarrow \dim V = \dim W_1 + \dim W_2 (T_{2a})$

例 2. \$W_1, W_2 \subseteq V\$, \$\dim(W_1 \cap W_2) \leq \min(\dim W_1, \dim W_2)\$

\square let \$\dim W_1 \leq \dim W_2\$, note \$W_1 \cap W_2 \subseteq W_1, \dots \dim \text{LHS} \leq \dim W_1; \dim(W_1 \cap W_2) = \dim W_1 + \dim W_2 - \dim(W_1 \cap W_2)

例: 例 2 i) \$W_1 = \{[\begin{smallmatrix} 1 & 0 \\ 0 & 0 \end{smallmatrix}]\}\$ basis \$= \{[\begin{smallmatrix} 0 & 1 \\ 0 & 0 \end{smallmatrix}]\}\$ \$W_1 \cap W_2 = \{[\begin{smallmatrix} 0 & 0 \\ 0 & 0 \end{smallmatrix}]\}\$ \$\dim = 1\$

例: 例 2 ii) \$W_1 = \{[\begin{smallmatrix} 0 & 1 \\ 0 & 0 \end{smallmatrix}]\}\$ \$W_2 = \{[\begin{smallmatrix} 0 & 0 \\ 1 & 0 \end{smallmatrix}]\}\$ \$W_1 \cap W_2 = \{[\begin{smallmatrix} 0 & 0 \\ 0 & 0 \end{smallmatrix}]\}\$ \$\dim = 3\$

2) T_{2c} \$\dim(V) = \dim(W) + \dim(V/W)\$ P. 11.4.

令 \$\beta_W = \{v_1, \dots, v_n\}\$, \$\beta_V = \{v_1, \dots, v_n\}\$, WTS \$\{v_{k+1} + W, \dots, v_n + W\}\$ basis for \$V/W\$

iii) L.Ind. \$\sum_{i=k+1}^n k_i(v_i + W) = \sum_{i=k+1}^n (k_i v_i + W) = \sum_{i=k+1}^n (k_i v_i) + W = 0 + W \therefore \sum_{i=k+1}^n k_i v_i \in W\$

但此部分 \$v_i\$ 不在 \$W\$ 中, \$\therefore \sum_{i=k+1}^n k_i v_i = 0\$, 由 L.Ind. \$\therefore k_i = 0\$

iv) span. \$\forall x + W \in V/W, x + W = (\sum_{i=1}^k a_i v_i + \sum_{j=k+1}^n a_j v_j) + W\$. 由 \$\sum a_i v_i \in W, \therefore = \sum a_j v_j + W\$

NOTE, 由 L. Ind. 有逆定理: let \$T: V \rightarrow V/W, v \mapsto v + W\$, 由定理 11.4 有

\$\text{Null}(T) = \{v \in V | v + W = 0 + W\}\$ that is, \$v \in W, \therefore \text{Null}(T) = W\$

\$\therefore\$ rank Thm, \$\dim V \leq \text{rank } T + \text{nullity } T = \dim V/W + \dim W\$ 由定理 11.4

3. 有 \$T: V \rightarrow Z\$ linear, def \$\bar{T}: V/\text{Null}(T) \rightarrow Z, v + \text{Null}(T) \mapsto T(v)\$

证 \$\bar{T}\$ well-defined 且是同构, 由 \$T(x) = \bar{T}(x + \text{Null}(T))\$

\square well-defined: \$v + \text{Null}(T) = v' + \text{Null}(T) \Rightarrow T(v) = T(v') \Leftrightarrow T(v - v') = 0, T(v) = T(v')

linear 且 \$T\$ onto \$\forall z \in Z, \exists v \in V\$ s.t. \$T(v) = z\$ (因 \$v - v' \in \text{Null}(T), T(v - v') = 0, T(v) = T(v') = z\$)

1-1: 令 \$\bar{T}(x + \text{Null}(T)) = T(x) = z, x \in \text{Null}(T)\$ def. \$x + \text{Null}(T) = \text{Null}(T) = \bar{T}^{-1}(z)\$

对 \$\forall x \in V, T(x) = \bar{T}(x + \text{Null}(T)) = \bar{T}^{-1}(z) = z\$ (由定理 11.4)

例 1. $V_1, V_2 \leq V$, if $\dim(V_1 + V_2) = \dim(V_1 \cap V_2) + 1$, $\exists v \in V_1 + V_2 - (V_1 \cap V_2)$

由 $V_1 \cap V_2 \leq V_i \leq V_1 + V_2$ $\dim(V_1 \cap V_2) \leq \dim V_i \leq \dim(V_1 + V_2)$ $\Rightarrow V_1 \cap V_2 = V_{i-1}$ (8-7)

由 $\dim V_i = \dim(V_1 \cap V_2)$ or $\dim(V_i) = \dim(V_1 + V_2)$ 同理 两种方法都成立
 $\Rightarrow \forall i, V_{i-1} \supseteq V_i$, $\forall i, V_{i-1} = V_1 + V_2$

$$\dim(V_1 \times V_2) = \dim(V_1) + \dim(V_2)$$

1. \sim is an equivalence relation on S . 因 equivalence relation 又称为 partition (Algebra P17. 2 Lemma)

为了叙述方便 $s \in S$, s 属于 M 个 equivalent class, 在各 equivalent class E_i 中选一个 c_i ;
将 $s \in c_i$ 记为:

$C \subseteq S$ 为 system of distinct representatives for \sim on S iff $\forall E_i$ contains 1 member in C

C 也称 set of canonical form

e.g. let $A, B \in M_{m \times n}(F)$, $A \equiv B$ iff \exists invertible $P \in M_m(F)$, $Q \in M_n(F)$, $B = P A Q$
 \Leftrightarrow A 通过 row/column operations reduce to B

事实上, \forall matrix R 会 reduce to $J_r = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix}$.

且 no distinct $J_s \equiv J_r \Leftrightarrow \text{rank}(J_s) = s = \text{rank}(P J_r Q) = \text{rank}(J_r) = r$

这样 $C = \{J_r \mid r \geq 0\}$ 为 set of canonical form. $A \equiv B$ iff A, B 有同 canonical form J_s

2. f func 了等价类

constant on equivalent class

$f: S \rightarrow X$ 为 invariant under \sim (on S) iff $a \sim b \Rightarrow f(a) = f(b)$

complete invariant under \sim iff $a \sim b \Leftrightarrow f(a) = f(b)$

e.g. 2. \sim equivalence relation \equiv (Reg 1), rank 为 complete invariant func

$\square A \equiv B \Leftrightarrow \text{rank}(A) = \text{rank}(B)$

e.g. 3. $A, B \in M_n(F)$, $A \sim B$ similar ($A \sim B$) iff \exists invertible P , $B = P^{-1}AP$

可记 $A \sim B$ 为 equivalent relation on $M_n(F)$

Jordan

normal canonical form

matrix similar 有多种 invariants 但无 complete invariants

det: $\square \det(B) = \det(P^{-1}) \det(A) \det(P) = \det(A)$, 为 invariant, 但不同 matrix 会有同 det

trace: $\square \text{tr}(B) = \text{tr}(P^{-1}AP) = \text{tr}(AP^{-1}P) = \text{tr}(A)$, 但 $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ trace 为 1 但非 similar

NOTE: complete invariant 可完全决定 ele 属于哪个 equivalent class, 只用施加 func

但 complete invariant 很少, 一般用 noncomplete 的逆否来判断 ele 不属于哪个 class

Inverse 4 $\leftarrow \boxed{15}$

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Date.

$$uv = A^{-1} \frac{A^{-1} u v^T A}{1 + v^T A^{-1} u}$$

证 1. (Moore Inverse Lemma) 对 $J \in A$, $A + uv^T J \in \mathbb{R} \Leftrightarrow v^T A^{-1} u \neq -1$.

$$\square \Rightarrow [A + uv^T A^{-1}] u \neq 0 \Leftrightarrow u + u v^T A^{-1} u \neq 0 \Leftrightarrow u(1 + v^T A^{-1} u) \neq 0$$

\Leftarrow Note $P = A + \frac{A^{-1} u v^T A^{-1}}{1 + v^T A^{-1} u}$ 是 $Q = A + uv^T$ 的逆. By check $PQ = QP = I$

$\because A = C = I$ reduces 证 4. lemma 2, #从 lemma 2 令 $J = A^{-1} X$ 还原
 $v = C Y$

证 2 (Sherman - Morrison - Woodbury) $A + uv^T$ 可逆, 有逆的 $U \in \mathbb{R}^{n \times k}$ $V \in \mathbb{R}^{k \times n}$ 令 $P = I - A^{-1} - A^{-1} U (V^T A^{-1} V)^{-1} V^T A^{-1}$

待证 $= I - A + uv^T$ 可逆 \Leftrightarrow $uv^T (I - V^T A^{-1} V)^{-1}$ 可逆, 即 $(A + uv^T)^{-1} = A^{-1} - A^{-1} U (V^T A^{-1} V)^{-1} V^T A^{-1}$

证 3 $(I + P)^{-1} = I - (I + P)^{-1} P$ $\square \quad \forall U \in I + P \quad U(I - U^{-1} P) = U - P = I$ 由 lemma 2 $\because A \in I$
 $B \in P$

证 4 (Push-Through) $U^T (I + V^T V)^{-1} = (I_d + U^T V)^{-1} U^T$

(lemma 1) $A(I + BA)^{-1} = (I + AB)^{-1} A$ $A(I + BA) = (I + AB)A \Leftrightarrow A(I + BA)^{-1} = (I + AB)^{-1} A$

(lemma 2) $(I + AB)^{-1} = I - A(I + BA)^{-1} B$ 考虑 $(I - A(I + BA)^{-1} B)(I + AB) = I + AB - (I + AB)^{-1} AB(I + AB)$
 $= I + AB - (I + AB)^{-1} (I + AB)(AB) = I + AB - AB = I$, $\therefore I - A(I + BA)^{-1} B$, lemma 1

\therefore 有 lemma 1 \rightarrow Push-Through

Date: / /

Ex 1. Mat 10x10 $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$ $\square A = I + uu^T$, $u = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ $\Rightarrow I - \frac{uu^T}{1+10} = \begin{pmatrix} 9/10 & -1/10 \\ -1/10 & 9/10 \end{pmatrix}$

$$1 \text{ Frobenius/energy } \|A\|_F = \|A^T A\|_F = \sqrt{\sum_{ij} a_{ij}^2} = \|\text{vec}(A)\|_2$$

$$\text{note } \|A\|_F^2 = \operatorname{tr}(AA^T) = \operatorname{tr}(A^TA) \quad \square (A^TA)_{ij} = \sum_k A_{ik}^T A_{kj} = \sum_k A_{ki} A_{kj}$$

P1a) 为 matrix $n \times n$

□ (iii) $\|AB\|_F = \sqrt{\text{tr}(AB^T AB)} = \sqrt{\text{tr}(B^T A^T AB)} = \sqrt{\text{tr}(B^T (A^T A) B)}$

$$1b) \text{ For } \vec{x}, \vec{y} \in \mathbb{R}^n \quad \|\vec{x} \otimes \vec{y}\|_F = \|\vec{x}\|_{\ell_2} \|\vec{y}\|_{\ell_2} = (\sum_i \|\vec{x}_i\|^2)^{\frac{1}{2}} (\sum_j \|\vec{y}_j\|^2)^{\frac{1}{2}} = \|\vec{A}\|_F \|\vec{B}\|_F$$

$$\square x \otimes y = xy^T \quad \text{LHS} = \sqrt{\sum_{ij} x_i y_j^2} = \sqrt{\sum x_i^2 \sum y_j^2} = \text{RHS}$$

1c) energy preserving orthogonal transform And Periodic signal $\|AP\|_2^2 = \|A\|_2^2$

$$\boxed{\|AP\|_F^2 = \text{tr}(A^T P^T A) = \text{tr}(AA^T) = \|A\|^2}$$

$$d) \text{ } AB \text{ ist orthogonal iff } \operatorname{tr}(AB^T) = 0 \quad \text{und} \quad \|A+B\|_F^2 = \|A\|_F^2 + \|B\|_F^2$$

$$\boxed{\text{□} \quad \text{tr}((A+\Delta)(A^T+\Delta^T)) = \text{tr}(AA^T + AB^T + BA^T + \Delta\Delta^T) \stackrel{P_2.1}{=} \text{tr}(AA^T) + \text{tr}(\cancel{AB^T}) + \text{tr}(\cancel{BA^T}) + \text{tr}(\Delta\Delta^T)}$$

e) $x \in \mathbb{R}^n$, $\forall a, b$ 有 x^T ~~Pr~~^a $b = x^T a + b^T$

$$\exists \quad \text{tr}((\mathbf{x}_b^T)^T \mathbf{x}_{a^T}) = \text{tr}(\mathbf{b}^T \mathbf{y}^T \mathbf{x}_{a^T}) = 0$$

$$f) \|A\|_F = \sqrt{\sum_{i=1}^{\min(n,m)} \sigma_i^2(A)} \quad \text{Pf SVD } \operatorname{tr}(A^T A) = \operatorname{tr}(V \Sigma^T \Sigma V^T) = \operatorname{tr}(\Sigma^T \Sigma) = \sum \sigma_i^2(A)$$

應用 1 (Ax = b) 在 A 與 b 的誤差為 ϵ 時，
 $(A + \frac{\epsilon}{\|A\|}I)\hat{x} = b + \frac{\epsilon}{\|A\|}b$ ，有 $\frac{\|\hat{x} - x\|}{\|x\|} \leq \frac{\|Ax - b\|}{(1 - \frac{\epsilon}{\|A\|})\|b\|} \left(\frac{\|Ab\|}{\|b\|} + \frac{\|A\|}{\|A\|} \right)$

$$\boxed{\text{有 } A^{-1}(A+\delta A)\hat{x} = A^{-1}(b+\delta b), \quad \hat{x} + A^{-1}\delta A\hat{x} = x + A^{-1}\delta b, \quad \hat{x}-x = A^{-1}\delta b - A^{-1}\delta A(x-\hat{x}) - A^{-1}\delta A\hat{x}}$$

$$\Rightarrow \|x - x'\| \leq \|A^{-1}\| \cdot \|s b\| + \|A^{-1}\| \cdot \|s A\| \cdot \|x - x'\| + \|A^{-1}\| \cdot \|s A\| \cdot \|x - x'\|$$

$$\leq \cos(\alpha) \frac{\|x\|}{\|y\|} \left\{ \frac{\|y\|}{\|x\|} + \cos(\alpha) \frac{\|x\|}{\|y\|} \|x-y\| + \cos(\alpha) \frac{\|x\|}{\|y\|} \cdot \|x\|\right\}$$

$$(1 - \text{cond}(A) \frac{\|Ax\|}{\|x\|}) \frac{\|x - x\|}{\|x\|} \leq \text{cond}(A) \left(\frac{\|xb\|}{\|b\|} + \frac{\|A\|}{\|A\|} \right)$$

应用2. 若 x st. $\|Ax^*\|_2 \leq \|Av\|_2$, $\forall v$, 则 $\Delta = \inf_{x \in \text{dom} A} \|Ax\|_2$. 有 $\Delta = \frac{1}{\|A^{-1}\|}$

$$\boxed{1} \text{ int } J(A^*) : \exists sA \neq 0, \|sA\| = \frac{1}{\|A^{-1}\|}, \text{ st. } A + sA \text{ singular}$$

def. of norm

$$\|sA\| = \max_{\|x\|=1} \|y x^T\| = \max_{\|x\|=1} \frac{\|y x^T\|}{\|x\|} = \max_{\|x\|=1} \|y x^T\| = \max_{\|x\|=1} \|y x^T\|$$

$\|x\|=1 \Rightarrow \|x^T\|=1$

$$= \max_{\|x\|=1} \|y x^T\| = \|y\| \quad \text{而} \|x\|=1, \therefore x^T=1$$

$$\therefore \frac{1}{\|A^{-1}\|} = \frac{1}{\|A^{-1}\|} \cdot \|A^{-1}\| = \frac{\|A^{-1}\| y\|}{\|A^{-1}\|} = \frac{\|A^{-1}Ax\|}{\|A^{-1}\|} = \frac{\|Ax\|}{\|A^{-1}\|} = \frac{\|Ax\|}{\|A\|} = \frac{\|Ax\|}{\|A\|} = \frac{1}{\|A\|} = \frac{1}{\|A\|}$$

且, $A + \delta A$ singular: $\exists x^*, A x^* + \delta A x^* = y - y x^T x^* = 0$, \therefore nullity $\neq 0$,

2) If $\|A\|$ smallest, then $\frac{1}{\|A\|} \leq \|A^{-1}\|$ (and singular if $\|A\| = 0$). $(A + \delta A)x = 0$, $Ax = -\delta Ax$, $x = -A^{-1}\delta A \cdot z$. $\|z\| \leq \|A^{-1}\| \cdot \|\delta A\| \cdot \|x\|$

Ex 1. i. $\|Ax\| = 1$, $\|A\| \leq ?$ ii. constant $\|A\| = 1$ s.t. $\|Ax^*\| = ?$ (obtainable)1311. $\|x\|_\infty = \max_i |x_i|$ induced matrix norm $\|A\|_\infty = \max_j \sum_{i=1}^n |a_{ij}|$

$$\boxed{\text{P49.4} \quad \text{由 } \|A\|_\infty = \max_{1 \leq j \leq n} \|Ax\|_\infty, \text{ 且 } \|Ax\|_\infty = \max_i \left| \sum_k a_{ik} x_k \right| \leq \max_i \sum_k |a_{ik}|. \text{ 由 } \|x\|_\infty = 1 \text{ 为定数} \Rightarrow \sum_k |a_{ik}| = \max_i \sum_k |a_{ik}|}$$

$$\text{令有 } \sum_k a_{ik} = \frac{1}{\sum_{j=1}^n a_{jk}} \cdot a_{jk} > 0 \text{ 因此 } 1 \|x\|_\infty = 1. \text{ 且由 } |\sum_k a_{ik} x_k| \leq \sum_k |a_{ik}| \leq \max_i \sum_k |a_{ik}| = \sum_k |a_{ik}|. \text{ 余同}$$

$$\therefore |\sum_k a_{ik} x_k| = \sum_k |a_{ik} x_k|$$

1312. $\|x\|_1 = \sum_i |x_i|$ induced matrix norm $\|A\|_1 = \max_i \sum_{j=1}^n |a_{ij}|$ 且 $\sum_i |a_{ij}| = \max_i \sum_{j=1}^n |a_{ij}|$

$$\boxed{\text{由 P41.1. } \|A\|_1 = \sup_{\|x\|_1 \leq 1} \|Ax\|_1 = \sup_{\|x\|_1 \leq 1} \sum_i |a_{ij} x_i| \leq \sup_{\|x\|_1 \leq 1} \sum_i |x_i| |a_{ij}| \leq (\max_i |a_{ij}|) \sup_{\|x\|_1 \leq 1} \sum_i |x_i| = \sum_i |a_{ij}|}$$
1313. $\Rightarrow \|A\|_1 = 5$, $\|A^T\|_1 = 2$ 希望 $\|Ax - b\| = 10^{-6}$, $\|x - x^*\| \approx ?$

$$\boxed{\|x - b\| = \|Ax - A^T x\| \leq \|A\|_1 \cdot \|x - x^*\|, \quad \|x - x^*\| \geq \frac{10^{-6}}{5}; \text{ 而 } \|x - x^*\| = \|A^T(Ax - b)\| \leq \|A^T\|_1 \cdot \|Ax - b\| = 2 \cdot 10^{-6}}$$
1314. norm equiv P46. Tii), 7639. i) $\|x\|_\infty \leq \|x\|_2 \leq \sqrt{n} \|x\|_\infty$

$$\boxed{\text{i)} } \quad \max_i |x_i| = \sqrt{(\max_i |x_i|)^2} \leq \sqrt{|x_1|^2 + \dots + |x_n|^2} \leq \sqrt{n (\max_i |x_i|)^2}$$

$$\text{ii)} \quad \max_i |x_i| \leq |x_1| + \dots + |x_n| \leq n \cdot \max_i |x_i|$$
1314. 用 induced norm/operator norm 但 略 $\|A\|_2 = \sqrt{(\lambda_1, \lambda_2, \dots, \lambda_n)}$

$$\boxed{\text{P i) eval 5 } \quad \text{由 } \lambda_1 = 10, \lambda_2 = 1, \lambda_3 = 0.1, \lambda_4 = 0.01, \lambda_5 = 0.001, \lambda_6 = 0.0001 \Rightarrow \lambda_1 \lambda_2 \lambda_3 \lambda_4 \lambda_5 \lambda_6 = 10 \cdot 1 \cdot 0.1 \cdot 0.01 \cdot 0.001 \cdot 0.0001 = 10^{-6} \Rightarrow \|A\|_2 = 10}$$

$$\text{ii) } \text{由 } \lambda_1 = 10, \lambda_2 = 1, \lambda_3 = 0.1, \lambda_4 = 0.01, \lambda_5 = 0.001, \lambda_6 = 0.0001 \Rightarrow \lambda_1 \lambda_2 \lambda_3 \lambda_4 \lambda_5 \lambda_6 = 10^{-6} \Rightarrow \|A\|_2 = 10$$
iii) rotation stretch by factor 2, $\therefore \|A\|_2 = 2$ 应用 2. 用 ϵ approx. $Ax = b$ 有 $\frac{\|x - x^*\|}{\|x\|} \leq \|A^{-1}\| \cdot \|A\| \cdot \frac{\|A - A^*\|}{\|A\|}$, where $\hat{A} \neq A$ has perturb
$$\boxed{\text{Pii) } \quad \text{由 } Ax + (\hat{A} - A)x + A(x - x^*) = b \Rightarrow x - x^* = -A^{-1}(\hat{A} - A)x. \quad \|x - x^*\| \leq \|A^{-1}\| \cdot \|\hat{A} - A\| \cdot \|x\|}$$

$$\therefore \text{LHS} \leq \text{cond. number} \cdot \frac{\|A\|_1}{\|A\|_1} \cdot \frac{\|\hat{A} - A\|_1}{\|\hat{A} - A\|_1} \cdot \frac{\|x\|_1}{\|x\|_1} = \text{cond}(A) \cdot \|\hat{A} - A\|_1 \cdot \|x\|_1$$
Note $\|A^{-1}\|_1 \cdot \|A\|_1 \geq \|A^T A\|_1 = 1$, 且 $\forall M, \exists A$ st. $\text{cond}(A) \geq M$

$$\boxed{\text{Piii) } \quad \text{consider } \epsilon < \frac{1}{M} \text{ 令 } A = \begin{pmatrix} 1 & 0 \\ 0 & \epsilon \end{pmatrix} \Rightarrow \|A\|_2 = 1 \quad \text{s.t. } \text{cond}(A) = \frac{1}{\epsilon} > M}$$

$$\text{设 } x = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \text{ 则 } \|x\|_2 = \|Vx\|_2, \text{ 但 } \|Vx\|_2 \text{ 为 } V \text{ 的 norm, 即 induced } \|A\|_2 = \|VAV^{-1}\|_2$$

$$\boxed{\text{Piv) } \quad \text{WTS } \forall \|x\|_2 = \|Vx\|_2 = 1 \Rightarrow \|Ax\|_2 = \|VAV^{-1}x\|_2. \text{ 但由 } \|VAV^{-1}\|_2 \text{ det. } \forall \|z\|_2 = 1 \text{ 有}$$

$$\|VAV^{-1}z\|_2 \leq \|VAV^{-1}\|_2 \text{ 因此 } z = Vx \Rightarrow \|VAV^{-1}Vx\|_2 = \|VAx\|_2 = \|VAU^{-1}\|_2$$

$$\therefore \text{由 } \|VAV^{-1}\|_2 \text{ det. 有 } \|VAV^{-1}z\|_2 = \|VAU^{-1}\|_2 \cdot \|U^{-1}z\|_2 = \|VAU^{-1}\|_2 \cdot \|U^{-1}z\|_2$$

$$\boxed{\text{Pvi) } \quad \text{cond}(A) = \max_{\|y\|=1} \frac{\|Ay\|}{\|Ay\|} \Rightarrow \|A\|_1 \cdot \|A^{-1}\|_1 = \max_{x \neq 0} \frac{\|Ax\|}{\|Ax\|} \cdot \max_{y \neq 0} \frac{\|A^{-1}y\|}{\|A^{-1}y\|} = \max_{x \neq 0} \|Ax\| \cdot \max_{y \neq 0} \frac{\|A^{-1}y\|}{\|A^{-1}y\|} = \frac{\max_{x \neq 0} \|Ax\|}{\min_{y \neq 0} \|A^{-1}y\|} = \frac{\max_{x \neq 0} \|Ax\|}{\min_{y \neq 0} \|y\|}$$

$$\text{2. 2nd perturbed } \tilde{x} \quad (Ax = b) \text{ 有 } \frac{\|b - \tilde{b}\|}{\|b\|} \leq \frac{1}{\|A\|_2} \leq \frac{\|x - \tilde{x}\|}{\|x\|} \leq \text{cond}(A) \frac{\|b - \tilde{b}\|}{\|b\|}$$

$$\boxed{\text{Pvii) } \quad \frac{\|A\|_1 \cdot \|x - \tilde{x}\|}{\|A\|_1 \cdot \|x\|} > \frac{\|Ax - A\tilde{x}\|}{\|A\|_1 \cdot \|A^{-1}\|_1} > \frac{\|b - \tilde{b}\|}{\|A\|_1 \cdot \|A^{-1}\|_1 \cdot \|b\|}}$$

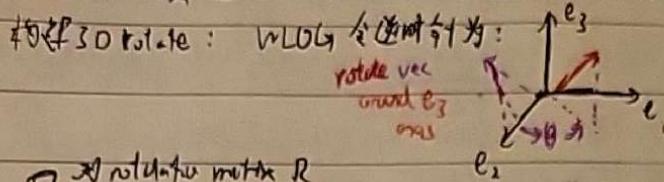
$$\boxed{\text{Pviii) } \quad \text{3. } \text{cond}(A) = \begin{cases} \frac{\lambda_{\max}}{\lambda_{\min}}, A \text{ pd.} \\ \infty, A \text{ not pd.} \end{cases}}$$

$$\begin{aligned} &\text{(in 2-norm)} \quad \begin{cases} \frac{\lambda_{\max}}{\lambda_{\min}}, A \text{ pd. evl. } \lambda_1 \geq \dots \geq \lambda_n > 0 \\ \infty, A \text{ not pd. evl. } \lambda_1 \geq \dots \geq \lambda_n > 0 \end{cases} \quad \therefore \|A\|_2 = \lambda_1 \\ &\quad \lambda_1, \lambda_2, \dots, \lambda_n \text{ 非零} \Rightarrow \lambda_1^2 \geq \dots \geq \lambda_n^2 > 0 \quad \therefore \|A\|_2 = \lambda_1 \\ &\quad 1, A \text{ unitary } \Rightarrow \lambda_1^2 = 1 \quad \therefore \|A\|_2 = 1 \end{aligned}$$

Rotation

Date:

$$1. \text{ 一般 2D 旋转 } \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix}$$



$$\left\{ \begin{array}{l} \text{rotate around } e_3 \text{ axis} \\ e_1' = \begin{pmatrix} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{pmatrix} \\ e_2' = \begin{pmatrix} \cos\theta & 0 & 0 \\ 0 & \cos\theta & 0 \\ -\sin\theta & 0 & 0 \end{pmatrix} \end{array} \right.$$

2. P 为 rotation matrix R

$$1) (\text{preserve distance}) \|x-y\| = \|(Rx-Ry)\|$$

$\square \text{ 由 4) \& P_{ij}, P_{ii}}$

$$2) (\text{preserve angle}) \cos(x, y) = \cos(Rx, Ry)$$

$$\square \frac{x^T y}{\|x\| \|y\|} = \frac{x^T R^T Ry}{\|x\| \|y\|} = \frac{x^T y}{\|x\| \|y\|}$$

3) \nexists commute \dots rotate order matter, $\underline{\text{1020}}$ rotation commutative

$$4) |\det R| = 1 \quad \square \text{ 由 4) \& P_{ij}, P_{ii}}$$

$$4) \text{ rotation } \Rightarrow \text{orthogonal}$$

$$\square \text{ 由 2D } Q = \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \quad Q^T Q = I$$

5) $\det R = 1 \quad \square \text{ 由 4) \& P_{ij}, P_{ii}}$ $\det \text{ linear } T \neq 0 \Rightarrow \text{det } T \neq 0 \quad \square \text{ 7. scale area}$

3 Householder Reflection

Gram-Schmidt num stable (cond(A) large) $\exists j: 1 \leq j \leq n, a_j \in \langle a_{1 \dots j-1} \rangle, \text{ from } \eta: a_j - \sum \frac{c_{ij} a_j}{\|a_j\|^2} \text{ di } \eta \text{ cancel terms}$
 $\therefore \text{reflect } a_j: \text{reflect } a_j \text{ st. } a_j \in \langle a_{1 \dots j-1} \rangle^\perp$ (numerically) \Rightarrow cancel error

1° def $P_n = I - 2vv^T$ where $\|v\|=1$, \Rightarrow reflection P householder Transform

$$P_i) P \text{ orthogonal } \quad P^T = I - 2vv^T = P; \quad P^T P = P^2 = I - 4vv^T + 4(vv^T)(vv^T) = I$$

ii) P reflect x to $n-1$ dim subspace which $\perp v$

$$\square \text{ 因 } P^T x = \text{proj}_v x \quad \therefore P x = x - v v^T x \quad \square$$

upper triangle [0]

2° (QR decmp) \Rightarrow reflection $\{P_i\}$ s.t. $P_n \cdots P_1 A = R$, $R = P_1^T \cdots P_n^T$, $\forall i P_i \text{ 只对 } R \text{ 中 } 1 \times 1$

$$P_1 P_1 A = \begin{pmatrix} r_{11} & 0 & \dots \\ 0 & \ddots & \vdots \\ \vdots & \vdots & \ddots \end{pmatrix}, P_2 P_1 A = \begin{pmatrix} r_{11} & r_{12} & \dots \\ 0 & r_{22} & \vdots \\ \vdots & \vdots & \ddots \end{pmatrix} \quad \text{reflect } a_1 \quad \text{②} \quad \text{③}$$

$$1° P_1: \quad P_1 \vec{a}_1 = r_{11} \vec{e}_1 = a_1 - 2v_1 v_1^T a_1, \quad v_1 = \frac{a_1 - r_{11} e_1}{\|a_1 - r_{11} e_1\|}, \quad \text{so } \|P_1 \vec{a}_1\|_2 = \|a_1\|_2 = \|r_{11} e_1\| = \|r_{11}\|_1$$

$$\therefore \text{pick } r_{11} = -\text{sgn}(\vec{a}_1) \cdot \|a_1\|_2, \quad v_1 \equiv \frac{a_1 - r_{11} e_1}{\|a_1 - r_{11} e_1\|}, \quad \text{denote } P_1 A = \begin{pmatrix} r_{11} & r_{12} & r_{13} & \dots & r_{1n} \\ 0 & \ddots & & & \\ a_2 & A_2 & & & \\ \vdots & & \ddots & & \\ a_n & A_n & & & \end{pmatrix}$$

$$2° P_2: \quad P_2(P_1 A)_{x_1} = \begin{pmatrix} r_{11} \\ 0 \\ \vdots \end{pmatrix} = (I - 2V_1 V_1^T) \begin{pmatrix} r_{11} \\ r_{12} \\ \vdots \end{pmatrix} \Rightarrow A_1 \downarrow I - 2V_1 V_1^T = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & \ddots & & \\ \vdots & & \ddots & 0 \\ 0 & 0 & \dots & 1 \end{pmatrix}$$

$$\text{是 } b P_2 = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & \ddots & & \\ \vdots & & \ddots & 0 \\ 0 & 0 & \dots & 1 \end{pmatrix}$$

$$\text{for } s.t.: \therefore r_{22} = -\text{sgn}(\vec{a}_2) \cdot \|a_2\|_2, \quad v_2 \equiv \frac{a_2 - r_{22} e_1}{\|a_2 - r_{22} e_1\|_2}, \quad \text{denote } P_2 P_1 A = \begin{pmatrix} r_{11} & r_{12} & r_{13} & \dots & r_{1n} \\ 0 & r_{22} & r_{23} & \dots & r_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix}$$

$$3° \text{ 由 } r_{33} = -\text{sgn}(\vec{a}_3) \cdot \|a_3\|_2, \quad v_3 \equiv \frac{a_3 - r_{33} e_1}{\|a_3 - r_{33} e_1\|_2}$$

Date.

15/2

Ex 1.1 $A = \begin{pmatrix} 0 & -4 \\ -5 & -2 \end{pmatrix}$ $b = \begin{pmatrix} -8 \\ 6 \end{pmatrix}$ \therefore sol $x = (A^T A)^{-1} A^T b = \begin{pmatrix} -\frac{7}{2} \\ 5/4 \end{pmatrix}$. Prove by QR using Gram

QGram $\left(\frac{q_1}{a_1}, \frac{q_2}{a_2} \right)$: 1° $u_1 = a_1$, $q_1 = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$ 2° $u_2 = a_2 - P_{u_1}(a_2) = \begin{pmatrix} -4 \\ 2 \end{pmatrix} - r_{12} \cdot q_1 = \begin{pmatrix} 4 \\ 0 \end{pmatrix}$
 $r_{11} = \|u_1\| = 5$ $r_{12} = q_1^T \cdot a_2 = 2$ $q_2 = \begin{pmatrix} -1 \\ 0 \end{pmatrix}$ $r_{22} = \|u_2\| = 4$

 $\therefore R = \begin{pmatrix} 5 & 2 \\ 0 & 4 \end{pmatrix}$ $Q = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \quad \text{Pf. } q_1, q_2 \perp$

Ex 1.2 Householder 1° $r_{11} = -\text{sign}(0) \|a_1\| = 5$ $v_1 = \frac{\begin{pmatrix} 0 \\ -5/2 \end{pmatrix}}{\|..||} = \begin{pmatrix} 0 \\ -1/\sqrt{2} \end{pmatrix}$ $P_1 = I - 2v_1 v_1^T = \begin{pmatrix} 0 & 0 & -1 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \end{pmatrix}$, $P_1 A = \begin{pmatrix} 0 & 0 \\ 0 & 4 \end{pmatrix}$

2° $r_{22} = -\text{sign}(-4) \|a_2\| = 4$, $v_2 = \frac{\begin{pmatrix} 0 \\ 4-4/2 \end{pmatrix}}{\|..||} = \begin{pmatrix} 0 \\ 1/\sqrt{2} \end{pmatrix}$ $P_2 = I - 2v_2 v_2^T = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$

 $\therefore R = \begin{pmatrix} 5 & 2 \\ 0 & 4 \end{pmatrix}$ $Q = P_1^T P_2^T = \begin{pmatrix} 0 & -1 & 0 \\ 0 & 0 & 1 \\ -1 & 0 & 0 \end{pmatrix}$ $\therefore \boxed{1.2}$

Jordan

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Date. /

$$\begin{pmatrix} -1 & 0 \\ -3 & \lambda_2 \end{pmatrix}$$

1. $\lambda = 2 \pm i$ similar to $\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$ L.Ind over $\mathbb{C}[T]$



x $\xrightarrow{\text{shear}}$ $x + v$ $\xrightarrow{\text{scale}}$ $\lambda x + v$

i) $\begin{pmatrix} \lambda_1 & 1 \\ 0 & \lambda_2 \end{pmatrix}$ λ_1 algebraic multi 2 \Rightarrow $\lambda_1 x + v$, L.Ind, $Ax = \lambda_1 x + v$.
 ii) $\begin{pmatrix} 3 & 1 \\ -1 & 5 \end{pmatrix}$ geometric \Rightarrow $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ \Rightarrow $\{v, x\}$ is basis

iii) $r \begin{pmatrix} \cos b & -\sin b \\ \sin b & \cos b \end{pmatrix} = \begin{pmatrix} a & -b \\ b & a \end{pmatrix}$ complex eval, $\begin{pmatrix} 2 & -2 \\ 2 & 2 \end{pmatrix}$

2×3 to \mathbb{C} form $\begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix} \begin{pmatrix} \lambda_1 & 1 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix} \begin{pmatrix} \lambda_1 & 1 & 0 \\ 0 & \lambda_2 & 1 \\ 0 & 0 & \lambda_3 \end{pmatrix} \begin{pmatrix} a & -b & 0 \\ b & a & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix}$

$$\text{Ex 1. } \begin{cases} x_1 + x_2 + 2x_3 + x_4 = 1 \\ x_1 - 2x_2 - x_4 = -2 \end{cases} \text{用 G.E.}$$

$$\boxed{\text{解 }} \text{rk} \begin{bmatrix} 1 & 1 & 2 & 3 \\ 1 & -2 & 0 & -1 \end{bmatrix} = \text{rk} \begin{bmatrix} 1 & 1 & 2 & 3 & 1 \\ 1 & -2 & 0 & -1 & -2 \end{bmatrix}, \therefore \exists \text{ sol. assign } \frac{x_2}{x_4} \text{ random, 有 } \begin{bmatrix} 1 & 1 \\ 1 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 2 \\ -2 + x_4 \end{bmatrix}$$

thus $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \text{ 为 } \begin{bmatrix} 1 & 1 \\ 1 & -2 \end{bmatrix}^{-1} \begin{bmatrix} 2 \\ -2 + x_4 \end{bmatrix} = \begin{bmatrix} -\frac{1}{3}x_3 - \frac{x_4}{3} \\ 1 - \frac{2}{3}x_3 - \frac{2}{3}x_4 \end{bmatrix}$ general sol. $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} \frac{2}{3} \\ \frac{-2}{3} \\ \frac{x_3}{3} \\ x_4 \end{bmatrix}$

应用 (minus 1 rank) 领域化为 REF, 并在 missing pivot 为 (0...-1..0), sol 为 -1 的情况

如 Ex 2 $A = \begin{pmatrix} 1 & 0 & 9 & 8 \\ 0 & 1 & 4 & 5 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$ 分为 $a \begin{pmatrix} 1 & 0 & 9 & 8 \\ 0 & 1 & 4 & 5 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} + b \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$

1. Gaussian Elimination is stable.

eg. $\begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \end{pmatrix} \Rightarrow \begin{pmatrix} x_1 & x_2 & x_3 \\ y_1 & y_2 & y_3 \end{pmatrix} \text{ (cond error)}$ 1. 1 是 big pivot 且其乘积相对大小

\therefore sol 为 2 位小数 pivot: $\Rightarrow \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \end{pmatrix} \Rightarrow \begin{pmatrix} x_1 & x_2 & x_3 \\ y_1 & y_2 & y_3 \end{pmatrix} \approx 1$

2. 由此为启发, impl Gaussian Elimination w/ pivoting, 使其大数问题也 stable

Ex 2 partial pivot $\begin{pmatrix} 6 & -2 & 3 & 4 & | & 16 \\ 12 & -8 & 6 & 10 & | & 26 \\ 3 & -13 & 9 & 3 & | & -19 \\ -6 & 4 & 1 & -16 & | & 34 \end{pmatrix} \Rightarrow \begin{pmatrix} 12 & -8 & 6 & 10 & | & 26 \\ 6 & -2 & 2 & 4 & | & 26 \\ 3 & -13 & 9 & 3 & | & -19 \\ -6 & 4 & 1 & -16 & | & 34 \end{pmatrix} \sim \begin{pmatrix} 12 & -8 & 6 & 10 & | & 26 \\ 0 & 2 & 4 & 1 & | & 26 \\ 0 & -11 & 9 & 3 & | & -19 \\ 0 & 0 & 4 & -16 & | & 34 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

$\text{max abs} = \text{row 2 swap} \quad \text{max abs} = \text{row 3 swap}$

 $\sim \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \Rightarrow \begin{pmatrix} 12 & -8 & 6 & 10 & | & 26 \\ 0 & -11 & 15 & 1/2 & | & 25/2 \\ 0 & 0 & 4 & -13 & | & -21 \\ 0 & 0 & 0 & 3/11 & | & 3/11 \end{pmatrix} \text{ sol 同上施密斯消元}$

Note: 在 partial pivot swap along process, 要预先知道那几行 swap, 先 swap 互换元 sol 两边端元边 swap, RP 不可 permute

(min)

2. 对 $AX=b$, $A = \begin{pmatrix} \bar{A} & 0 \\ 0 & I_m \end{pmatrix}$, $b = \begin{pmatrix} c \\ d \end{pmatrix}$ 时 least square sol 同 $\bar{A}x=c$, 因:

假设 A 为 full form, $\sqrt{\text{change of basis}} P^{-1}PA = \begin{pmatrix} A & 0 \\ 0 & I_m \end{pmatrix}$, 有 $\|b - Ax\|_2^2 = \|P(b - Ax)\|_2^2 = \|P(b - Ax)\|_2^2 = \|P(b - Ax)\|_2^2$

$$= \|\begin{pmatrix} c - \bar{A}x \\ d \end{pmatrix}\|_2^2 = \|c - \bar{A}x\|_2^2 + \|d\|_2^2, \text{ min when } \bar{A}x=c$$

eg. $A = \begin{pmatrix} 1 & 0 & 1 \\ 3 & 3 & 2 \\ 3 & 3 & 4 \\ -1 & 6 & 3 \end{pmatrix}$ $b = \begin{pmatrix} 4 \\ 2 \\ 2 \\ 1 \end{pmatrix}$ $\text{rank}(A) = \text{rank}(\bar{A}) = 3$ $\text{rank}(A^T) = \text{rank}(\bar{A}^T) = 3$ $P = P^T = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$ $\text{IR}^2 = \text{RA} \oplus N(\bar{A})$ $\text{base } \bar{A}$

$\therefore P^T b = \begin{pmatrix} 2.64 \\ -1.13 \\ -3.32 \\ 2.06 \\ 4.58 \end{pmatrix}$ $\text{NOTE 实际得记算 } N(\bar{A}^T), \text{ 令 } P = \begin{pmatrix} q_1 & q_2 & q_3 & q_4 & q_5 \\ p_1 & p_2 & p_3 & p_4 & p_5 \end{pmatrix} = \begin{pmatrix} P \\ Q \end{pmatrix}$

有 \bar{A} 上 $\triangle \tilde{P}^T A = \tilde{A} = \begin{pmatrix} q_1 & q_2 & q_3 & q_4 & q_5 \\ p_1 & p_2 & p_3 & p_4 & p_5 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$

可 QR 実行 QR decmp: $PA = R = \begin{pmatrix} \bar{R} & 0 \\ 0 & I_m \end{pmatrix}$, $A = P^{-1}R = P^T R = 0$ by (inner prod)

3. (LS w/ QR)

$\|Ax-b\| = \|(QRx-b)\| = \|(Rx-Q^Tb)\| = \|\begin{pmatrix} \bar{R} & 0 \\ 0 & I_m \end{pmatrix}x - \begin{pmatrix} \bar{Q}^T & 0 \\ 0 & I_m \end{pmatrix}b\| = \|\begin{pmatrix} \bar{R}x - \bar{Q}^Tb \\ 0 \end{pmatrix}\| = \|\bar{R}x - \bar{Q}^Tb\| + \|0\| = \|\bar{R}x - \bar{Q}^Tb\|$

Sys of Equations I

Date: / /

1. $\exists A_{mn}, Ax=b \exists \text{ sol} \Leftrightarrow \text{rk } A = \text{rk } [A|b]$
- $\square \Rightarrow b = L.$ (as A 's col = $\sum_{i=1}^n x_i a_i$, thus $b \in \langle a_1 \dots a_n \rangle$, $\text{rk } A = \dim \langle a_1 \dots a_n \rangle = \dim \langle a_1 \dots a_n, b \rangle = \text{rk } [A|b]$)
- \Leftarrow 令 $\text{rk } A = r$ 则 A 有 r L. ind col $a_1 \dots a_r$, $[A|b]$ 也有 r 个 col, $\therefore b$ 可由 $a_1 \dots a_r$ L. 表示.
- NOTE, x unique $\Leftrightarrow \text{rk } A = \text{rk } [A|b] = n$ (即 $b = \sum_{i=1}^n x_i a_i$, 即 x 为 \vec{x} sol)
- $\square \Rightarrow$ 令 $k < n$, 则 $\exists y \neq 0$ s.t. $Ay = 0$, 且 $x \neq y$ sol, $x+y$ 也 sol. \therefore 无解.
- \Leftarrow 由 col, 若 x 为唯一, $Ax = Ay = b$, $A(x-y) = b$ 可知 A 有 n L. ind col, $\therefore x = A^{-1}b$ (AP 例 7.4)

2. $A_{mn}, m=n, \text{rk } A = m, Ax=b$ 有 sol \Leftrightarrow 有 $m-n$ 个 random, 且 m 个 ind col
- $\square \Leftarrow$ 令 $a_1 \dots a_m$ L. ind $Ax=b \Rightarrow x_1 a_1 + \dots + x_m a_m = b - x_{m+1} a_{m+1} - \dots - x_n a_n$, 令 $\begin{cases} x_{m+1} \in \Phi_{m+1} \\ \vdots \\ x_n \in \Phi_n \end{cases}$ 为 random
- $\therefore \begin{bmatrix} a_1 & \dots & a_m \\ \vdots & & \vdots \\ a_m & & \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} = \dots, \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} = B^{-1} [b - \varphi_{m+1} a_{m+1} - \dots - \varphi_n a_n]$
- (见 P11.1.2, P11.2.1, P11.2.2)
3. (least square) $\min_x f(x) = \|b - Ax\|_2^2 \Leftrightarrow A_m$ 有
- $m=n$ 且 A 可逆. 则 \exists sol $x^* = A^{-1}b, f(x^*) = 0$ (note $\exists m=n x^* \in \text{ker } A^T$)
 - $m > n$, $f(x) = \|b - Ax\|_2^2$ 无解. 见 P11.3
 - $A^T A$ 为正定, $\therefore A^T A x^* = A^T b$ (见 P11.3, P11.2.2, P11.2.3)
 - $A^T A$ 为正定: $x^* = (A^T A)^{-1} A^T b = A^{-1}b$
 - \nexists ! x^* 为唯一 sol, 但有 $\exists f(x^*)$ (P11.2.2)

- \square 由 P11.1.3 及 $y = 0$; 令 x^* 为 $f(x)$ 的极小值点 $\Leftrightarrow A^T A x^* = A^T b$ 由 orthogonal decomposition
- $\Rightarrow b = c + d$ 由 P11.2.2, 令 $d \perp Ax - c$, $f(x) = \|c + d - Ax\|_2^2 = \|c - Ax\|_2^2 + \|d\|_2^2$, $x^* \in Ax = c$ 为唯一
- 令 x^* 为 $A^T A x = A^T c + 0 = A^T(c+d) = A^T b$ ($\text{ker } A^T \cap \text{ker } A = \{0\}$) $\Leftrightarrow A^T A x^* = A^T(c+d) = A^T c$
- $\therefore A^T(Ax^*-c) = 0$, $\therefore Ax^*-c \in N(A^T)$ ($N(A^T) = \{0\}$) $\therefore Ax^*-c = 0$
- NOTE** $\begin{cases} \text{若 } A \text{ full-rank} \\ \text{或 } A \text{ rank } n \end{cases} \Rightarrow \text{Pseudoinverse: } \exists B = A^T A, y = A^T b, \text{ 令 } Bx = y, \text{ 令 } B \text{ 为 } \text{Pseudoinverse}$ (P11.2.2, P11.2.3)
- 1° Cholesky decom $B = G G^T$ 2° solve $G^T x = y$ by forward sub 3° solve $G x = z$ by backward sub

- P11.1. (Solve normal equation by numerical method)
- \Rightarrow nonsquare A_{mn} ($m > n$) def¹ $\text{cond}(A) = \|A\|_2 / \|A^T\|_2 = \frac{\|A\|_2}{\|A^T\|_2} = \frac{\|A^T b\|_2}{\|A\|_2} = \frac{\|x\|_2}{\|A\|_2} \leq \text{cond}(A) \frac{\|b\|_2}{\|A\|_2}$ (relative error bound by cond(A))
- $\therefore \text{cond}(A) = \frac{\|A\|_2}{\|A^T\|_2}$ 为 least square: points \tilde{x} 为 sol $\tilde{x} \in \frac{\|x\|_2}{\|A\|_2} \leq \text{cond}(A) \frac{\|b\|_2}{\|A\|_2}$
- \square 1° SVD $A = U \begin{pmatrix} \Sigma & 0 \\ 0 & 0 \end{pmatrix} V^T$, $\therefore \|A\|_2 = \sqrt{\sum \Sigma_i^2} = \sigma_1$, 令 $A^T = (V \Sigma^T U^T)^{-1} V \Sigma^T U^T$
- $= (V \Sigma^T U^T)^{-1} V \Sigma^T U^T = V \Sigma^{-1} U^T$
- $\therefore \|A^T\|_2 = \sqrt{\sum \Sigma_i^2} = \sigma_1$ (SVD) $\therefore \|A^T\|_2 = \sqrt{\sum \Sigma_i^2} = \sigma_1$ (SVD)
- $\therefore \|x - \tilde{x}\|_2 \leq \|A^T\|_2 \cdot \|b - \tilde{b}\|_2$ 且 $b = c + d$, $Ax = c$, $\therefore \|x\|_2 \leq \|A\|_2 \cdot \|x\|_2$ (deli 得力)

1. elementary row operation

elementary

1) non scaling: - row \times scalar. $A_{i,:} \rightarrow \alpha A_{i,:}$

2) non exchange: \leftrightarrow rows. $A_{i,:} \leftrightarrow A_{j,:}$

3) non adjustment: - row \times scalar $+ g\text{-row}$. $A_{j,:} \rightarrow A_{j,:} + \alpha A_{i,:} - i \neq j$

Gaussian Elimination $n \times n$ matrix
 (A_{ij}) cost is row $n \rightarrow \frac{1}{2}n(n-1)$
 scaling adjoint \neq operation.
 If row \times scalar \neq scaling \Rightarrow T. $n(n-1)$ cost
 adjoint $\sim n(n-1)$
 $\Rightarrow E_{1:n} \cdot \text{operation} R \sim k \cdot \frac{n(n-1)}{2} = \frac{n(n-1)}{2} = \frac{n^2-n}{2}$

2. matrix $R \rightarrow$ in row echelon form iff

i) 只有 0 行 $\neq R$ 底部, ii) $\neq 0$ 行, first nonzero entry $\neq 1$, iii) $\forall 2 \leq i \leq n$, 第 i 行 leading entry \neq

row echelon form \neq reduced row echelon form iff A col has leading entry 在其他 position 0

例 2.1.

3. $A, B \in M_{m,n}(F)$, A from equivalent to $A \rightarrow B$ by elementary row operation $\Leftrightarrow B$.

且 $A \sim B$ 为 equivalent relation elementary matrix

且, $T_1: A \sim B$, 2) $B = EA$; 因 row operation \Rightarrow column operation, $B = AE'$

□ row operation: 1) $E = \begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{bmatrix}$ 2) $E = \begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{bmatrix}$ 3) $E = \begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{bmatrix}$

由 $B_{i,:} = (EA)_{i,:} = E_{i,:} A$ 看出, 且 E 由 $A \rightarrow B$ 为操作于 I_n 上 (是 \neq)

且 col operation 1) $E = \begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{bmatrix}$ 2) $E = \begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{bmatrix}$ 3) $E = \begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{bmatrix}$

由 $B_{:,j} = AE_{:,j}$ 看出

易知当 A row $[col] \rightarrow B$, A^T 对应 $(col)[row] \rightarrow B^T$ □ $B = EA \rightarrow B^T = A^T E^T$

T_2 : E (elementary mat) invertible, 且 invertible type

row $\xrightarrow{\text{row}} [col]$ (不是 $[row]$)

且 $WLOG$, E 由 row op type: 得到, 由 T_1 可以对 I_n 的 row type: $I_n \sim E I_n \dots$ 有 $E, I_n = E' E$

且 prod of elem mat elem $\times \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ sum of ... elem $\times \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$

且 transpose of ... elem \checkmark type 1, 2 symmetric \checkmark type 3 row \times type 3 col \sim $A \sim B$ by $[row] \rightarrow A \sim B$ by $[col]$ $= \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$

$A \sim B$ by $\xrightarrow{\text{row}} B \sim A$ by $\xrightarrow{\text{row}}$ $\times \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ by $\xrightarrow{\text{row}}$, 但不同时行

T_2 用于求简单的逆矩阵 如 $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$, $\begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$, $\begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$ (undo op)

例 $A = \begin{pmatrix} 2 & 1 & 1 & 4 \\ 3 & 1 & 2 & 3 \\ 1 & 2 & 1 & 0 \end{pmatrix}$ $B = \begin{pmatrix} 2 & 1 & 1 & 4 \\ 1 & 1 & 0 & -5 \\ 1 & 2 & 1 & 0 \end{pmatrix}$ $C = \begin{pmatrix} 3 & 5 & 2 & 4 \\ -1 & -1 & 0 & -5 \\ 1 & 2 & 1 & 0 \end{pmatrix}$ \neq st. $A = FC$

□ $A = EB$, note 要找 $B = ?$ C \neq st. $F = E?$, $F = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$, $? = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 3 \\ 0 & 0 & 1 \end{pmatrix}$ $\therefore F = EB?$

Sys Eq 2

$$\begin{array}{l} \text{1. } x_1 - 2x_2 + x_3 = 0 \\ \quad \rightarrow 2x_2 - x_3 = 8 \\ \quad - 4x_1 + 5x_2 + 9x_3 = -9 \\ \quad \Downarrow A \cdot X = b \end{array} \Rightarrow \begin{array}{l} \text{coefficient matrix} \\ \left[\begin{array}{ccc|c} 1 & -2 & 1 & 0 \\ 0 & 2 & -1 & 8 \\ -4 & 5 & 9 & -9 \end{array} \right] \\ \text{augmented matrix} \\ \left[\begin{array}{ccc|c} 1 & -2 & 1 & 0 \\ 0 & 2 & -1 & 8 \\ -4 & 5 & 9 & -9 \end{array} \right] \end{array}$$

then, row reduction $\left[\begin{array}{ccc|c} 1 & -2 & 1 & 0 \\ 0 & 2 & -1 & 8 \\ -4 & 5 & 9 & -9 \end{array} \right] \xrightarrow{\text{echelon form}} \left[\begin{array}{ccc|c} 1 & 0 & 0 & 29 \\ 0 & 1 & 0 & 16 \\ 0 & 0 & 1 & 3 \end{array} \right] \xrightarrow{\text{back substitution}} \left[\begin{array}{ccc|c} 1 & 0 & 0 & 29 \\ 0 & 1 & 0 & 16 \\ 0 & 0 & 1 & 3 \end{array} \right]$
 Row 1st 1st leading entry

例 1. $\begin{cases} x_1 + 3x_2 = 1 \\ 2x_1 + dx_2 = y \end{cases}$ 1. 2nd col st. consistent (无解) 3. 1st col st. inconsistent (无解)

$$\square \left[\begin{array}{cc|c} 1 & 3 & 1 \\ 2 & d & y \end{array} \right] \sim \left[\begin{array}{cc|c} 1 & 3 & 1 \\ 0 & d-6 & y-2 \end{array} \right] \quad \text{if } d=6 \text{, } y=2 \text{, unique soln} \quad \text{if } d \neq 6 \text{, no soln}$$

\therefore 若有 raw $\left[\begin{array}{ccc|c} 0 & \dots & 0 & 1 \\ 0 & \dots & 0 & b \end{array} \right]$ 不一致, 则不一致 $\left\{ \begin{array}{l} \text{无解, unique soln if } rkA = n \\ \text{no soln if } rkA < n \end{array} \right.$

不同的基代表的东西不一样. $\left[\begin{array}{c} 1 \\ 0 \\ 0 \end{array} \right]$ 和 $\left[\begin{array}{c} 0 \\ 1 \\ 0 \end{array} \right]$ 与 $\left[\begin{array}{c} 0 \\ 0 \\ 1 \end{array} \right]$ 不一样. 因此坐标轴一定由坐标基 L. Comb.

2. linear combinations. given $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p$ in \mathbb{R}^n , given scalars c_1, c_2, \dots, c_p , 在一维空间上运动

vector \vec{y} defined by $\vec{y} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_p \vec{v}_p$

c_1, \dots, c_p : weight, 可写为 $[c_1, c_2, \dots, c_p]^T$

Span if $\vec{v}_1, \dots, \vec{v}_p$ in \mathbb{R}^n , set of all linear combinations

of $\vec{v}_1, \dots, \vec{v}_p$ is denoted by $\text{span}\{\vec{v}_1, \dots, \vec{v}_p\} = \{c_1 \vec{v}_1 + \dots + c_p \vec{v}_p \mid c_i \in \mathbb{R}\}$

3. matrix equation $Ax = b$.

$$[\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n] \left[\begin{array}{c} x_1 \\ x_2 \\ \vdots \\ x_n \end{array} \right] = x_1 \vec{a}_1 + x_2 \vec{a}_2 + \dots + x_n \vec{a}_n \quad (\vec{a}_n \text{ 不是 } 3.1/2 \text{ 里})$$

$\exists A \in \mathbb{R}^{m \times n}$, the following statement are equivalent:

a. $\forall b \in \mathbb{R}^m$, $Ax = b$ has a solution (不唯一)

b. $\forall b \in \mathbb{R}^m$ is a linear combination of A (线性组合)

c. the columns of A span \mathbb{R}^m ($\exists m \times n, \text{ full rank}$)

d. A has a pivot position in each row

例 1. could a set of n vectors in \mathbb{R}^m span all of \mathbb{R}^m when $n < m$?

\square let $A = [v_1 \ v_2 \ \dots \ v_n]$ then d. A ^{cannot} have a pivot position 得力

Date:

$$\text{Ex } \begin{bmatrix} 5 & 1 & -8 & 4 \\ -2 & 7 & 3 & -5 \end{bmatrix} \begin{bmatrix} -5 \\ 1 \\ 3 \\ -2 \end{bmatrix} = \begin{bmatrix} 5 \cdot -5 + 1 \cdot 24 - 8 \\ -2 \cdot 5 + 7 \cdot 9 + 10 \end{bmatrix} = \begin{bmatrix} -8 \\ 16 \end{bmatrix}$$

$$2^{\circ} \begin{bmatrix} 5 \\ -2 \end{bmatrix} \cdot 5 + \begin{bmatrix} 1 \\ 7 \end{bmatrix} \cdot (-1) + \begin{bmatrix} -8 \\ 3 \end{bmatrix} \cdot 3 + \begin{bmatrix} 4 \\ -5 \end{bmatrix} \cdot (-2) = \begin{bmatrix} -8 \\ 16 \end{bmatrix}, \text{ 例 } Ax = \begin{bmatrix} a_1 & \dots & a_n \end{bmatrix} \cdot x = \sum a_i x_i$$

$\{ \sin(x), \cos(x), 1 \} \subseteq \mathcal{F}(\mathbb{R}, \mathbb{R})$ L.Ind

$$\boxed{\begin{cases} x=0 & a_1+a_2=0 \\ x=\frac{\pi}{2} & a_1+a_2=0 \\ x=\frac{\pi}{4} & a_1+a_2=0 \end{cases} \Rightarrow a_1=a_2=0}$$

SD/Wank w/

Ex 2 describe all sols of $Ax=0$ in parametric vector form

$$\boxed{A = \begin{bmatrix} 1 & 3 & -3 & 7 \\ 0 & 1 & -4 & 5 \end{bmatrix}, \begin{bmatrix} 1 & 3 & -3 & 7 & 0 \\ 0 & 1 & -4 & 5 & 0 \end{bmatrix} \sim \begin{bmatrix} 0 & 0 & 9 & -8 & 0 \\ 0 & 1 & -4 & 5 & 0 \end{bmatrix}, \text{ 3 pivot position}} \quad \begin{cases} x_1 = -9x_3 + 8x_4 \\ x_2 = 4x_3 - 5x_4 \end{cases} \rightarrow x_3, x_4 \text{ free var}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -9x_3 + 8x_4 \\ 4x_3 - 5x_4 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} -9 \\ 4 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} 8 \\ -5 \\ 0 \\ 1 \end{bmatrix}$$

L.Ind 何谓线性无关?

$\begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array}$ $\begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array}$

$$\begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array}$$

$$\begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array}$$

$$\begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array}$$

$$\begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array}$$

$$\begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nearrow \text{L.dep} \\ \downarrow \end{array} \quad \begin{array}{c} \nwarrow \text{L.dep} \\ \downarrow \end{array}$$

Ex 3 A is 3×2 with 2 pivot positions. $\text{Do } Ax=0$ has nontrivial sol? $\text{Do } Ax=b$ has at least one sol for every b ?

$$\boxed{A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \quad \begin{array}{l} x_1=0 \\ x_2=0 \end{array} \quad \begin{array}{l} 1. \text{ sol trivial} \\ 2. \text{ no pivot position each row. N.O} \end{array}}$$

Ex 4 A 7×5 , if columns L.Ind, how many pivot positions

$\boxed{5}$, 这样无 free variable, $\boxed{3}$. $\mathcal{F} \# 1$

A 5×7 , if columns span \mathbb{R}^5 , how many pivot positions

$\boxed{5}$, 每行都有一个 pivot position $\boxed{1}$. \mathcal{T}

Ex 5 let $\vec{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $\vec{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$, $\vec{y}_1 = \begin{bmatrix} 2 \\ 5 \end{bmatrix}$, $\vec{y}_2 = \begin{bmatrix} -1 \\ 6 \end{bmatrix}$, $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ linear, which $\vec{e}_1, \vec{e}_2, \vec{y}_1, \vec{y}_2$ Find images of $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ (pp: -10#(3) basis map to \mathbb{R}^2 , J#(3) $\forall x$ column form to \mathbb{R}^2)

$$\boxed{T\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = T(x_1\vec{e}_1 + x_2\vec{e}_2) = x_1\begin{bmatrix} 2 \\ 5 \end{bmatrix} + x_2\begin{bmatrix} -1 \\ 6 \end{bmatrix}}$$

let $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, $\vec{v}_1 = \begin{bmatrix} 2 \\ 5 \end{bmatrix}$, $\vec{v}_2 = \begin{bmatrix} -1 \\ 6 \end{bmatrix}$, $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ linear, $\vec{x} \mapsto x_1\vec{v}_1 + x_2\vec{v}_2$

find A s.t $T(\vec{x})$ is $A\vec{x}$ for any \vec{x} (pp: 10#(T)_{PP})

$$\boxed{T(\vec{x}) = x_1\vec{v}_1 + x_2\vec{v}_2 = \begin{bmatrix} \vec{v}_1 & \vec{v}_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad A = \begin{bmatrix} 2 & 7 \\ 5 & -3 \end{bmatrix}}$$

见图 1

matrix tech

Ex 6 each matrix T is linear. True, $T(u+v) = A(u+v) = Au + Av = T(u) + T(v)$

Codomain of T $\vec{x} \mapsto A\vec{x}$ is set of all L.Comb of columns of A False

应为 range, codomain \mathbb{R}^m , if $A \in \mathbb{R}^{m \times n}$

$\{u, v\}$ L.Ind $\Leftrightarrow \{u+v, u-v\}$ L.Ind. $\checkmark \frac{a+b}{2}(u+v) + \frac{a-b}{2}(u-v) = 0$

1. Solutions of nonhomogeneous systems $\vec{A}\vec{x} = \vec{b}$.

$$\text{eg. } \vec{X} = \begin{bmatrix} \vec{x}_1 \\ \vec{x}_2 \\ \vec{x}_3 \end{bmatrix} = \begin{bmatrix} \vec{p} \\ \vec{0} \\ \vec{0} \end{bmatrix} + \vec{x}_3 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \vec{p} + \vec{x}_3 \vec{v} = \vec{p} + t \vec{v} \quad (t \in \mathbb{R})$$

(like through \vec{p} parallel to \vec{v})

2. Solutions of homogeneous systems $\vec{A}\vec{x} = \vec{0}$

$$\text{eg. } \vec{X} = \begin{bmatrix} \vec{x}_1 \\ \vec{x}_2 \\ \vec{x}_3 \end{bmatrix} = \vec{x}_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \vec{x}_2 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + \vec{x}_3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = t_1 \vec{v}_1 + t_2 \vec{v}_2 + t_3 \vec{v}_3 \quad (t_i \in \mathbb{R})$$

3. If $\vec{A}\vec{x} = \vec{b}$ has a solution \vec{p} , -> solution set $\{\vec{w} = \vec{p} + t\vec{v} \mid t \in \mathbb{R}\}$
 or: if $\vec{A}\vec{x} = \vec{b}$ has a solution, solution is unique precisely when $\vec{A}\vec{x} = \vec{0}$ only has trivial solution.

Vector space V , $S = \{\vec{v}_1, \dots, \vec{v}_p\} \subseteq V$; x_1, \dots, x_p scalar (in \mathbb{F})

4. Linear Independent $x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_p \vec{v}_p = \vec{0}$ only trivial solution

Linear dependent x_1, \dots, x_p not all zero.

Free var = \exists non-triv. soln.

#1 (columns of A are L.Ind iff $\vec{A}\vec{x} = \vec{0}$ has only trivial solution)

#2 $S = \{\vec{v}_1, \dots, \vec{v}_p\}$ L.dep iff at least one of vectors is L.comb. of others

(可推出: 若 L.Ind $S \subseteq V$, let $v \in V-S$, $S \cup \{v\}$ L.dep $\Leftrightarrow v \in \text{span}(S)$)

$\square \Rightarrow a_1 s_1 + \dots + a_n s_n + a v = \vec{0}$, $a \neq 0$ (因 S L.Ind. $\therefore v$ 不是 $\{s_1, \dots, s_n\}$ L.comb., $v \in \text{span}(S)$)

$\Leftarrow v = a_1 s_1 + \dots + a_n s_n$, note $a_1 s_1 + \dots + a_n s_n + (-1)v = \vec{0}$, $-1 \neq 0$, $\therefore S \cup \{v\}$ L.dep.

一些注意點見圖例 6-8

#3 any set $\{\vec{v}_1, \dots, \vec{v}_p\}$ in \mathbb{R}^n is L.dep if $p > n$ [eg. 4x5 matrix has L.dep]

$\text{p var} \geq n \text{ var} \Rightarrow \text{not has free var}$

#5 If $\{\vec{v}_1, \dots, \vec{v}_p\}$ in \mathbb{R}^n contains zero vector, then L.dep $\square \vec{v}_k = \vec{0} \Rightarrow \sum a_i \vec{v}_i = \vec{0}$

#6 If $\{\vec{v}_1, \dots, \vec{v}_p\}$ is L.dep, then $\{\vec{v}_1, \dots, \vec{v}_{p-1}, \vec{v}_p\}$ is L.dep; (因 $\{\vec{v}_1, \dots, \vec{v}_p\}$ L.dep, \vec{v}_p 不是 L.Ind, \vec{v}_p 是 $\{\vec{v}_1, \dots, \vec{v}_{p-1}\}$ L.dep, $\therefore \{\vec{v}_1, \dots, \vec{v}_{p-1}\}$ L.dep)

5. $\vec{A}\vec{x} = \vec{b}$ to $\vec{A}\vec{u} = \vec{0}$, \vec{A} max

Transformation/mapping T from \mathbb{R}^n to \mathbb{R}^m , $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$
 each \vec{x} in \mathbb{R}^n to $T(\vec{x})$ in \mathbb{R}^m ; \mathbb{R}^n : domain of T , \mathbb{R}^m : codomain of T

$T(\vec{x})$ in \mathbb{R}^m is image of \vec{x} , set of all images in range of T

real space homomorphism, 例 V, W 有同代映射 $\vec{x} \mapsto aT(\vec{x}) + b$ 仍成立

T 可以記為 $\vec{v} \mapsto \vec{v}$ 和 \vec{v} 是連續的不是連續的滿足, 證化 $T: V \rightarrow W$

6. T is linear if i) $T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v})$ for all \vec{u}, \vec{v} in domain

($T(\vec{0}) = \vec{0}$ 有用結果) ii) $T(c\vec{u}) = cT(\vec{u})$ for all scalars c and for all \vec{u} in domain

$T(\vec{u} + d\vec{v}) = cT(\vec{u}) + dT(\vec{v})$

Date.

(12) 1. 对应线性 $\begin{cases} 2u+v+w=5 \\ 4u-6v=-2 \\ 2u+7v+2w=9 \end{cases}$ 行行精是 plane 的交点

但仅考虑， $u\begin{bmatrix} 2 \\ -6 \\ 7 \end{bmatrix} + v\begin{bmatrix} 1 \\ -7 \\ 2 \end{bmatrix} + w\begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 5 \\ -2 \\ 9 \end{bmatrix}$ - comb of lins

证 1. $A_{mk} B_{kn} = \sum_{i=1}^n a_{ki} b_{in} \quad \square \quad LHS = \sum_{i=1}^n a_{ki} b_{in} \quad RHS_{ab} = \sum_{i=1}^n a_{ki} b_{in} \text{ 且 } V \otimes U = U V$

(13) 2. $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$, rotate each point in \mathbb{R}^2 about origin through an angle θ .

$\square \begin{bmatrix} 1 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix} \quad \begin{bmatrix} 0 \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix} \quad A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$

(1, 0)

(13) 3. Linear $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$. i) if \mathbb{R}^n onto \mathbb{R}^m ii) if one to one, ~~then~~ relation of m & n

\square i) Repeat each row at pivot position, $n \geq M$; ii) col L.hd. max (转上, 逆化简见 $\star\star\star P_i$)

(13) 4. $\begin{bmatrix} 1 & h & -3 \\ 0 & 2 & 6 \end{bmatrix} \quad h \neq 0$ (consistent)

$\square \begin{bmatrix} 1 & h & -3 \\ 0 & 4+2h & 0 \end{bmatrix} \quad h \neq -2, \text{ else } 0$

$\begin{bmatrix} 1 & h & -3 \\ 4 & 8+k & 0 \end{bmatrix} \quad h \neq -8$ i) no sol ii) unique iii) infinite many

$\square \begin{bmatrix} 1 & h & 2 \\ 0 & 8+4h & k+h \end{bmatrix} \quad i) 8+4h=0 \quad ii) 8+4h \neq 0 \quad iii) 8+4h=0$

(13) 5. $T_1(a_1, a_2, a_3) = (a_1+a_2+a_3, a_2-a_3, 2a_1-3)$

$\square T_1(0) = (1, 0, -3) \neq 0$ 非 L.T.

(13) 6. $A_{2 \times 3}$, one free variable $Ax=0$ is a line in \mathbb{R}^3 由 $\mathbf{v}_1, \mathbf{v}_2$ 为向量, 由 $\mathbf{v}_1 + t\mathbf{v}_2$ 构成
the free variable $Ax=0$ is a plane in \mathbb{R}^3 $(\mathbf{v}_3 = \mathbf{v}_1 + t(\mathbf{v}_2 - \mathbf{v}_1))$

(13) 7. if w is L.comb of u & v in \mathbb{R}^n , then w is L.comb of u and v . $\frac{w}{u} = 0 \bar{u} + 1 \bar{v}$

(13) 8. if none of vectors in $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ in \mathbb{R}^3 is a multiple of other vector; $\{S\mathbf{v}_1, S\mathbf{v}_2, S\mathbf{v}_3\}$ L.Ind. $\{(\mathbf{v}_1)(\mathbf{v}_2)(\mathbf{v}_3)\}, (\mathbf{v}_1)(\mathbf{v}_2)(\mathbf{v}_3), (\mathbf{v}_1)(\mathbf{v}_3)(\mathbf{v}_2), (\mathbf{v}_2)(\mathbf{v}_1)(\mathbf{v}_3), (\mathbf{v}_2)(\mathbf{v}_3)(\mathbf{v}_1), (\mathbf{v}_3)(\mathbf{v}_1)(\mathbf{v}_2), (\mathbf{v}_3)(\mathbf{v}_2)(\mathbf{v}_1)$, 则 $\{S\mathbf{v}_1, S\mathbf{v}_2, S\mathbf{v}_3\}$ 不可用 2×2 multiple.

(13) 9. $\mathbf{v}_1, \mathbf{v}_2$ in \mathbb{R}^4 , \mathbf{v}_2 not a scalar multiple of \mathbf{v}_1 , then $\{\mathbf{v}_1, \mathbf{v}_2\}$ L.Ind. $\{(\mathbf{v}_1)(\mathbf{v}_2)\}$ 为 zero vec. L.Ind.

两个都不是各自 scalar multiple

$T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear determined by 1xmn identity matrix (linear map)

A linear T is matrix T

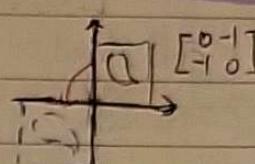
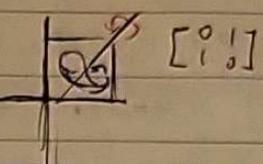
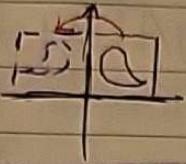
1. $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear iff \exists unique matrix A s.t.

$$T(\vec{x}) = A\vec{x}, \forall \vec{x} \in \mathbb{R}^n \quad (\text{e.g. col vector, row } j=1)$$

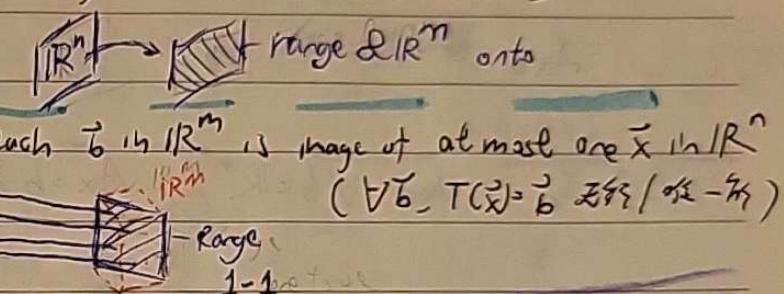
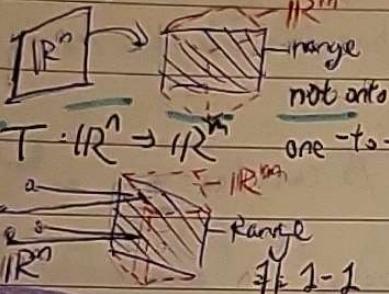
In fact, $A = [T(\vec{e}_1) \dots T(\vec{e}_n)]$ (A mxn)

$$\begin{aligned} \square \Rightarrow \vec{x} = I_n \vec{x} = [\vec{e}_1 \dots \vec{e}_n] \vec{x}, T(\vec{x}) = T([\vec{e}_1 \dots \vec{e}_n] \vec{x}) = \vec{x}_1 T(\vec{e}_1) + \dots + \vec{x}_n T(\vec{e}_n) = [T(\vec{e}_1) \dots T(\vec{e}_n)] \begin{bmatrix} \vec{x}_1 \\ \vdots \\ \vec{x}_n \end{bmatrix} \end{aligned}$$

2. 半直映射图:
standard matrix



3. $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ onto \mathbb{R}^m if each \vec{b} in \mathbb{R}^m is image of at least one \vec{x} in \mathbb{R}^n
(range of T is codomain \mathbb{R}^m) ($\forall \vec{b}$, $T(\vec{x}) = \vec{b}$ exist at least one sol.)



4. $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, T one-to-one iff $T(\vec{x}) = \vec{0}$ trivial sol only.

$\square \Rightarrow$ linear, $T(\vec{0}) = \vec{0}$ ($\vec{0}$ 的有用结论) \Leftrightarrow let T not 1-1, let $T(\vec{u}) = T(\vec{v}) = \vec{b}$,

$$T(\vec{u}) \neq T(\vec{u} - \vec{v}) = T(\vec{u}) - T(\vec{v}) = \vec{0}, \text{矛盾.}$$

逆: $T: V \rightarrow W$ linear, T 1-1 $\Leftrightarrow \text{Null}(T) = \{\vec{0}\}$

5. $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, A standard matrix for T . λ rank/nullity 定理 4

i) T onto \mathbb{R}^m iff columns of A span \mathbb{R}^m \square 由定理 4

ii) T one-to-one iff columns of A l. ind. \square 由定理 4. dep. free variable, 无无关

\square $AB = [Ab_1 \dots Ab_m] = \begin{bmatrix} a_1^T B \\ \vdots \\ a_n^T B \end{bmatrix}$, 也 = $\sum_{i=1}^n a_i b_i^T$ (由定理 4) $\rightarrow Ax = \sum a_i x_i$

6. $AB + BA$, in general: $AB = BA$ \Leftrightarrow $B = C$; $AB = 0 \Leftrightarrow A = 0$ or $B = 0$

$$I_m A = A = A I_n$$

从 transformation 角度看,

都是先 $C \# B \# A$, 然后 $\#$ 结论

AB, AC, BC 都是 standard matrix

从 transformation 角度看,

都是先 $C \# B \# A$, 然后 $\#$ 结论

AB, AC, BC 都是 standard matrix

从 transformation 角度看,

都是先 $C \# B \# A$, 然后 $\#$ 结论

AB, AC, BC 都是 standard matrix

从 transformation 角度看,

都是先 $C \# B \# A$, 然后 $\#$ 结论

AB, AC, BC 都是 standard matrix

得力

5

13) 1. $(AB)^T = (BA)^T$

$$\square (AB)_{ij}^T = (AB)_{ji} = \sum_k A_{jk} B_{ki} \text{ 而 } (BA)^T_{ij} = B \sum_k B_{ik} A_{kj}^T = \sum_k A_{jk} B_{ki}$$

13) 2. $AB \text{ nxn}, AB \text{ invertible}, \text{ 证 } A, B \text{ 可逆}$

$$\square \text{ let } C: (AB)^{-1}, \text{ consider } BC, A(BC) = (AB)C = I \quad \text{由 } CA \neq B^{-1} \Rightarrow AB \text{ 不可逆} \quad \text{且 } AB \text{ 可逆} \Leftrightarrow C \neq I$$

$$\text{而 } C = (I - BA)^{-1}, CAB = B^{-1}XB, I = B^{-1}XB \Rightarrow X = BIB^{-1} = I \Rightarrow C = I$$

13) 3. $AB \text{ last col 全为 } 0, B \text{ 无零 column. 证 } A \text{ L.Ind.}$ \times

$$\square AB = [B_{1,1} \dots B_{1,n}], B_{1,1} \neq 0, B_{1,i} \neq 0 \Rightarrow \text{有 nontrivial sol. } A \text{ col L. dep}$$

$$\text{or: } (AB)_{:,1} = 0 = A(B)_{:,1} \Rightarrow \sum_{i=1}^{n-1} t_i A_{i,1} \text{ where } B_{:,1} = \begin{pmatrix} t_1 \\ t_n \end{pmatrix} \text{ 而 } t_i \neq 0, \therefore \{A_{:,1}\} \text{ L. dep}$$

B 有 cols, L. dep by AB 有 col 也 L. dep

$$\square (AB)_{:,i} = A(B)_{:,i} \text{ 而 } B_{:,i} \text{ 已 L. dep. 由 } (AB)_{:,i} \text{ 也 } = 0$$

13) 4. $A = I_n, Ax = 0 \Rightarrow \text{only trivial sol. } A \text{ can't have more col than rows.}$ \square

$$\square Ax = 0, (Ax = 0 = x) \text{ 为 trivial sol.}$$

(i) $AD = I_m$, then $\forall b \in \mathbb{R}^m, Ax = b \text{ has a sol, } A \text{ can't have more row than col}$

$$\square AD_b = b \quad \text{且各 row 有唯一解, 各行各列有 pivot pos.}$$

(ii) A as $m \times n$, $\exists n \times m C$ and D s.t. $CA = I_n$ and $AD = I_m$, prove $m = n, C = D$

$$\square \text{ 可逆即同时满足 i) & ii) } - (C_{nm}(A_{mn})D_{nm}) = D = C \quad \therefore \text{由 i) & ii) } m = n$$

证明 A 可逆, $Ax = 0$ 只有 trivial sol. A span \mathbb{R}^n . 满足 iii.

13) 5. $A \leftarrow A + \varepsilon B, \varepsilon \text{ small 证 } A^{-1} \leftarrow A^{-1} - \varepsilon A^{-1}BA^{-1}$ $\square A^{-1} = (A + \varepsilon B)^{-1} = [A(I + \varepsilon A^{-1}B)]^{-1} = (I + \varepsilon A^{-1}B)^{-1}A^{-1}$
 $\text{Optim P. 3} = (I - \varepsilon A^{-1}B + \varepsilon^2(A^{-1}B)^2 - \dots)A^{-1} \approx A^{-1} - \varepsilon A^{-1}BA^{-1}, \text{ if you neglect } \varepsilon^2(A^{-1}B)^2 - \dots \text{ term}$

13) 6. A invertible $n \times n, B_{n \times p}$, 证 $[AB] \sim [IX], X = A^{-1}B$

$$\square (E_p - E_D)A = I, \therefore (E_p - E_D)B = X = A^{-1}B$$

13) 7. $A = A^T, \text{ diag } = 0, \text{ 且 } \forall x, x^T A x = 0 \quad \square$

$$\begin{aligned} & 1^\circ A + A^T = 0 \\ & 2^\circ (x^T A x)^T = x^T A x, = -x^T A x, \therefore 2x^T A x = 0 \end{aligned}$$

13) 8. $ABX \text{ nxn}, AX, A-AX \text{ 可逆}, (A-AX)^{-1} = X^T B, \text{ 证 } B \text{ 可逆, 并 solve } X$

NOTE 除了证 $B^T B = BB^T = I$ 外 (这里此法较难), 用 $B = X(A-AX)^{-1}$ 可逆 $\Leftrightarrow X$ 可逆

$$\square X = B(A-AX), X = (I-BA)^{-1}BA$$

13) 9. $\square AB-1 \text{ 可逆} \Rightarrow BA-1 \text{ 可逆} \quad \square \text{ why? } BA-1 \text{ 可逆} \Rightarrow AB-1 \text{ 可逆}$

13) 4. 2

$$\therefore BA-1 \text{ rank } = 0 \quad \therefore BA \text{ rank } = 1 \quad \square$$

13) 10. 证 $I-AB \text{ 可逆} \Rightarrow I-BA \text{ 可逆}$ $\square 1 - (I-AB) = AB - I \text{ 可逆} \quad \square 2 \text{ 若 } (AB) \rightarrow 0 \text{ p.v.}$

$$(I-AB)^{-1} \exists, (I-AB)^{-1} = I + ABA + ABA + \dots = I + A(I+BA+ABA+\dots)B = I + A(I-BA)^{-1}B,$$

Optim P. 3

symmetric: $A = A^T$ (if A is symmetric, $A + A^T$ symmetric)

skew-symmetric: $A^T = -A$ ($A - A^T$ skew-symmetric)

Invertible &
Transpose

$$[A \ B]^T = [A^T \ B^T]$$

$$W^T = \begin{bmatrix} X^T \\ Y^T \end{bmatrix}$$

$$\begin{aligned} (rA+sB)^T &= rA^T+sB^T \\ (rA^T+sB^T)^T &= rA+sB \end{aligned}$$

Date

1. Transpose $A_{m \times n} \rightarrow A^T_{n \times m}$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}^T = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}^T = \begin{bmatrix} 1 & 3 & 1 \\ 2 & 4 & 2 \end{bmatrix}^T = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

由逆可得 $A^T = (A^T)^T = A$

2. $A_{n \times n}$ invertible if $\exists! C_{n \times n}$ s.t. $CA = I_n = AC$, $C \neq A^{-1}$
to C 不唯一, $B = BI = B(AC) = (BA)C = IC = C \Rightarrow B = C$

3. 对于 2×2 的 $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ if $|A| = ad - bc \neq 0$. $A^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$
如果 $|A| = 0$, not invertible ($A \sim \begin{bmatrix} a & b \\ 0 & d - \frac{bc}{a} \end{bmatrix}$ 不可逆)

4. $A_{n \times n}$ invertible $\Leftrightarrow A^T \in \mathbb{R}^n$, $A\vec{x} = \vec{b}$ has unique sol $\vec{x} = A^{-1}\vec{b}$ 可逆,
且唯一 way. 由 \vec{x} 有 $A\vec{u} = \vec{b}$, $A^{-1}A\vec{u} = \vec{u} = A^{-1}\vec{b} = \vec{x}$

$$[A \ b] \sim [I \ x]$$

col of A L.Ind.
span \mathbb{R}^n

SP. $(A^{-1})^{-1} = A$, i.e. A is invertible, then A^{-1} invertible $\square A A^{-1} = I = A^{-1}A$

2. $(AB)^{-1} = B^{-1}A^{-1}$ $\square AB \cdot (B^{-1}A^{-1}) = AIB^{-1} = I$, 由 $(B^{-1}A^{-1})AB = I$ 可逆

3. $(AT)^{-1} = (A^{-1})^T$ $\square (A^{-1})^T A^T = ((AA^{-1})^T)^T = I^T = I$ 4. $r \neq 0$ ($rA^T = r^{-1}A^{-1}$ 可逆)

6. 对 $A'_{m \times n}, B'_{n \times p}$, 考虑 Augment matrix $m \times (n+p)$ $(A+B)$; note. $\forall M_{m \times n}, M(A+B) = (MA | MB)$

\square 对 $M(A+B)$ 有 n 个 $n \times n$. $M(A+B)_{ij}, j=1 \dots n$, 为 $\sum_k M_{ik} A_{kj} = MA_{ij}$. 同时 $j=n+1 \dots n+p$. $M(A+B)_{ij} = MB_{ij}$

以此用行消元. $A_{n \times n} \Leftrightarrow (A | I_n)$ element by row operation $\Leftrightarrow (I_n | A^{-1})$ (类似左互易)

$\square \Rightarrow$ 对 $A^{-1}(A | I_n) = ((I_n | A^{-1}))$, 因 $A^{-1} = E_p \cdots E_1$, $E_p \cdots E_1(A | I_n) = (I_n | A^{-1})$ 由 element
 \Leftrightarrow 令 $M = \text{row op. 矩阵和}$, 中 $M = E_p \cdots E_1$, $E_p \cdots E_1(A | I_n) = (I_n | B') = (MA | MI_n)$ $\therefore MA = I_n$
WTS $B = A^{-1}$. 由 $MA = I_n$, $M = A^{-1} = B$

7. Linear transform to special case of affine transformation: $f(x) = Ax + b$

Def. T affine iff \forall scalar λ $f(\lambda x + (1-\lambda)y) = \lambda f(x) + (1-\lambda)f(y)$

$$\text{应用} \quad X = A_{11}^{-1} A_{21}, \quad Y = A_{11}^{-1} A_{12}$$

$$\square \text{ If } A_{11} \text{ 可逆, } [A_{11} \ A_{21}] = [I \ 0] [A_{11} \ 0] [I \ Y] \Leftrightarrow S = A_{21} - A_{11}^{-1} A_{12}$$

\square 由 A_{21} 可逆, 令 $S = A_{11} - A_{11}^{-1} A_{12}$, S 称为 Schur complement
to $[A_{11} \ A_1] \in \mathbb{R}^{n \times n}$ 都可逆, S 也可逆.

\square Lemma $\begin{bmatrix} B & 0 \\ 0 & C \end{bmatrix}$ 可逆 iff B, C 都可逆 (B, C square)

$$\square (\Rightarrow) \begin{bmatrix} * & * \\ * & W \end{bmatrix} \begin{bmatrix} B & 0 \\ 0 & C \end{bmatrix} = \begin{bmatrix} * & * \\ * & W \end{bmatrix} \xrightarrow{W=C} (\Leftarrow) \begin{bmatrix} B & 0 \\ 0 & C \end{bmatrix} \begin{bmatrix} B^{-1} & 0 \\ 0 & C^{-1} \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}$$

$$\begin{bmatrix} I & 0 \\ -X & I \end{bmatrix} \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} I & -Y \\ 0 & I \end{bmatrix} = I \begin{bmatrix} A_{11} & 0 \\ 0 & S \end{bmatrix} I, \text{ 故为三对角矩阵, 且 } A \text{ 可逆} \\ (\text{即 } S \text{ 为 } A_{11} \text{ 的 Schur complement}) \text{ 因此 } \begin{bmatrix} A_{11} & 0 \\ 0 & S \end{bmatrix} \text{ 为 } S \text{ 可逆.}$$

应用: a standard set of differential equations is transformed by Laplace transforms

into $\begin{bmatrix} A - sI_n & B \\ C & I_m \end{bmatrix} \begin{bmatrix} \vec{x} \\ \vec{u} \end{bmatrix} = \begin{bmatrix} \vec{y} \\ \vec{0} \end{bmatrix}$, where $A_{m \times n}, B_{n \times m}, C_{m \times n}, s$ a variable, \vec{x}, \vec{y} function

of s , \vec{u} in \mathbb{R}_m (\vec{u}_{input}), \vec{x} in \mathbb{R}^n (state vector), \vec{y} in \mathbb{R}^m (output)

有 transfer function $W(s)$ s.t. $W(s)\vec{u} = \vec{y}$ (input \Rightarrow output). $\Rightarrow A - sI_n$ 可逆, $W(s)$ 为全阶可逆

$$\square \begin{cases} (A - sI_n)\vec{x} + B\vec{u} = \vec{0} \\ C\vec{x} + I_m\vec{u} = \vec{y} \end{cases} \Rightarrow \begin{cases} \vec{x} = (A - sI_n)^{-1}(-B\vec{u}) \\ \vec{y} = (I_m - C(A - sI_n)^{-1}B)\vec{u} \end{cases} \Rightarrow \text{由 } (A - sI_n) \text{ 的 Schur complement}$$

由上得 斜尔逆有可逆情况, 则可找出 $W(s)^{-1}$, output \Rightarrow input

$$\square \vec{x} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \xrightarrow{\text{逆用 U, rotate}} V_1 \quad \text{设 } \cos(\theta_1 + \theta_2) = \cos\theta_1 \cos\theta_2 - \sin\theta_1 \sin\theta_2$$

$$\begin{cases} \text{设 } V_2 \\ \text{由 } V_1 = \begin{pmatrix} \cos\theta_1 & -\sin\theta_1 \\ \sin\theta_1 & \cos\theta_1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} \cos\theta_1 \\ \sin\theta_1 \end{pmatrix} \quad V_2 = \begin{pmatrix} \cos\theta_2 & -\sin\theta_2 \\ \sin\theta_2 & \cos\theta_2 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} \cos\theta_2 \\ \sin\theta_2 \end{pmatrix} \\ \cos(\theta_1 + \theta_2) = \cos(\text{angle between } V_1, V_2) = \frac{\langle V_1, V_2 \rangle}{\|V_1\| \|V_2\|} \end{cases}$$

\square 1: $A_{m \times n}$ to form a basis of \mathbb{R}^m , by $m \geq n$

\square 2: $\text{span } \mathbb{R}^m, \text{ if } \text{Col } A \neq \mathbb{R}^m$ (由 L.I. \Rightarrow $n \leq m \Rightarrow n = m$)

if $A_{m \times m}$, $\text{Col } A \neq \mathbb{R}^m$, by $\text{Nul } A \neq \{0\}$, A 不可逆

\square 由四.3, $\text{rk } A \neq n$: nullity $\neq 0$, 由非满秩不可逆 OR PP 时从 non pivot pos., $\therefore Ax=0$ 有 nontrivial sol

$A_{3 \times 5}$ has 3 pivot col, Is $\text{Col } A = \mathbb{R}^3$? Is $\text{Nul } A = \mathbb{R}^2$?

$\square \text{ Col } A = \mathbb{R}^3$, 但 nullspace $\leq \mathbb{R}^5$ ($\text{dim } N(A) \neq \text{Nul } A$)

$A_{4 \times 7}$ has 3 pivot col, Is $\text{Col } A = \mathbb{R}^3$? What is $\text{dim } \text{Nul } A$?

\square $\text{Col } A \leq \mathbb{R}^4$ nullity = 7 - rk A = 7 - 3 = 4

Each line in \mathbb{R}^n is one-dimensional subspace of \mathbb{R}^n . \times

由 line 不经 origin offspace, dim only def for subspace

$\square A_{6 \times 4}, B_{4 \times 6}, AB \neq \text{full rank} \quad \checkmark \quad \square B_{6 \times 6} \text{ full rank, } B \text{ L.dep.} \therefore Bx=0 \text{ 有 nontrivial sol, 两边同乘 } A$

$ABx=AO=0, \therefore AB \text{ nullity} \neq 0, \text{ not full rk}$

Invertible 2
VS 5

Ans

Date: / /

$A_{n \times n}$. 下同 同理

即不需证明满足 $CA = AC = I$, \square square

a. A is invertible; A^T is invertible; there is $n \times n$ C s.t. $CA = I$; there is $n \times n$ D s.t. $AD = I$

b. A row equivalent to $n \times n$ Identity matrix

c. A has n pivot position; cols of A span \mathbb{R}^n ; $\forall b \in \mathbb{R}^n$, $A\vec{x} = \vec{b}$ has one unique sol.

d. $A\vec{x} = \vec{0}$ has only trivial sol; cols of A L. Ind.

\Rightarrow L. Ind. \Leftrightarrow unique sol.

即 $A\vec{x} = \vec{0}$ 只有唯一解

f. linear $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is one-to-one; maps \mathbb{R}^n onto \mathbb{R}^n ; T bijective; T invertible

見出課題四

2. let $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ be L.T. and let A be standard matrix for T . \star 例題 4 & A, R \square

T is invertible iff A is invertible. A^{-1} would be standard matrix for T^{-1} .

3. If $A_{m \times n}$, $B_{n \times o}$, A partition into x cols and y rows, B partition into y rows and z rows

then they are comfortable for block multiplication

eg. $AB = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} = \begin{bmatrix} A_{11}B_1 + A_{12}B_2 \\ A_{21}B_1 + A_{22}B_2 \end{bmatrix}$ (i,j) in $\cup_k(A) \cap \cup_k(B) \Rightarrow a_{ik} b_{kj}$
 $\therefore \sum (i,j) \in (i,j) \in AB$

In general $AB = [c_{11}, c_{12}, \dots, c_{1n}] \begin{bmatrix} B_1 \\ B_2 \\ \vdots \\ B_m \end{bmatrix} = [c_{11} \text{ row}_1 + \dots + c_{1n} \text{ row}_n]$

4. block upper triangular $A = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix}$, assume A_{11} p.p., A_{22} $q \times q$, A invertible

$\exists A^{-1} = \begin{bmatrix} A_{11}^{-1} & A_{11}^{-1}A_{12}A_{22}^{-1} \\ 0 & A_{22}^{-1} \end{bmatrix}$

$\square \nexists AB = I_{p+q}$ $\begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix} \begin{bmatrix} B_1 & B_2 \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} I_p & 0 \\ 0 & I_q \end{bmatrix}$

$\begin{cases} A_{11}B_1 + A_{12}B_{21} = I_p \\ A_{11}B_{21} + A_{12}B_{22} = 0 \\ A_{21}B_{21} = 0 \\ A_{21}B_{22} = I_q \end{cases} \Rightarrow \begin{cases} A_{11}B_1 = I_p \therefore B_1 = A_{11}^{-1} \\ A_{12}B_{21} = 0 \therefore B_{21} = A_{12}^{-1} \\ A_{11}B_{22} = 0 \therefore B_{22} = A_{11}^{-1} \\ A_{12}B_{22} = I_q \therefore B_{12} = -A_{11}^{-1}A_{12}A_{22}^{-1} \end{cases}$

vector space

用三點可推出 subspace & vector space, $H \subseteq \mathbb{R}^n$ 一部分, 但保留代數結構

5. Subspace of \mathbb{R}^n is any set H , (a) zero vector in H b) each $\vec{u}, \vec{v} \in H$, $\vec{u} + \vec{v} \in H$

(b) $\forall k \in \mathbb{K}$, $k\vec{u} \in H$ \Rightarrow 必須要非空, 見 P. 9. 1 NOTE

c) each $\vec{u} \in H$, λ scalar, $\lambda\vec{u} \in H$. \neg 只不過原點不是 subspace, $\lambda \neq 0$ 且 $\lambda \neq 1$ 時

$\vec{v}_1, \dots, \vec{v}_p \in \mathbb{R}^n$, All L. Comb of $\vec{v}_1, \dots, \vec{v}_p$ is subspace in \mathbb{R}^n ; column space of $A_{m \times n}$ is subspace in \mathbb{R}^m

(The null space of $A_{m \times n}$ is set of solutions of homogeneous equation $A\vec{x} = \vec{0}$, a subspace in \mathbb{R}^n)

這 col space of \mathbb{R}^m , 为 to be claim. . . col space 也用得方 output to the 集合 \square (由 \mathbb{R}^m)

一直有 col space -> 后因由其是 subspace. 因 A 的 col 即直接合基 basis 所在的向量, 造成它们即 col space

-> 由其可被线性变换得到. 无压缩的情况下, null space 由可直接到的向量 $\lambda \vec{u}$ 成.

$H_1 = \{(a-b, b-a, a, b) : a, b \in \mathbb{R}\}$, H_1 is subspace of \mathbb{R}^4
 $H_2 = \{(a, b, c, d) : a-2b+5c-d=0 \text{ and } c-a=b\}$, H_2 is subspace of \mathbb{R}^4

Date: / / set of homogeneous equations ($\mathbb{R}^{4 \times 4}$)
 $\begin{cases} a-2b+5c-d=0 \\ -a+b+c=0 \end{cases}$ He subspace

(2) | $A = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 2 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}$ find basis for null space of A .

$$\boxed{\text{if } Ax=0, [A|0] = \begin{bmatrix} 1 & -2 & 1 & 3 & 0 \\ 0 & 1 & 2 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}, \vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = x_2 \begin{bmatrix} 2 \\ 0 \\ -2 \\ 1 \end{bmatrix} + x_4 \begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \end{bmatrix} + x_5 \begin{bmatrix} 3 \\ 0 \\ 2 \\ 1 \end{bmatrix}}$$

spanning = Null(A)
L.I. \rightarrow basis for null
Note, $x_4, x_5 \neq 0$

(3) 1. $C \cap N(A^T) = \{0\}$ \rightarrow the $A \in C$, 若 $A \in N(A^T)$, $A^T A = 0 \Rightarrow \text{rank}(A^T A) = \text{rank}(A)$
 $\therefore A = 0$ pp intersection to 0

(3) 2. find basis for col space of $A = \begin{bmatrix} 1 & 0 & -3 & 5 & 0 \\ 0 & 1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$ (pivot column \rightarrow basis)

$\boxed{V \in \mathbb{R}^4 \quad \vec{v} = \sum c_i \vec{a}_i, \text{ 但 } \vec{a}_1, \vec{a}_2 \text{ 线性无关.} \therefore \text{L.I. basis } \{a_1, a_2, a_5\}}$

(3) 3. $\vec{v}_1 = \begin{bmatrix} 3 \\ 6 \\ 2 \\ 2 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \\ 1 \end{bmatrix}, \vec{x} = \begin{bmatrix} 3 \\ 1 \\ 2 \\ 7 \end{bmatrix}, \beta = \{\vec{v}_1, \vec{v}_2\}$, then β is basis for $H = \text{Span}(\vec{v}_1, \vec{v}_2)$ since \vec{v}_1, \vec{v}_2 L.I.

If \vec{x} in H , find coordinate vector $\boxed{\text{因 } \vec{x} \in H, \vec{x} = \sum x_i \vec{v}_i = \begin{bmatrix} 3 \\ 1 \\ 2 \\ 7 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \\ 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ 6 \\ 0 \\ 0 \end{bmatrix} \Rightarrow \begin{cases} x_1 = 2 \\ x_2 = 3 \end{cases}}$

$$\text{坐标向量 } [\vec{x}]_{\text{坐标}} = \underbrace{Q}_{\text{基底}} [\vec{x}]_{\beta}$$

$$\begin{bmatrix} 3 & -1 \\ 6 & 0 \\ 2 & 1 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 3 \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \\ 2 \\ 7 \end{bmatrix}$$

(3) 4. $\begin{bmatrix} 1 & 3 & 3 & 3 & -9 \\ -2 & -2 & 3 & 3 & 2 \\ 3 & 3 & 1 & 7 & 2 \\ 3 & 4 & -1 & 1 & 8 \end{bmatrix}$ 可化为 $\begin{bmatrix} 1 & 0 & -3 & 5 & 0 \\ 0 & 1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$ 且知 x_3, x_4 为 free variable, x_1, x_2, x_5 为 pivot variable, 故 $\{x_1, x_2, x_5\}$ 为 basis, 但只有 $\begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}$ 是而不是 $\begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 0 \end{bmatrix}$ 和 $\begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \end{bmatrix}$, 即 $\{x_1, x_2, x_5\}$ 不是 basis

pivot col 与自由变量无关, 各 pivot col 为 L.I., 有 free variable 即可被它们表示

(3) 5. $\alpha_1, \dots, \alpha_r$ L.I. $a \in \text{span}(\alpha_1, \dots, \alpha_r)$, $v \notin \text{span}(\alpha_1, \dots, \alpha_r)$ $\exists t \in \mathbb{R}$ s.t. $v = t\alpha_1 + \dots + t\alpha_r$

$\boxed{0 = \sum_{i=1}^r b_i \alpha_i + t(b_1 + \dots + b_r)}$ 若 $b_1 + \dots + b_r \neq 0$, $t(b_1 + \dots + b_r) = -\sum_{i=1}^r b_i \alpha_i$, 则 $v \in \text{span}(\alpha_1, \dots, \alpha_r)$
 $\therefore b_1 + \dots + b_r \neq 0$, 而 $\alpha_1, \dots, \alpha_r$ L.I. $\therefore b_1, \dots, b_r = 0 \therefore v \in \text{span}(\alpha_1, \dots, \alpha_r)$

$\boxed{k \geq 2, a_i \in \mathbb{R}^n, \{a_1, \dots, a_k\} \text{ L.dep.} \Leftrightarrow \exists \text{ scalar } b_1, \dots, b_k \text{ st. } \sum b_i a_i = 0, \sum b_i \neq 0}$

$\boxed{\det \vec{a}_i = \begin{vmatrix} 1 & & & \\ a_{i1} & \dots & a_{i(n+1)} \end{vmatrix} \in \mathbb{R}, \vec{a}_i \text{ RL dep.} \Rightarrow \text{有非全 } 0 \text{ 行 st. } \sum b_i \vec{a}_i = 0 \Rightarrow \begin{bmatrix} \sum b_i \\ \vdots \end{bmatrix} = 0}$

从 AB , A diagonal matrix, \vec{v} scalar times rows of B \Rightarrow AB is $\begin{bmatrix} 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \end{bmatrix}$, AB is $\begin{bmatrix} 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \end{bmatrix}$

If $A \vec{v} = \vec{v}$, then $A = \vec{v}$ or $\vec{v} = 0$ \times $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \vec{v} = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$

$(A+B)(A-B) = A^2 - AB + BA - B^2$, only when A, B commute $\Rightarrow A^2 = B^2$

$\frac{1}{4}$ square matrix is a product of elementary matrices. $\times \text{op BA} = E_p \cdot E_1, E_2, E_3 \text{ 等} \Rightarrow$ $\boxed{A \sim T}$

$A \sim T, r \neq n, (rA)^{-1} = r^{-1}A^{-1}$

$\boxed{n \times n \det(CA) = C^n \det(A)}$

$\boxed{\text{op } AB + BA, \det BA = \det AB}$ \checkmark $\boxed{\det(A+B) = \det A + \det B}$

$$\begin{bmatrix} 3 \\ -1 \end{bmatrix} \begin{bmatrix} 6 \\ 2 \end{bmatrix} \in \text{L.I. Ind.} \quad \text{For } \text{span}(A^T) \text{ basis.} \quad \text{IR}^2 \text{ 是 IR}^3 \text{ 的子空间, } H \text{ 是 } K \text{ 的子集.}$$

(dim 2 = 2 < 3 Rank 2)

1. Basis for vector space V (P_E) is L.I. set in V that spans H . $H = \{\begin{bmatrix} x \\ y \end{bmatrix}\}$ is IR^2 subspace.

e.g. $\{e_1, \dots, e_n\}$ standard basis for IR^n , thus $\{e_1, \dots, e_n\}$ is basis for IR^n .

e.g. $\forall V = \text{span}(A)$, $\{A\}$ is basis for V .

2. if set $\beta = \{b_1, \dots, b_p\}$ is basis for subspace H , for each $\vec{x} \in H$, coordinate of \vec{x} relative to basis β is $\vec{x} = c_1 \vec{b}_1 + c_2 \vec{b}_2 + \dots + c_p \vec{b}_p$, and $[\vec{x}]_{\beta} = \begin{bmatrix} c_1 \\ \vdots \\ c_p \end{bmatrix}$ is coordinate vector relative to β . $(\vec{x} = [\vec{b}_1, \dots, \vec{b}_p] [\vec{x}]_{\beta})$

The dimension of a nonzero subspace H is number of vectors in any basis for H . $(\dim[H] = \dim[\vec{b}_1, \dots, \vec{b}_p] \vec{x} = [\vec{x}]_{\beta})$

The rank of A is the dimension of col space of A . $\{ \dim \text{of subspace } \text{col } A \text{ is defined as } 0 \}$
即 $\text{rank}[A]$ 等于 A 为满秩, $\text{rank} = 1$, A 不满秩, $\text{rank} = 2, \dots$, 即 rank 是 A 中非零列数。

3. (Rank Theorem) If A has n col, $\text{rank } A + \dim(\text{Null}(A)) = n$:

direct proof: $T: V \rightarrow W$ linear, $\text{nullity}(T) + \text{rank}(T) = \dim(V)$

\square let $\dim(V) = n$, $\text{nullity}(T) = k$, $\{v_1, \dots, v_k\}$ basis for $\text{Null}(T)$, $\{v_{k+1}, \dots, v_n\}$ $\Rightarrow V$ basis $\{T(v_1), \dots, T(v_n)\}$ $\Rightarrow S = \{T(v_1), \dots, T(v_n)\}$ $\Rightarrow S$ is $\text{Range}(T)$ basis $\Rightarrow S$ span: $\text{Range}(T) = \text{span}(T(v_1), \dots, T(v_n)) = \text{span}(T(v_{k+1}), \dots, T(v_n)) = \text{span}(S)$
 $\Rightarrow S$ L.I. Ind. let $\sum_{i=1}^k a_i T(v_i) = J$, $\Rightarrow T(\sum_{i=1}^k a_i v_i) = 0$, $\sum_{i=1}^k a_i v_i \in \text{Null}(T)$, $\therefore \sum_{i=1}^k a_i v_i \in \{v_{k+1}, \dots, v_n\}$ $\Rightarrow \sum_{i=1}^k a_i v_i \in \sum_{i=k+1}^n b_i v_i$
由 $\{v_1, \dots, v_n\}$ \Rightarrow $\sum_{i=1}^k b_i v_i - \sum_{i=k+1}^n a_i v_i = 0$, $a_i = b_i = 0$ $\therefore \{T(v_{k+1}), \dots, T(v_n)\}$ basis, $\text{rank } k = n - k$

4. $A_{n \times n}$, A 可逆除了前述上页的定理外:

a). Col of A form basis of IR^n

b). Col space $A = \text{IR}^n$: $\dim(\text{Col } A) = n$; $\text{rank } A = n$

\square 以 T 表示: $\text{linear } T: V \rightarrow W$, T bijective $\Leftrightarrow \text{rank}(T) = \dim(T)$; T 1-1, $\therefore \text{nullity } = 0$.

$\Rightarrow \dim(V) = \dim(W)$, $\text{range}(T) = W$, $\text{rank } T = \dim V$ \Rightarrow (i, j) cofactor C_{ij}

5. Determinant $\det A_{n \times n} = a_{11} \det A_{11} - a_{12} \det A_{12} + \dots + (-1)^{i+j} a_{ij} A_{ij} = \sum_{j=1}^n (-1)^{i+j} a_{ij} \det A_{ij}$

\Rightarrow 三阶矩阵, \det 证明 $\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \sim \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} - a_{12} a_{11} & a_{23} - a_{13} a_{11} \\ 0 & 0 & a_{33} \end{bmatrix} \Rightarrow \det A = a_{11} a_{22} a_{33} + \dots$ \Rightarrow $\det A$ 如 A 为 3×3 , $\Delta \neq 0$, $\Rightarrow \Delta$ 有主意义, 有 S_3 意义

cofactor, 不只是 ΔC_{ij} , 也对 V row / col cofactor expansion.

6. Row operations: 1° 两行互换, \det 变号 2° - 行加到 i -行, \det 不变 3° - $i \times k$, $\det \propto k$

推论 1° 两行相同, $\det = 0$. 2° 两行成比例, $\det = 0$ \Rightarrow A 为满秩 \Leftrightarrow $(\exists \text{ pivot position}) \det \neq 0$

7. 行列式可看成代数变换 $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ 因此扩大的区域扩大了多少倍, 因此只要 $\det \neq 0$, 即化为不纯用当前维度表示 $(\text{不纯 } 0 \dots)$, A^{-1} 可用初等行变换求出, A^{-1} 有在

已知 $\det(A_1 A_2) = \det(A_1) \det(A_2)$ 伸展两次

\Rightarrow $\det(A_1 A_2) = \det(A_1) \det(A_2)$ \Rightarrow $\det(A_1 A_2) = \det(A_1) \det(A_2)$

应用 $A = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ $B = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ Pauli spin matrices, $AB = -BA$, anticommute

应用, 寻找 $(-2, 0)(0, 3)(1, 3)(-1, 0)$ 因成形而选取的矩阵 $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 2 & 3 \end{pmatrix}, \begin{pmatrix} 1 & 3 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix}$ 使得 $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 1 & 3 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 3 & 1 \end{pmatrix}$

例) to $A^n = 0$ for some $n > 1$, $(I-A)^{-1}$?

$$\boxed{\text{口 } (I-A)(\underbrace{I+A+A^2+\dots+A^{n-1}}_{(I-A)^{-1}}) = I + A^n = I}$$

应用, let a, b positive, find \vec{u}, \vec{v} s.t. $\frac{x_1^2}{a^2} + \frac{x_2^2}{b^2} = 1$

口 从单位圆映射到 $\frac{x_1^2}{a^2} + \frac{x_2^2}{b^2} = 1$ 公方程

$$A\vec{u} = \vec{x}, \vec{u} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}, \vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} au_1 \\ bu_2 \end{bmatrix}$$

$$\text{设 } A = \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix}$$

$$\text{area} = |\det A| \cdot A_{\text{area}} = ab\pi(1) = \pi ab$$

例) Vandermonde $V = \begin{bmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^{n-1} \end{bmatrix}, \vec{y} = (y_1, \dots, y_n) \in \mathbb{R}^n, \vec{c} = (c_0, \dots, c_{n-1}) \in \mathbb{R}^n$

$$V\vec{c} = \vec{y}, p(t) = c_0 + c_1 t + \dots + c_{n-1} t^{n-1}$$

a) show $p(x_1) = y_1, p(x_2) = y_2, \dots, p(x_n) = y_n$. $p(t)$ interpolating polynomial for $(x_i, y_i) \in \mathbb{R}^2$, graph pass through points

b) to x_1, \dots, x_n distinct. show col of V L.Ind.

$$\boxed{\text{口 a) } \begin{bmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^{n-1} \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ \vdots \\ c_{n-1} \end{bmatrix} = \begin{bmatrix} c_0 + c_1 x_1 + \dots + c_{n-1} x_1^{n-1} \\ c_0 + c_1 x_2 + \dots + c_{n-1} x_2^{n-1} \\ \vdots \\ c_0 + c_1 x_n + \dots + c_{n-1} x_n^{n-1} \end{bmatrix} = \begin{bmatrix} p(x_1) \\ p(x_2) \\ \vdots \\ p(x_n) \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}}$$

b) $n-1$ 元多项式是含 $n-1$ 解, to $\vec{y} = 0$, $p(x_1) = p(x_n) = 0$ 不能同时成立. $c_0 = c_1 = \dots = c_{n-1} = 0$ 使 $p(x) = 0$.

因 x_1, \dots, x_n 不同, 即 $n-1$ 未知数有 n 个解, 于是, 一且 $c_0 = c_1 = \dots = c_{n-1} = 0$ 才使 $p(x) = 0$. c R 为 trivial sol. 且 $c \perp \text{Ind}$

应用 证明 $\begin{bmatrix} A & 0 \\ C & D \end{bmatrix} = \det A \cdot \det D = \det \begin{bmatrix} A & B \\ 0 & D \end{bmatrix}$ (若 A 有特征值 0 三阶时)

$$\begin{bmatrix} A & 0 \\ C & D \end{bmatrix} = \begin{bmatrix} A & 0 \\ 0 & D \end{bmatrix} \begin{bmatrix} I & 0 \\ C & D \end{bmatrix} \therefore \det \begin{bmatrix} A & 0 \\ C & D \end{bmatrix} = \det A \cdot \det D$$

$$\det \begin{bmatrix} A & B \\ 0 & D \end{bmatrix} = \det \begin{bmatrix} A & B \\ 0 & D \end{bmatrix}^T = \det \begin{bmatrix} A^T & 0 \\ 0 & D^T \end{bmatrix} = \det A^T \det D^T = \det A \det D$$

应用. 用外积差推理论 $\vec{u} \times \vec{w} = \begin{vmatrix} \vec{i} & \vec{j} & \vec{k} \\ \vec{u}_1 & \vec{u}_2 & \vec{u}_3 \\ \vec{w}_1 & \vec{w}_2 & \vec{w}_3 \end{vmatrix}$

口 考虑 $f(\vec{z}) = \det \begin{bmatrix} \vec{y} & \vec{u}_1 & \vec{w}_1 \\ \vec{z} & \vec{u}_2 & \vec{w}_2 \\ \vec{v} & \vec{u}_3 & \vec{w}_3 \end{bmatrix}$ 可以看成 \vec{z} 的 \mathbb{R}^3 input \mathbb{R}^3 output \mathbb{R} , $f(\vec{x}) = A \vec{x}$

表 $A_{1 \times 3}$ (基底向量 $\vec{u}, \vec{v}, \vec{w}$ 为 0) $[A_{11}, A_{12}, A_{13}] \begin{bmatrix} \vec{y} \\ \vec{z} \\ \vec{v} \end{bmatrix}$ 在 \mathbb{R}^3 上的点乘, 几何意义为 (A_1, A_2, A_3)

在 (x, y, z) 上的点乘, (x, y, z) 长度, 为 $\vec{u}, \vec{v}, \vec{w}$ 得 $\begin{cases} A_1 \vec{x} = (u_2 w_3 - u_3 w_2) \vec{x} \\ A_2 \vec{y} = (v_1 w_3 - v_3 w_1) \vec{y} \\ A_3 \vec{z} = (w_1 u_2 - w_2 u_1) \vec{z} \end{cases}$ 可以 (x, y, z) 换为 (i, j, k) 成为 1 向量.

因此 $A[\vec{z}]$ 为一个向量, 定义为 $\vec{u} \times \vec{w}$ 较空不改变 \vec{z} 的向量.

例) $\text{if } A \quad A^* = \begin{bmatrix} 3 & 5 & 4 \\ 1 & 0 & 1 \\ 2 & 1 & 1 \end{bmatrix}$ find A^{-1}

$$\boxed{\text{口 } A^{-1} = \frac{\det A}{\det A} \begin{bmatrix} 1 & -1 & 5 \\ 1 & -5 & 1 \\ 1 & 1 & -5 \end{bmatrix}}$$

$|A|=0$, BP 将其 transform 成矩阵为 0, 被压缩, 难度减少, 但这样就有一个 input
多个 output, 不是一个函数可做的, 如矩阵 A 在这个压缩后的面上, 有 sol. $\exists A \text{ 使 } Ax = b$ 无解
(压缩)

Invertible & eigen

1. A_{nn} , $\det AT = \det A$, BP col operation \Leftrightarrow row operation 故可利用

2. Cramer's Rule. A_{nn} , 可逆, For any \vec{b} in \mathbb{R}^n , unique \vec{x} is

$$x_i = \frac{\det A_i(\vec{b})}{\det A}, i=1, 2, \dots, n, \text{ where } A_i(\vec{b}) \text{ is } A \text{ replacing } i\text{th by } \vec{b}$$

$$A \cdot I_i(\vec{x}) = A[\vec{e}_1, \dots, \vec{x}, \dots, \vec{e}_n] = [A\vec{e}_1, \dots, A\vec{x}, \dots, A\vec{e}_n] = [\vec{a}_1, \dots, \vec{b}, \dots, \vec{a}_n] = A_i(\vec{b})$$

$\det A \cdot \det(I_i(\vec{x})) = \det(A_i(\vec{b})) \Rightarrow \det(I_i(\vec{x})) = \vec{x}$ 即是 \vec{x} , 对称行适用, 无关行不出

3. A 非 singular

$$A^{-1} = \frac{\text{adj } A}{\det A}$$

jth col of A^{-1} is \vec{x} that satisfies $A\vec{x} = \vec{e}_j$, 而 A^{-1} i th entry $x_i = \frac{\det A_i(\vec{e}_j)}{\det A}$

note $\det A_i(\vec{e}_j) = (-1)^{i+j} \det A_{ij} = C_{ji}$. But $A^{-1} \in \mathbb{R}^{n \times n}$, $A^{-1} = \frac{1}{\det A} \begin{bmatrix} C_{11} & C_{12} & C_{13} \\ C_{21} & C_{22} & C_{23} \\ C_{31} & C_{32} & C_{33} \end{bmatrix}$
注意 $\text{adj } A \cdot A = A \cdot \text{adj } A = \det A \cdot I$ ($A \cdot \det A = \text{adj } A, \forall i \in A$)

4. eigenvalue 参见图

eigenvector of A_{nn} is non zero vector \vec{x} st. $A\vec{x} = \lambda \vec{x}$ for some scalar λ

λ eigenvalue of A_{nn} if \exists nontrivial sol \vec{x} for $A\vec{x} = \lambda \vec{x}$.

即 λ is eigenvalue iff $(A - \lambda I)\vec{x} = \vec{0}$ has nontrivial sol. The set of all sub is Null(A - \lambda I)

This set a subspace of \mathbb{R}^n , eigenspace of A corresponding to λ

1. eigenvalue of triangular matrix are entries on main diagonal BP [-1, 1] \lambda \neq 1, 2 \rightarrow 1

□ 由 triangular dot to diagonal $\neq 0$

2. 0 is eigenvalue iff A not invertible. [5]. T

□ $A\vec{v} = 0, \vec{v} \neq 0 \therefore A\vec{v} = 0$ nontrivial sol ($\vec{v} \neq 0$) $\therefore A$ 非 $1-1$, nullity $\neq 0$ 且非 full-rk

3. DEF polynomial $f(t) \in P[\mathbb{F}]$ split over \mathbb{F} iff

$$\exists a_1, \dots, a_n, c \in \mathbb{F}, f(t) = c(t-a_1) \dots (t-a_n)$$

(不重根)

e.g. $\lfloor (t^2+1) \rfloor (t-2)$ 不 split over \mathbb{R} , 但 split over \mathbb{C} $(t-i)(t+i)$

BP: 若 a_1, \dots, a_n 为 distinct $\neq 0$ (互不相等), \neq repeated.
algebraic multiplicity > 1

diagonalizable T characteristic polynomial split over \mathbb{F}

□ [5]. T show $f(\lambda) \neq 0$, 且对应 to ind. of. basis. \therefore 对某 $\beta, [T]_\beta = [\lambda_1, \dots, \lambda_n]$, 且 $f(\lambda) = (\lambda_1 - t) \dots (t - \lambda_n) = (-1)^n (t - \lambda_1) \dots (t - \lambda_n)$

* 即是 split 不- \neq diagonalizable.

4. Cayley-Hamilton \forall square A satisfy characteristic fm.

$$BP \quad f(\lambda) = (-1)^n \lambda^n + k_{n-1} \lambda^{n-1} + \dots + k_1 \lambda + k_0, \text{ 有 } (-1)^n A^n + k_{n-1} A^{n-1} + \dots + k_1 A + k_0 = 0$$

deli 得力

Date.

$$\text{Ex 1} A = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 0 \\ 2 & -1 & 8 \end{bmatrix} \text{ eigenvalue } = 2, \text{ find basis for eigenspace for } \lambda = 2$$

$\square A - 2I = \begin{bmatrix} 2 & -1 & 6 \\ 2 & 1 & 0 \\ 2 & -1 & 8 \end{bmatrix}, \therefore (A - 2I)x = 0 \begin{bmatrix} -2 & 1 & 6 \\ 0 & 0 & 0 \end{bmatrix} \xrightarrow{\text{to}} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_2 \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix}$

Ex 2 $A_{n \times n}$, 如 $\sum a_{ij} = s$, 该特征值为 s ; $B_{n \times n}$, 如 $\sum a_{ij} = s$, 该特征值为 s

$\square A \begin{bmatrix} 1 & 1 & \dots & 1 \end{bmatrix} = \begin{bmatrix} a_{11} + \dots + a_{1n} \\ a_{21} + \dots + a_{2n} \\ \vdots \\ a_{n1} + \dots + a_{nn} \end{bmatrix} = s \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$

应用. dynamic system. $A = \begin{bmatrix} .95 & .05 \\ .05 & .97 \end{bmatrix}$, dynamic system defined by $\vec{x}_{k+1} = A\vec{x}_k$, $\vec{x}_0 = \begin{bmatrix} .6 \\ .4 \end{bmatrix}$

1. $\det \begin{bmatrix} -.95-\lambda & .05 \\ .05 & .97-\lambda \end{bmatrix} \Rightarrow \lambda = 1/0.92$ 且 $\vec{v}_1 \begin{bmatrix} 3 \\ 5 \end{bmatrix}$ $\vec{v}_2 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ 为特征向量

$$\vec{x}_0 = c_1 \vec{v}_1 + c_2 \vec{v}_2 (\vec{v}_1, \vec{v}_2 \text{ basis for } \mathbb{R}^2) = \begin{bmatrix} \vec{v}_1 & \vec{v}_2 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$$

$$\text{系数} [c_1, c_2] = [\vec{v}_1, \vec{v}_2]^{-1} \vec{x}_0 = \begin{bmatrix} 3 & 1 \\ 5 & -1 \end{bmatrix}^{-1} \begin{bmatrix} .6 \\ .4 \end{bmatrix} = \begin{bmatrix} .125 \\ .225 \end{bmatrix}$$

$$\vec{x}_1 = A\vec{x}_0 = c_1 A\vec{v}_1 + c_2 A\vec{v}_2 = c_1 \vec{v}_1 + c_2 (.92) \vec{v}_2; \quad \vec{x}_2 = A\vec{x}_1 = c_1 \vec{v}_1 + c_2 (.92)^2 \vec{v}_2$$

$$\therefore \vec{x}_k = c_1 \vec{v}_1 + c_2 (.92)^k \vec{v}_2 = (.125) \begin{bmatrix} 3 \\ 5 \end{bmatrix} + (.225)(.92)^k \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad k \rightarrow \infty, \vec{x}_k = .125 \vec{v}_1$$

A 可为 Markov chain 的 migration matrix, \vec{x}_0 为人口, \vec{x}_k 为 k 年后人口, 可知会稳定下来

Ex 3 $A = \begin{bmatrix} 1 & 3 & 3 \\ 3 & -5 & 3 \\ 3 & 3 & 1 \end{bmatrix} \neq A^k; \quad \boxed{1} = \begin{bmatrix} 2 & 4 & 3 \\ -4 & -6 & -3 \\ 3 & 3 & 1 \end{bmatrix}; \quad C = \begin{bmatrix} 5 & -8 & 1 \\ 9 & 9 & 2 \end{bmatrix} \quad P = \begin{pmatrix} 1 & -1 & 1 \\ 1 & 5 & 1 \\ 1 & 1 & 1 \end{pmatrix}$

$\boxed{1} \det(A - \lambda I) = 0 \begin{cases} \lambda_1 = 1 & v_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \\ 2 & v_2 = \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix} \end{cases} \quad (v_1 \text{ is L.I.}) \quad \therefore A = PDP^{-1} \quad \therefore A^k = P D^k P^{-1}, D^k = \begin{pmatrix} 1 & & \\ & 2 & \\ & & 2 \end{pmatrix}$

2. B 有 $\lambda_1 = 1$, $v_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$, $\lambda_2 = \lambda_3 = -2$, $v_2 = \begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix}$ $\notin \mathbb{R}^3$ basis, 不可对角化

3. C 有 $\lambda = 5, 0, -2$ distinct, 可对角化

例 If $AP = PD$, D diagonal, then nonzero col of P must be eigenvector of A . \checkmark 常考证明

A has n L.I. eigenvectors, e.g. A^T 也是 \checkmark 因 $A^T A$ 可对角化, $A^T = (PDP^{-1})^T = (P^T)^{-1} D^T (P^T)$, 可对角化, 有

可逆与可对角化无矛盾. $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ 可对角化但不可逆; $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ 可逆但不可对角化

vs 6

Date.

1. Subspace $\{x \mid Ax=0\}$ set of all sols + a system of homogeneous linear equation \rightarrow Null space ①
 已知为 set of all L. Comb. of certain specified vectors \rightarrow Col space ②
 (1) : $Ax=0$, $x=0$ 为解; \bar{v}, \bar{v}' in Null, $A(\bar{v}+\bar{v}') = A\bar{v}+A\bar{v}'=0$; $c_1\bar{v}=A\bar{v}=0$
 (2) : 见前例, 令 $\bar{v}_1, \dots, \bar{v}_p$ span Col, weight 全为 0 则有; $\bar{v}_1, \bar{v}_2, \dots, \bar{v}_p$ 为 L. Comb.
 从 $\# \text{Null } A$ 的过程中, 得到 ① 简化为 ② (即 ① + ② 为 Sol 之和 - 些是 L. Comb)
 ①② K 代表基底者, ① 为 Kernel, ② 为 range

元

$$\left\{ \begin{bmatrix} a \\ b \\ c \end{bmatrix} : a+b+c=0 \right\} \text{ 为 subspace} \quad \text{由 homogeneous equation} \quad \left\{ \begin{bmatrix} b-2d \\ 5+d \\ b+3d \end{bmatrix} : b, d \in \mathbb{R} \right\} \times \text{ to get } \begin{bmatrix} b-2d \\ 5+d \\ b+3d \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow b=d, d \in \mathbb{R} \quad \text{且} \begin{bmatrix} 1 & 1 & -1 \\ 1 & 2 & -4 \\ 2 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 & 1 & -1 \\ 1 & 2 & -4 \\ 2 & 0 & -1 \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

例 用 subspace 解释为何 $\begin{cases} x_1 - 3x_2 - 3x_3 = 0 \\ -2x_1 + 4x_2 + 2x_3 = 0 \\ -x_1 + 5x_2 + 7x_3 = 0 \end{cases}$ 已知解为 $x_1=3, x_2=2, x_3=-1$ 且解的个数已为 3
 \square 由于 $\bar{x} \in \text{Null } A$, $A\bar{x}=0$ (且 $\text{Null } A$ 为 subspace)

同理, 且 Col Space 为 $\begin{cases} 5x_1 + x_2 + 3x_3 = 0 \\ -9x_1 + 2x_2 + 5x_3 = 1 \\ 4x_1 + x_2 - 6x_3 = 9 \end{cases}$ 有解时, $5x_1 + x_2 - 3x_3 = 0$ 也有解. $(\bar{x}) = \begin{bmatrix} 0 \\ 4 \\ 5 \end{bmatrix}$, 也在 Col A)

应用 IP_n polynomial space, $\vec{p}_n(t) = a_0 + a_1 t + \dots + a_n t^n$ 为 Vector Space, 在 statistical trend analysis (ATA)

(Lagrange Interpolation) 对 $C_0, \dots, C_n \in \mathbb{F}$ 且 distinct, DEF $f_i(x) = \frac{(x-C_0)(x-C_1)\dots(x-C_{i-1})(x-C_{i+1})\dots(x-C_n)}{(C_i-C_0)(C_i-C_1)\dots(C_i-C_{i-1})(C_i-C_{i+1})\dots(C_i-C_n)}$

$= \prod_{k=0, k \neq i}^n \frac{x-C_k}{C_i-C_k}$, 证 $\{f_0, \dots, f_n\} \rightarrow IP_n(\mathbb{F})$ 为 basis

$\square \sum_{i=0}^n a_i f_i = \text{zero poly}$, $\text{Rp}(t) \neq 0$, $\therefore \sum_{i=0}^n a_i f_i(C_j) = 0 \forall j = 0 \dots n$

但 $f_i(C_j) = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases} \therefore \sum a_i f_i(C_j) = a_j$ (只对 $i=j$ 为 1)
 $\therefore \{f_0, \dots, f_n\}$ 为 $IP_n(\mathbb{F})$ 的 basis

因此对 poly $p(x) \in IP_n(\mathbb{F})$, $p(x) = \sum_{i=0}^n a_i f_i$, 对 $j=0 \dots n$, $p(C_j) = \sum a_i f_i(C_j) = a_j$

$\therefore p = \sum_{j=0}^n p(C_j) f_j$, 可知对 poly 其对应的 n+1 distinct C_i , $p(C_i) = 0$.

且 $p = \sum a_i f_i = 0$, zero func!!

ex 1 已知 poly 通过 (1, 2) (2, 5) (3, 4), 找出 poly 的组合.

\square 对 poly p $\begin{cases} p(1)=1 \\ p(2)=5 \\ p(3)=4 \end{cases}$ 选 $\begin{cases} C_0=1 \\ C_1=2 \\ C_2=3 \end{cases}$ 则 $f_0 = \frac{(x-2)(x-3)}{(1-2)(1-3)} = \frac{-x^2+6x+5}{2}$, $p = f_0 + 5f_1 + 4f_2 = -3x^2+6x+5$

ex 2. $C_0 \dots C_n$ distinct, def $T: IP_n(\mathbb{F}) \rightarrow \mathbb{F}^{n+1}$, $f \mapsto \begin{pmatrix} f(0) \\ f(1) \\ \vdots \\ f(n) \end{pmatrix}$, 且 T isomorphism

\square Linearity 例证, 且 $\dim IP_n(\mathbb{F}) = n+1$ 令 $T(f) = 0$, \therefore 对各 C_i , $f(C_i) = 0 \therefore f$ 为 zero vec.

应用 Verify $S = \{1, t, t^2, \dots, t^n\}$ is a basis for IP_n . (S 不 standard basis)

\square S span IP_n ; $c_0 \cdot 1 + c_1 \cdot t + \dots + c_n \cdot t^n = \vec{0}(t) = 0$. 代数上多于 $n+1$ 个方程, 且 $c_0=c_1=\dots=c_n=0$, S L. Ind

Span

set of V vector by the basis of S - 基底

1-(The spanning set theorem) 若 finite set $S = \{v_1, \dots, v_p\}$, $\text{span}(S) = \text{Vec. Space } V$, 且 S 是 subset of V 的 basis

□ 若 $S = \emptyset$, $\text{span}(S) = \{0\}$, S 不为 basis. 若 S 有非 0 元, 构造 basis by 步骤 ④:

(4): pick v_i , note $\{v_i\}$ L.Ind, 可能找出来的加上可构成 L.Ind. 的 v_i, \dots 直到找到 $\beta = \{u_1, \dots, u_k\}$ L.Ind.

$\text{WTS } \beta$ basis, 即 β spans V : i) $\text{span}(\beta) \subseteq V$ 因 $\beta \subseteq S$, $\text{span}(\beta) \subseteq \text{span}(S) = V$

ii) $V \subseteq \text{span}(\beta)$ - 若 $v \in V$ 由定理, 自然 $v = \sum_{i=1}^k c_i u_i$, $v \in \text{span}(\beta)$, $v \notin S$, v 为 S 的 L.Comb. 其中 $c_i \in \mathbb{R}$.

iii) β L.Ind (已证), 由 ③, ④ 2 部份) $v \in \text{span}(\beta)$

eg: 有 $S = \left\{ \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$ span \mathbb{R}^3 . pick $\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} = 2 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$, 不 include, $\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ 不 include, $\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$ 不 include, $\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} + 2 \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$

JK: A 为出法变量, $\begin{pmatrix} 1 \\ 2 \\ 3 \\ 1 \\ 0 \\ 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 4 \\ 2 & 1 \\ 3 & 1 \end{pmatrix}$, $\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ exclude, basis = $\left\{ \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$

试 find basis for $x+2y+z=0$ in \mathbb{R}^3 .

□ 有 $S = \{1, 2, 1\}$ 为 homogeneous equation. Null(A), $A = [1 2 1]$, $\{[1], [0]\}$

let $v_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$, $v_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$, $v_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$, let H be set of vectors in \mathbb{R}^3 whose 2nd 3rd entry are equal

Then $\forall x \in H$ is L.Comb. of v_1, v_2, v_3 : $\begin{bmatrix} x \\ 1 \\ 1 \end{bmatrix} = s \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + (t-s) \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$

但 $\{v_1, v_2, v_3\}$ 不是 basis for H , 因 v_1, v_2, v_3 甚至不在 H 里, 基本上是 L.Ind 且 span

2. Coordinate

(线性关系也叫基底) 例 2 例题见回 E-5

比第 10.3, 2d-4 given T
unique. 见文字, 有 10 个 THM
show T given output, 有反向一个 T st.
basis \rightarrow output

$B = \{b_1, \dots, b_n\}$ 为 V 的 basis, 对 $x \in V$, 若 $x = \sum a_i b_i$, 令 word $[x]_B = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$, 简单上,

对 $T: V \rightarrow F^n$, $x \rightarrow [x]_B$ 为 linear 且 bijective

□ 对 $x = \sum a_i b_i$, $T(x+k)y) = \begin{bmatrix} a_1 + k\beta_1 \\ \vdots \\ a_n + k\beta_n \end{bmatrix} = T(x) + kT(y)$, \therefore linear; $1-1$, $\therefore T(x) = 0 = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \therefore a_i = 0 \therefore x = \sum a_i b_i = 0$

onto: $\dim V = n = \dim F^n$ (由 \dim 同构且 $1-1$ 保 \Rightarrow onto)

推广, 已知 vector 表示 Vector space - ele, 1D manifolds 都是基底 L TN $(T(v_i)) \in W$ (平行于 L) $\rightarrow \sum a_{ij} w_i$ ($a_{ij}: 1 \rightarrow m$ unique, 由 \dim T_i)

对 $T: V \rightarrow W$, $B = \{v_1, \dots, v_n\}$, $V = \{w_1, \dots, w_m\}$, 对 $v_j (j: 1 \rightarrow n) \rightarrow \sum a_{ij} w_i$ ($a_{ij}: 1 \rightarrow m$ unique, 由 \dim T_i)

把 a_{ij} 放进 matrix $(m \times n) \times [T]_B^r$, 对 $[x]_B$, $T([x]_B) = [x]_r$, $[T]_B^r = [[T(v_1)]_r, \dots, [T(v_n)]_r]$

对 $x = \sum v_i$, $T(x) = \sum T(v_i)$, 同 $[T]_B^r[x]_B = [[T(v_1)]_r, \dots, [T(v_n)]_r]$ ($[x]_B = x, [T(v_i)]_r$)

3-10 例 4 总可用 matrix 表示 vector space, 通过映射到 F^n back L.Ind 时 $\exists f_1(x) + f_2(x) + f_3(x) = 0$ 且 $f_1(x) + f_2(x) + f_3(x) = 0$ 但 $f_1(x) + f_2(x) + f_3(x) = 0$ 且 $f_1(x) + f_2(x) + f_3(x) = 0$

即 \exists coordinate mapping \cong isomorphism from V to \mathbb{R}^n , 是坐标的维度不同, 但 act alike

(3) i) $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2, (a_1, a_2, a_3) \rightarrow (2a_1 + 3a_2 - a_3, a_1 + a_3)$, 找 $[T]_B^r$, ii) $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3, [a_1, a_2, a_3] \rightarrow \begin{bmatrix} a_1 - a_2 \\ a_1 \\ 2a_1 + a_2 \end{bmatrix}$, $\beta = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$, $T = \begin{bmatrix} 2 & 3 & -1 \\ 1 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}$

□ i) $[T]_B^r = \begin{bmatrix} 2 & 3 & -1 \\ 1 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 3 & -1 \\ 1 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}$

3. DEF $\mathcal{L}(V, W)$ 为 all L.T. from V to W , 且其为 V .S. (addition, scalar multiply, \neq af, \neq f.g. 互不相容)

(矩阵表示法也叫 linear) 对 $T, U \in \mathcal{L}(V, W)$: $[T+U]_B^r = [T]_B^r + [U]_B^r$, $[kT]_B^r = k[T]_B^r$ (Li 得力)

□ i) 有 a_{ij}, b_{ij} st. $T(v_j) = \sum a_{ij} w_i$, $U(v_j) = \sum b_{ij} w_i$, $(T+U)(v_j) = \sum (a_{ij} + b_{ij}) w_i \therefore a_{ij} + b_{ij} = \text{右}$

ii) $kT(v_j) = k \sum a_{ij} w_i \therefore \text{左} = k a_{ij} = \text{右}$

***3

VS 7 & dim 3

Date.

DEFINITION 1

\mathcal{T}_1 $\beta = \{s_1, \dots, s_n\}$ is V.S. if subset. β is basis $\Leftrightarrow \forall v \in V, v \notin \beta \Rightarrow \text{unique L.comb.}$

$\square \Rightarrow \exists P_2 P_{iv} \Leftarrow$ i) β L.ind. of V . \Rightarrow 表示部不同基底. $\text{coeff} = 0$
ii) span, that is $\forall v \in V$ has L.comb.

\mathcal{T}_2 (Replacement) if V.S. V , if n -ele G , $\text{span}(G) = V$, if m -ele L .Ind. L , $\forall L \subseteq V$, $\exists H \subseteq G$, $n-m$ ele L .Ind. H from V

\square induction on m : $m=0$, $L=\emptyset$, take $H=G$; if $m+1$, $L=\{v_1, \dots, v_m\}$, $\exists \{v_1, \dots, v_m\}$ L.ind. \Rightarrow $\{v_1, \dots, v_m\}$ L.ind. by inductive hypothesis.

\square $\{u_1, \dots, u_{m+1}\} \subseteq G$ 且 $\{v_1, \dots, v_m\} \cup \{u_{m+1}\}$ span V . \Rightarrow $\forall v \in V, a_1v_1 + \dots + a_mv_m + b_1u_{m+1} = v_{m+1}$.
 \square 若 $a_1, \dots, a_m \neq 0$ (若 $a_1, \dots, a_m = 0$, $v_{m+1} \in \text{span}(\{v_1, \dots, v_m\})$), $L=\{v_1, \dots, v_m\} \cup \{u_{m+1}\}$ L.dep, $\exists H$, $\exists b_1 \neq 0$, $b_1 \neq 0$, $u_1 = (-b_1^{-1}a_1)v_1 + \dots + (-b_1^{-1}a_m)v_m$.
 \square \Rightarrow $\{u_1, \dots, u_{m+1}\} \not\subseteq L$.Ind. $\Rightarrow n-m-1$ -ele $H=\{u_2, \dots, u_{m+1}\}$, $\text{span}(LUH)=V$.

\mathcal{P} Replacement proof
a.) If vector space V basis $\beta = \{b_1, \dots, b_n\}$, then any set in V with more than n vectors is L.dep

\square i) 若 L hol. 对大于 n 的子集 S , S 有 n ele. $\Rightarrow S$ L.dep. 由 \mathcal{T}_2 , $n+1 \leq n$, $\exists H$

ii) $\omega_1 > n$, 为 free variable, L.dep \mathcal{T}_3 , 4. \mathcal{T}_4

b) V has basis n vector, then V basis has n vectors

\square 若 β, β_2 是 basis, β_1 n vector, β_2 m vector, 若 $m > n$, i) β_2 L.dep. $\therefore m \leq n$. \square ii) β_1 L.dep. $\therefore n \leq m \therefore m = n$

c) V L.ind. subset L 可以扩展为 V 的 basis

\square 若 $\dim(V) = n$, $|L| \geq n$, 由 a) L.dep; 若 $|L| \leq n$, 有 $H \subseteq \beta$ st L span V , $|LH| \leq n$

i) (basis theorem) $\dim(V) = n$, 由 V has spanning set at least n ele, $\exists n$ ele β 为 V basis;

ii) L Ind L 有 n ele $\Rightarrow V$ has basis β * in vector in \mathbb{R}^n span \mathbb{R}^n 为 basis

\square i) Let G spanning set, $H \subseteq G$ basis, $|H| = n$ $|G| \geq |H|$, $|G| \geq n$. 若 $|G| = n$, $G = H$

ii) β 不是 G 的 spanning set, 有 $n-n=0$ subset H st. L span V . $\therefore L$ span \neq L.ind. L 为 basis

\mathcal{T}_3 $W \subseteq V \Rightarrow \dim(W) \leq \dim(V)$, β bases for W . β extend to bases for V 若 $W = V$ 则 β 为 V 的 basis

\square i) let $\dim(V) = n$, if $W = \{0\}$, $\dim(W) = 0 \leq n$; if $\exists x, x \in W \wedge x \neq 0$, 由 2.1 节定理 ④ β 为 L .ind. $\{x_1, \dots, x_k\}$, 由 P_a , $k \leq n$ 且 L .Dep. 全 x_i L.dep, $\forall x_i \in W$, $x_i \in L$.dep, $\therefore x_i \in \text{span}(x_1, \dots, x_k)$, $\therefore \text{span}(x_1, \dots, x_k) = W$, basis

ii) $\square P_b$

3) 1. $\exists \{u_1, \dots, u_t\}, \{v_1, \dots, v_t\}$ s.t., u_i 为 V 的 L.comb. $\forall L$ L.dep \neq P_d

\square $\text{span} U \subseteq \text{span} V$, $\forall \text{span} U, \text{span} V$ vec space. $\therefore \text{span} U \subseteq \text{span} V$, $\dim(\text{span} U) \leq \dim(\text{span} V) \leq t$
 \square \mathcal{T}_3 为真, 但因 V 不是 L.ind, 不是 $[u_1, \dots, v_t]$ 的 L.ind. $\therefore L$.dep $\leq t$

$\exists \{v_1, \dots, v_p\}$ span V , $\dim V \leq p$ ✓; $\{v_1, \dots, v_p\}$ L.ind in V , $\dim V > p$ ✓

$\exists \{v_1, \dots, v_p\}$ L.dep in V , $\dim V \leq p$ ✗; If any set of p element in V fail to span V , $\dim V > p$ ✓

3) 3. $\dim(d_1, d_2, d_3, d_4)$ L.dep; $\forall d_1, d_2, d_3 < d_4$, $d_4 \notin \langle d_1, d_2, d_3 \rangle$

\square 有 $c_1, c_2, c_3, c_4 \in \mathbb{C}$, st. $c_1d_1 + c_2d_2 + c_3d_3 = 0$, $c_4 \neq 0$, 有 d_4 L.dep $\therefore d_4 = \frac{c_2}{c_4}d_2 + \frac{c_3}{c_4}d_3$

\square $\exists a_4 \in \text{span}(d_1, \dots, d_3) = \text{span}(d_2, d_3)$ $\therefore d_4$ L.dep. $\therefore \dim(d_1, \dots, d_4) < 4$ $\square \mathcal{T}_2$

Rank 1

$\exists A \in M_{n,n}(\mathbb{R})$ such that $A = A^T$

~~打个~~ $\forall x \in \mathbb{R}^n, Ax = u \Leftrightarrow x = V^{-1}u$

\Rightarrow ist $A \in \mathbb{R}^{n \times m}$, $x \in \mathbb{R}^m$, $a \in \mathbb{R}^n$: $\exists k \in \mathbb{N}$ mit $A = a a^T$

217. 4 equation 42 variable to homogeneous system 有 2 答 (不是 multiply), 其他 4 未知數 2 答

It is a nonhomogeneous system with some of well mixed to 33.

A homogeneous system with same of rank. rank of system is n^2 since n^2 equations have n^2 unknowns.

$\boxed{A_{4 \times 42}}$ 由 $Ax = 0$ 的解集的维数为 2, nullity $= 2$, $\text{rk } A = 40$. $\therefore \text{Col } A = \mathbb{R}^{40}$. $\forall x, Ax = 5$ 有 5 个

$$T_4 \text{ Ark} = k, \text{ so } A = \sum_{i=1}^k \text{rk}=1-\text{matrix} \quad \square \quad A = U \Sigma V^T, \text{ so } \Sigma = \sum_{i=1}^k \underline{\Sigma_i}, \quad A = \sum_{i=1}^k U \Sigma_i V^T$$

T_5 (Eckart-Young) 當 $r=r_A$ 且 $k=k_B$, $\sum_{i=1}^r \hat{u}_i u_i^\top$ 約等於 A 而 $k=k_B$ 時， $\hat{A}(k)=\sum_{i=1}^k \hat{u}_i u_i^\top$

1) $\hat{A}(k) = \text{argmin}_{\hat{A}} \| A - \hat{B} \hat{A} \|_2$ EP rk \hat{A} approx; 在 t_2 次方 norm 是 1

$$\text{ii) } HA - A^T A^{-1} = \sigma_{k+1}$$

应用: Rank 1 特征值分解 / singular value decomposition [1]

Let $A_{2 \times 3}$, $\text{rank}(A) = 1$. The first col of A , $\vec{u}_1 \neq 0$. There exists $\vec{v} \in \mathbb{R}^3$ s.t. $A = \vec{u}_1 \vec{v}^T$

□ rank $A=1$, \vec{u}_1 为 basis for $C(A)$, $\vec{v}_1 = x\vec{u}_1$, $\vec{u}_2 \equiv y\vec{u}_1$, $\vec{v} = \begin{bmatrix} x \\ y \end{bmatrix}$

(also $\vec{u}_1 = \vec{0}$, $T_{\vec{u}_1} \vec{u}_2$ is basis, & $\vec{u}_1 = x\vec{u}_2$, $\vec{v} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$; $\vec{u}_1, \vec{u}_2 = \vec{0}$, \vec{u}_2 is basis, $\vec{v} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$)

PP A has rank 1 iff outer product $\tilde{u}\tilde{v}^T$, i.e. $A = \tilde{u}\tilde{v}^T$ for $\tilde{u} \in \mathbb{R}^m$, $\tilde{v} \in \mathbb{R}^n$

$x = a\beta_1 + b\beta_2 + c\beta_3$ 新的 x 相当于 $ar_1 + br_2 + cr_3$ 在本质上也只是 identity map \hookrightarrow 但不会对 value 做什么, 而是只转换为另一组表示 $(r_1, r_2, r_3) \sim (r_1, r_2, r_3)$

3. Basis change 若 given V , 有 $B = \{x_1, \dots, x_n\}$, $B' = \{x'_1, \dots, x'_n\}$, 求了找出 过渡 (change) matrix, $\det B' - x_i$

Prove I_V bijection, thus exists inverse. and $[v]_p = [I_V(v)]_\beta = [I_v]_p^\beta [v]_\beta$

之所以這樣 $\det Q$ 為 $I_1^{\alpha_1} I_2^{\alpha_2} \cdots I_n^{\alpha_n}$, 因現定理知 $[v]_{\rho'} \rightarrow [v]_{\rho}$ 而 $[v]_{\rho} = [T(v)]_{\rho} = Q[v]_{\rho}$

若已有 $T: V \rightarrow V$ 在 β 中，知道 T 在另一 basis 的表现行 expression:

$$\text{由 } \begin{bmatrix} T \end{bmatrix}_\beta = (Q^{-1}[T]_\beta Q)^\beta = [I_V]_\beta^\beta [T]_\beta = [I_V T]_\beta^\beta = [T I_V]_\beta^\beta = [T]_\beta, \text{ 且 } \\ \text{左邊 } \xrightarrow{\text{PPT, 右} \rightarrow \text{左边}, \beta \rightarrow \beta' \rightarrow \beta} \xrightarrow{\square} Q[T]_{\beta'} = [I_V]_{\beta'}^\beta [T]_\beta = [I_V T]_{\beta'}^\beta = [T I_V]_{\beta'}^\beta = [T]_{\beta'}, \text{ 且 } \\ \text{右邊 } \xrightarrow{\beta' \rightarrow \beta} \xrightarrow{\beta \rightarrow \beta} \xrightarrow{\square} \text{得证.}$$

(2) 指出该 $y=2x$ 对称轴为 $(\text{mer } T \text{ 且})$
 $y=2x$ 为 $(1, 2)$ 的直角坐标系中的一条直线，且 $T(1, 2) \rightarrow (1, 2)$ 。
 $[T]_B = [A^{-1} D E]_{P \rightarrow B}$ $P \in B$

$$\text{打} (2,1) \text{ 太阳点, note } (1,2)(\text{been Indepdt. 且 } (1,2) \rightarrow (2,-1)) \text{ 为 } \beta' \text{ 的一个倍数} \\ T(-2,1) \rightarrow (2,-1) \quad [1,0] = Q^{-1} \cdot [1,2]$$

$$***5 \text{ bwdZ \& Decomposition}$$

$\beta = \sum_{i=1}^n v_i u_i' = [v_1 \dots v_n] [u_1 \dots u_n]' = [v_1 \dots v_n] P = Q [u_1 \dots u_n]$
 $P = [u_1 \dots u_n]$
 $Q = [v_1 \dots v_n]$
 $P^{-1} = [u_1 \dots u_n]$
 $Q^{-1} = [v_1 \dots v_n]$
 $X = \sum x_i v_i = [x_1 \dots x_n] Q$
 $x_i = \sum v_i' u_i = [v_1 \dots v_n] P^{-1} X$
 $P^{-1} = [u_1 \dots u_n]$
 $Q^{-1} = [v_1 \dots v_n]$
 $T: V \rightarrow W, \exists [T]_P^r = P^{-1} [T]_P^r Q$
 $[T]_P^r = [I_W T I_V]_P^r = [I_W]_P^r [T]_P^r [I_V]_P^r = P [T]_P^r Q$
 $R Q = S$
 $[I_V]_P^r [I_V]_P^r = [I_V]_P^r = S$

应用 LU Factorization (pol. Cholesky, Option Pg. 1)

i. If A_{mn} 可以元数 interchange 而化为 echelon form, D^n

$$A = \begin{pmatrix} 1 & \dots & 0 \\ 0 & \dots & 1 \\ \vdots & \ddots & \vdots \end{pmatrix} \begin{pmatrix} D & & \\ & E & \\ & & U \end{pmatrix}$$

A 可由 elementary row operation reduce to U
 L 由 diagonal = zero, $E_p F_p \rightarrow E, A \rightarrow U$, $\therefore A = (E, U)$
 L product 由上而下
 U echelon form
 A 可能

P.i) LU 不唯一, 但若 require diag(L)=1..1, 用 $D \neq I_n$, $A = LU = LD^{-1}U$

4D 例 U, D 为 L 的 $D^{-1}U$ 为上 $\therefore D$ 为 L 的 diag matrix 由 $D \neq I_n$, $D_{ii} \neq 1$, $(LD)_{ii} = \sum L_{ij} D_{jj} = D_{ii} \neq 1$

$$\text{ii) } \det A = \prod_{i=1}^m U_{ii} \quad \boxed{\det L \det U = 1 \cdot \prod U_{ii}}$$

iii) solve Sys Eq $Ax=b$ 方便: i) solve $-Ly = b$ 用 forward backward sub 由 L 为, ii) solve $-Lx = y$

iv) 若 A 为 LU $\left(\begin{smallmatrix} L & U \\ 0 & 1 \end{smallmatrix}\right)$. 若有 LU 可能 at least 1 4LU 不可能

v) (PLU) 为满元 w/ pivot 有 $PA = LU$

$$\text{例 1. } A = \begin{pmatrix} 3 & -3 & 9 & 3 \\ -6 & 4 & 1 & -18 \\ 1 & -2 & 2 & 4 \\ 1 & 2 & -4 & 6 \end{pmatrix} \xrightarrow{\text{max 1st col } P_1 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}} \begin{pmatrix} 6 & -2 & 2 & 4 \\ -6 & 4 & 1 & -18 \\ 1 & -2 & 2 & 4 \\ 1 & 2 & -4 & 6 \end{pmatrix} \xrightarrow{E_1 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ -1/6 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}} \begin{pmatrix} 6 & -2 & 2 & 4 \\ 0 & 2 & 3 & -14 \\ 0 & -12 & 8 & 1 \\ 0 & 4 & 2 & 2 \end{pmatrix} \xrightarrow{\text{scale factor 1/2}} \begin{pmatrix} 6 & -2 & 2 & 4 \\ 0 & 1 & 3/2 & -7/2 \\ 0 & -12 & 4 & 1 \\ 0 & 4 & 1 & 1 \end{pmatrix} \xrightarrow{\text{scale factor 1/12}} \begin{pmatrix} 6 & -2 & 2 & 4 \\ 0 & 1 & 1/2 & -1/2 \\ 0 & -12 & 1 & 1/2 \\ 0 & 4 & 1 & 1/2 \end{pmatrix} \xrightarrow{\text{scale factor 1/12}} \begin{pmatrix} 6 & -2 & 2 & 4 \\ 0 & 1 & 1/2 & -1/2 \\ 0 & -12 & 1 & 1/2 \\ 0 & 4 & 1 & 1/2 \end{pmatrix}$$

$$\begin{aligned} P_2 &= \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \xrightarrow{\text{3rd row, } P_3 = 1} \begin{pmatrix} 6 & -2 & 2 & 4 \\ 0 & 1 & 1/2 & -1/2 \\ 0 & -12 & 1 & 1/2 \\ 0 & 4 & 1 & 1/2 \end{pmatrix} \xrightarrow{E_3 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}} \begin{pmatrix} 6 & -2 & 2 & 4 \\ 0 & 1 & 1/2 & -1/2 \\ 0 & 0 & 1 & 1/2 \\ 0 & 0 & 0 & 1 \end{pmatrix} \\ &\xrightarrow{\text{scale factor 1/12}} \begin{pmatrix} 6 & -2 & 2 & 4 \\ 0 & 1 & 1/2 & -1/2 \\ 0 & 0 & 1 & 1/2 \\ 0 & 0 & 0 & 1 \end{pmatrix} = U \end{aligned}$$

swap along 3消元 1消元 first (compute) 3消元.

if find new elementary \tilde{E} st. $\tilde{E}_3 \tilde{E}_2 \tilde{E}_1 P_3 P_2 P_1 A = U$, 有 $\begin{cases} \tilde{E}_1 = P_3 P_2 E_1 P_1 \\ \tilde{E}_2 = P_3 P_2 P_1 \\ \tilde{E}_3 = P_3 P_2 P_1 \end{cases} = P_3 P_2 E, P_2 \neq P_3$

$$\therefore PLU \text{ 有 } \begin{cases} P = P_3 P_2 P_1 \\ L = \tilde{E}_1^{-1} \tilde{E}_2^{-1} \tilde{E}_3^{-1} \\ U = \tilde{E}_3 \tilde{E}_2 \tilde{E}_1 P_1 A \end{cases}$$

$$\text{例 2. LU on } \begin{pmatrix} 2 & 4 & -1 & 5 & -2 \\ -4 & -5 & 3 & 8 & 1 \\ 2 & 5 & 4 & 1 & 8 \\ 0 & 0 & 7 & -3 & 1 \end{pmatrix} \sim \begin{pmatrix} 2 & 4 & -1 & 5 & -2 \\ 0 & 1 & -1 & 3 & 1 \\ 0 & 0 & 1 & 2 & -3 \\ 0 & 0 & 0 & 4 & 7 \end{pmatrix} \sim \begin{pmatrix} 2 & 4 & -1 & 5 & -2 \\ 0 & 1 & -1 & 3 & 1 \\ 0 & 0 & 1 & 2 & -3 \\ 0 & 0 & 0 & 1 & 5 \end{pmatrix} = U$$

$\therefore \text{L} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

例 3 LU 快速解 A^{-1} ; find $Ax = e_1$, 即化为解 sys eq

$Ax = e_1$

\therefore SGD - UV decompose, $A = UV$, SGD min $\|UV - A\|_2^2$ when U, V free param

1.9) Leontief Input-Output Model

1. econ divided in n sectors that produce goods & services $\vec{x} \in \mathbb{R}^n$ production vector
 open sector consume goods & services, \vec{d} final demand vector, \vec{c} goods & services demands by open sector
 producer 为了生产, 需要 input, intermediate demand
 Leontief 假设 $\vec{x} = \vec{c}$. Intermediate demand + $\vec{d} = \vec{x}$, 完全平行
 为使, 假设 for each sector, \exists unit consumption vector in \mathbb{R}^n , input needed per unit of output
 可得 intermediate demand = $x_1 \vec{c}_1 + x_2 \vec{c}_2 + \dots = C\vec{x}$, $C = [\vec{c}_1 \vec{c}_2 \dots]$, \vec{c}_i i th sector
 这样 $\vec{x} = C\vec{x} + \vec{d} \Rightarrow (I - C)\vec{x} = \vec{d}$

2. 例) 一 econ 有 manufacturing, Agriculture, Services $\equiv 3$ sectors.

Purchased from	M	A	S	Inputs consumed per Unit of Output	to M 生产 100 units, 需要
M	.50	.20	.10	1	.50
A	.40	.30	.10	.20	.20
S	.20	.30	.30	.10	.10

由 $C = \begin{bmatrix} .50 & .40 & .20 \\ .20 & .30 & .10 \\ .10 & .10 & .30 \end{bmatrix}$

由 $\vec{d} \rightarrow 50M, 30A, 20S$ - \vec{x} 可用以下求到

$$I - C = \begin{bmatrix} .5 & -.4 & -.2 \\ -.2 & .7 & -.1 \\ -.1 & -.1 & .7 \end{bmatrix} \quad [I - C] \vec{d} = [I \vec{x}] \quad \vec{x} = \begin{bmatrix} 119 \\ 78 \\ 78 \end{bmatrix}, \text{ 生产水平}$$

3. 实际中又必须 economically feasible, 即无负数 entries; col sums of $C < 1$, 因 sectors 应 input < output
 且如 C, \vec{d} 无 negative entries 且 col sums of $C < 1$, $I - C$ 有逆.

假设 \vec{d} 确定, industries 从 \vec{x} 生产水平正好为 \vec{d} , 即生产的正好被全消耗, 由
 intermediate demand $\vec{C}\vec{d}$, 而 $\vec{C}\vec{d}$, 需 intermediate demand $C(C\vec{d}) \dots$

$$\vec{x} = \vec{d} + C\vec{d} + C^2\vec{d} \dots = (I + C + C^2 + \dots) \vec{d}, \text{ 根据 } (I - C)(I + C + C^2 + \dots + C^{m-1}) = I - C^{m+1} \quad ①$$

If col sum of $C < 1$; $C^m \rightarrow 0$: $I - C^{m+1} \rightarrow I$

$$\text{因此 由 } ① (I - C)^{-1} = I + C + C^2 + \dots + C^{m-1}$$

4. $\vec{d} \rightarrow \vec{d} + \alpha\vec{d}$, $\vec{x} \rightarrow \vec{x} + \alpha\vec{x}$, 在 production level change with demand $(\vec{d} + \alpha\vec{d} = (I - C)\vec{x} + \alpha(I - C)\vec{x})$

而如 $\alpha\vec{d}$ 只有 3 sectors 提高需求, $[I - C] \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} (I - C)^{-1} \vec{d} \\ \vdots \\ (I - C)^{-1} \vec{d} \end{bmatrix} = \vec{x}$, 即 3 个 sector 提高, 每个 sector 都提高生产率

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notable for $\mathbb{R}^3 \rightarrow \mathbb{R}^2$, 但有 SR 的时候, 对应是 (单) 的 \mathbb{R}^2 ele.
有 grid map, 且我们自己关心其余的 \mathbb{R}^2

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左用 固像子

$$L \begin{array}{|c|c|c|c|} \hline & 0 & 0 \\ \hline 0 & 0 & 0.5 & 0.5 \\ \hline 0 & 0 & 0.12 & 0 \\ \hline 1 & 0 & 0 & 1 \\ \hline 4 & 0 & 0 & 1 \\ \hline \end{array} D = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0.5 & 0.5 & 0 \\ 0 & 0 & 0.12 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0.25 \\ 0 & 1 \end{bmatrix} \text{ shear transform}$$

2. 但是 matrix multiplication 也即 linear Transformation 不能直接平移, 因此引入 homogeneous coordinate

(x, y) in \mathbb{R}^2 可视作 $(x, y, 1)$ in \mathbb{R}^3 , 且 $(x, y) \mapsto (x+h, y+k)$ 可用 $(x, y, 1) \mapsto (x+h, y+k, 1)$

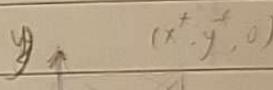
standard matrix for L.T. $\begin{bmatrix} 1 & 0 & h \\ 0 & 1 & k \\ 0 & 0 & 1 \end{bmatrix}$ ($\begin{bmatrix} 1 & 0 & h \\ 0 & 1 & k \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} x+h \\ y+k \\ 1 \end{bmatrix}$)

而非平移, RP L.T. in \mathbb{R}^2 及 \mathbb{R}^3 都是 $\begin{bmatrix} A & v \\ 0 & 1 \end{bmatrix}$, A 为 \mathbb{R}^2 或 \mathbb{R}^3 to standard matrix

to $\begin{bmatrix} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$ counter-clockwise rotation $\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ reflection $y=x$ $\begin{bmatrix} s & 0 & 0 \\ 0 & t & 0 \\ 0 & 0 & 1 \end{bmatrix}$ scale x by s, y by t

3. 3D TS 理, 假设为 (X, Y, Z, H) if $H \neq 0$, 且 $X = \frac{X}{H}, Y = \frac{Y}{H}, Z = \frac{Z}{H}$ 为 homogeneous

e.g. $(10, -6, 14, 2) \mapsto (-10, 6, -14, 1)$ 即为 $(5, -3, 7)$ to homogeneous



4. 3D 的投影到 2D 平面

即 (x, y, d) 者, 通过 (x, y, z) 到 viewing plane $(x^*, y^*, 0)$

$$\text{即 } \frac{x^*}{d} = \frac{x}{d-z}, x^* = \frac{x}{1-\frac{z}{d}}, \text{ 同理 } y^* = \frac{y}{1-\frac{z}{d}}$$

$$(x, y, d) \quad \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1/d \\ 0 & 0 & -1/d & 1 \end{bmatrix} \quad (x^*, y^*)$$

$$\text{即 } \begin{bmatrix} x^* \\ y^* \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 0 \\ 1 \end{bmatrix} \quad \begin{bmatrix} x \\ y \\ 0 \\ 1 \end{bmatrix} \in \mathbb{R}^4 \quad \begin{bmatrix} x \\ y \\ 0 \\ 1 \end{bmatrix} \mapsto \begin{bmatrix} x^* \\ y^* \\ 0 \\ 1 \end{bmatrix} \in \mathbb{R}^4$$

2. 用 difference equation

1. vector space S of discrete-time signal 为 signal, function sampled at discrete time

to form (y_0, y_1, \dots) 且 y_i 为 \mathbb{F} . 记为 $\{y_k\}$ $\begin{array}{ccccccc} \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ y_k = (-1)^k & & & & & & \text{intervals} \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ s_{y_k} = (1, -1, 1, -1, \dots) \end{array}$

L. hd for 3 signals $\Rightarrow c_1 u_k + c_2 v_k + c_3 w_k = 0$ for all k 有 $c_1 = c_2 = c_3, k \in \mathbb{Z}, \forall k \geq 0$

无论是否 L.hd, 上式 k 也可写为 $k+1, k+2, \dots$ 由 $\begin{bmatrix} u_k & v_k & w_k \\ u_{k+1} & v_{k+1} & w_{k+1} \\ u_{k+2} & v_{k+2} & w_{k+2} \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$

左 matrix \rightarrow Cassonati matrix of signal, det \rightarrow Cassonati

If Cassonati matrix $\text{J} \neq 0$ for at least one value of $|c|$, $\Rightarrow c_1 = c_2 = c_3 = 0, \{u_k\}, \{v_k\}, \{w_k\}$ L.hd

2. given signal $\{z_k\}$, $a_0 y_{k+n} + a_1 y_{k+n-1} + \dots + a_{n-1} y_{k+1} + a_n y_k = z_k$ for all k

\Rightarrow linear difference / linear recurrence equations of order n . If $\{z_k\}$ zero sequence, homogeneous
在 digital signal processing, linear difference equation is linear filter, a_0, \dots, a_n \Rightarrow filter coefficients
 $\{y_k\}$ input, $\{z_k\}$ output, $\{z_k\} \neq 0$ \Rightarrow filtered out, $\{y_k\}$ goes sol

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3. Sol for homogeneous difference equation has form $y_k = r^k$ for some r .

Eg. Find $y_{k+3} - 2y_{k+2} - 5y_{k+1} + 6y_k = 0 \quad \forall k$.

$$r^k(r-1)(r+2)(r-3)=0, \quad r^k, (-2)^k, 3^k \text{ are sol.}$$

$\begin{aligned} & r^k - 2r^{k+2} - 5r^{k+1} + 6r^k = \\ & 3^k(27-1+(-5)+6) = 0 \end{aligned}$

4. Given $a_1 \dots a_n$, mapping $T: S \rightarrow S$, $S \subseteq \mathbb{R}^n$ by $w_k = y_{k+n} + a_1 y_{k+n-1} + \dots + a_n y_k$

\exists T linear, \exists sol for homogeneous ($\{w_k\} = \{0\}$) $\Rightarrow T$ h.s. kernel

(1) If $a_n \neq 0$, and given $y_0 \dots y_{n-1}$ specified, $y_{k+n} + a_1 y_{k+n-1} + \dots + a_n y_k = z_k$ for all k has unique sol if $y_0 \dots y_{n-1}$ specified.

□ Define $y_n = z_0 - [a_1 y_{n-1} + \dots + a_{n-1} y_1 + a_n y_0]$, $\forall k \in \mathbb{Z}$ define $y_{n+k} =$

2' set H of all sols of $y_{k+n} + a_1 y_{k+n-1} + \dots + a_n y_k = 0$ for all k is n -dim Vector

□ H 为 S subspace $\Leftrightarrow T$ h.s. kernel, $\forall \{y_k\} \in H$, $F: H \rightarrow \mathbb{R}^n$, $\{y_k\} \mapsto (y_0, y_1, \dots, y_{n-1})$
可证 F linear. T . 由 (1), \exists unique $\{y_k\}$ in H st. $F(\{y_k\}) = (y_0, y_1, \dots, y_{n-1})$, F onto, F is isomorphism, $\dim H = \dim \mathbb{R}^n = n$ (由上题推得)

(3) find basis for sol of all sols for $y_{k+3} - 2y_{k+2} - 5y_{k+1} + 6y_k = 0 \quad \forall k$

由 $\{1^k, (-2)^k, 3^k\}$ 为基, 且由 \mathbb{Z}^3 (Cartesian, $k=0 \begin{bmatrix} 1 & 1 & 1 \\ 1 & -2 & 3 \\ 1 & 4 & 9 \end{bmatrix}$) 可证, L.h.d
any L.h.d of 3 vectors in 3-dim space is basis.

*** 9.2
basis 3

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例 1. 算 $\begin{pmatrix} Q \\ P \end{pmatrix}$

$$i) v_1' = v_1 + 3v_2 - 4v_3$$

$$v_2' = 2v_1 - v_2 + 5v_3$$

$$v_3' = 4v_1 + 5v_2 + 3v_3$$

$$ii) v_1 = v_1' + v_2' + 3v_3'$$

$$v_2 = 2v_1' - v_2' + 4v_3'$$

$$v_3 = 3v_1' + 5v_3'$$

注 2

$$\square) Q = [I_{v_i}]_{P \rightarrow P'}^T = (R^{-1}) = \begin{pmatrix} 1 & 2 & 4 \\ 3 & -1 & 5 \\ 4 & 5 & 3 \end{pmatrix}^{-1}$$

$$ii) Q = \begin{pmatrix} 1 & 2 & 4 \\ 3 & -1 & 5 \\ 4 & 5 & 3 \end{pmatrix}$$

$$iii) i) A = [v_1' \dots v_3']^T [v_1 \dots v_3], \text{ 由 } \begin{pmatrix} 1 & 2 & 4 \\ 3 & -1 & 5 \\ 4 & 5 & 3 \end{pmatrix} \begin{pmatrix} v_1' \\ v_2' \\ v_3' \end{pmatrix} = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} \Rightarrow [v_1 \dots v_3] \begin{pmatrix} 1 & 2 & 4 \\ 3 & -1 & 5 \\ 4 & 5 & 3 \end{pmatrix}^T = [v_1' \dots v_3']$$

$$iv) \begin{pmatrix} 1 & 2 & 4 \\ 3 & -1 & 5 \\ 4 & 5 & 3 \end{pmatrix} \begin{pmatrix} v_1' \\ v_2' \\ v_3' \end{pmatrix} = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$$

注 1. ~~for~~ \mathcal{B} 为正交基 for \mathbb{R}^3 , 则 Q 为正交

$\square)$ ~~有~~ $Q = (Q \leftarrow T) \leftarrow P \rightleftharpoons Q = T^{-1} \leftarrow P \leftarrow P(v_i)$

但 $-$ change bases matrix 不 orthogonal

应用 2. $\forall A_{mn} \in \mathbb{R}$, $P \in \mathbb{R}^n$ basis $\{x_1, \dots, x_r\}_{r \leq n}$, $r \in \mathbb{R}^m$ basis $\{y_1, \dots, y_{m-r}\}_{r \leq m}$

$$\text{def} [L_A]_P^r = \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix} \quad \square A_i x_i \in R(A) \quad L_A(x_i) = 0 \quad \text{if } A = T \begin{pmatrix} E_r & 0 \\ 0 & 0 \end{pmatrix} B^{-1}$$

$\Rightarrow rk = r = mn(n, m)$ (full-rank) 由 $\{m \in \mathbb{C}^n | L_A(m) = 0\} = \{0\}$

$$\text{eg 1. } A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \quad P = \left\{ \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \right\}$$

$$r = \left\{ \begin{pmatrix} 1 \\ 4 \\ 7 \end{pmatrix}, \begin{pmatrix} 2 \\ 5 \\ 8 \end{pmatrix}, \begin{pmatrix} 3 \\ 6 \\ 9 \end{pmatrix} \right\}$$

$$[L_A]_P^r = \begin{bmatrix} I_3 & 0 \\ 0 & 0 \end{bmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 3 & 0 \\ 4 & 5 & 6 & 0 \\ 7 & 8 & 9 & 0 \end{pmatrix}$$

$$\text{eg 2. } A = \begin{pmatrix} -1 & 1 \\ 2 & -1 \end{pmatrix} \quad P = \left\{ \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\} \quad [L_A]_P^r = \begin{bmatrix} (-1)_r & (1)_r \\ (2)_r & (-1)_r \end{bmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

3. 例 2. \forall orthonormal basis $(e_i, e_j, e_j) = \delta_{ij} = \sum_{i,j} e_i e_j^T$, $\forall x = \sum a_i e_i$, i.e. x basis \mathbb{R}^n .

$\Rightarrow \delta_{ij} = \langle x, e_i \rangle$ $\square) \forall x = a_1 e_1 + \dots + a_n e_n, A \in \mathbb{R}^{n \times n}, \langle x, e_i \rangle = \langle a_1 e_1 + \dots + a_n e_n, e_i \rangle = a_i$

~~***~~ 10. Inner Map 2

10. Given Map T
 $T: V \rightarrow W$, V, W vector spaces, T linear. (i) $\text{Null}(T) \leq V$, $\text{Range}(T) \leq W$

$$\text{Null}(T) = \{x \in V : T(x) = 0_W\}$$

$$\text{Range}(T) \rightarrow T(v_j) = 0_{\mathbb{R}^n} \therefore 0_{\mathbb{R}^n} \in \text{Range}(T)$$

ii) If $T(\alpha) = x$, $T(\beta) = y$, note $T(\alpha + \beta) = x + y$. $\therefore x + y \in T(\mathbb{R})$

iii) Let $T(x) = x$

$\text{age}(T) \quad \text{not } T(c,d) \rightarrow cT(d) = \text{crossover}(T)$

T linear $T: V \rightarrow W$, $\beta = \{v_1, \dots, v_n\}$ a basis, (2) $\text{Range}(T) = \text{span}(T(\beta)) = \text{span}\{Tw_1, \dots, Tw_n\}$

\square 证②: note $T(\beta) \in \text{span}\{T(v_i)\}$. 对① $\text{span}(T(\beta)) \subseteq \text{Range}(T)$, 因 $T(\beta) \in W$, 且 $\text{Range}(T) \subseteq W$, 故 $T(\beta) \in W$.

$\text{Range}(T) \subseteq \text{span}(T(\beta))$, let $x \in V$, $T(x) = x - P_{\text{Range}(T)}(x)$, $x = \sum \beta_i v_i$, T linear, $\therefore x = \sum \beta_i T(v_i)$

$\{3\} \ni T: P_1(\mathbb{R}) \rightarrow M_{2,2}(\mathbb{R}), f(x) \mapsto \begin{bmatrix} f(1) & f(2) \\ 0 & f(0) \end{bmatrix}, T$ Range(T) basis

$\square P_2[\mathbb{R}]$ 的 basis $(\beta = \{1, x, x^2\})$, $R_{2,2}(T) = \text{span}\{T(1), T(x), T(x^2)\} \rightarrow \left\{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & 4 \\ 0 & 0 & 1 \end{bmatrix}\} \xrightarrow{\text{1st}} \dim = 3$

例 2 设 $T: P_2(\mathbb{R}) \rightarrow P_3(\mathbb{R})$, $f(x) \mapsto 2f'(x) + \int_0^x 3f(t)dt$, 是 \mathbb{R} 上的线性映射, 且 T 为单射.

对 $\{v_1, \dots, v_n\}$, 存在 $W_1, \dots, W_n \in W$, 使得存在唯一的线性映射 $T: V \rightarrow W$, 使得 $T(v_i) = w_i$

$\square \quad \forall x \in V, \quad x = \sum a_i v_i, \text{ DEF } T: V \rightarrow W, x \mapsto \sum a_i w_i.$

且 $\forall v_i = \sum_{j \neq i} c_j v_j + v_i$, $T(v_i) = w_i$. 且唯一: 若 $\exists U: W \rightarrow V$, $\forall x \in V$, $U(x) = T(x)$.

P_i 对 linear $T: V \rightarrow W$, $\dim V < \dim W$, T 不能 onto (这叫单射; T biject $\Rightarrow \dim V = \dim W$)

日 ~~由~~ $\text{rank } T \leq \dim V < \dim W$; nullity(T) = $\dim V - \text{rank } T \Rightarrow \dim V - \dim V = 0$

ii) $T^{-1} \Leftrightarrow T \text{ has L.ind subset map } \rightarrow W \text{ has L.ind subset } \Leftrightarrow \text{Null}(T) = \{0\}$

\Leftarrow 该证：假设不成立，即 $\exists x \neq y$ 且 $T(x) = T(y)$ 。对 $z = x - y \neq 0$ 有 $\nexists l \in \text{Ind}$. 但 $T(lz) = T(x) - T(y) = 0$, so l L.dep. 矛盾

iii) $T \vdash T$, 且 $S \subseteq V$. S L.ind $\Leftrightarrow T(S)$ L.ind.

(iv) T^{-1} 且 onto . 证 T 将 $\beta_{ij} \rightarrow \alpha_{ij}$ (过手可逆). 因此 $\dim V = \dim W$, $V \cong W$ (图 2). 由 $\text{bijection } T$, 而由

由 $P_{\text{ind}}(T(v_i)) \subset \text{L. Ind.}$, 且 $\text{Range}(T) = \text{Span}\{T(v_i)\}$. 而 T onto, $\therefore \text{Range}(T) = W$. $\therefore \{T(v_i)\} \text{ span. } \{T(v_i)\} = P_W$ unique.)

④ 若 $S \subseteq V$ 且 $T(S) \perp \text{Ind}_f$, 则 $T(S) \perp \text{Ind}_f$ 或者 $\exists a_i \in T(S), \forall i, a_i T(s_i) = 0, \exists i \neq j, s_i \neq s_j$, 由 (herity), $T(\sum a_i s_i) = 0$

vi) \Rightarrow choose R basis for V . Then 任选可逆矩阵 R , 使 $R^{-1}A = \text{diag}(r)$

Ex: If $\{r_j = \sum_{i=1}^n q_{ij} \beta_i\}$, $n = \{r_j\}_{j=1}^n$ is V bases, $\{Q\}$ is change basis matrix

$$\sum_j a_j r_j = 0 = \sum_j a_j \sum_i Q_{ij} \beta_i = \sum_i (\sum_j a_j Q_{ij}) \beta_i = 0 \text{ (by basis, } \forall i, \sum_j a_j Q_{ij} = 0)$$

$$(a_1 \dots a_n) \begin{pmatrix} Q_{11} & Q_{1n} \\ Q_{21} & Q_{2n} \end{pmatrix} = (a_1 \dots a_n) Q = (0 \dots 0) = 0 \quad Q \text{ is } \mathbb{R}^n \text{ basis}$$

证了. 例证不-1, $\exists x \neq 0$ s.t. $T(x) = 0$. 考虑 $L.\text{Ind}(\beta) = \{u_1, \dots, u_n\}$, $x = \sum a_i u_i$, $T(x) = 0 = \sum a_i T(u_i)$, 但 $a_i \neq 0$ 时 $T(u_i) \in L.\text{Ind}$.

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这个图. Ex 145 & 2, 图的什么名

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = a(1) + b(2)$$

1. 因为 B col vector, $B = (b_1, \dots, b_n)^T$, $AB = \sum_{i=1}^n b_i A_{:, i}$ $\square P_1$ (i&ii) \square 3. ii)
 \square : $\{e_1, \dots, e_n\}$ 为 \mathbb{F}^m 的 basis, $B = \sum_i b_i e_i$, $\therefore A(b, e_1 + \dots + e_n) = b_1 A e_1 + \dots + b_n A e_n = \sum_i b_i A_{:, i}$
 类似地, A row vector, $A = (a_1, \dots, a_n)$ $AB = \sum_{i=1}^n a_i B_{i,:}$ $\square AB = (B^T A^T)^T$ 行上行相加
 $(ab) \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = a(1) + b(2)$

2. $T: V \rightarrow W$, $\exists \text{ rank}(T) = \text{rank}(L_A)$ $\text{nullity}(T) = \text{nullity}(L_A)$

- $\square \text{rank } T = \dim(R(T)) \leq W$ $\text{rank } L_A = \dim(R(L_A)) \leq \mathbb{F}^m$ $\xrightarrow{\phi_r: V \rightarrow \mathbb{F}^m}$ $\xrightarrow{\phi_r: R(T) \rightarrow R(L_A)}$ $\xrightarrow{\phi_r: L_A \rightarrow \mathbb{F}^m}$
 由固例 1 想到 $\phi_r: W \rightarrow \mathbb{F}^m$, WTS: $\phi_r: R(T) \rightarrow R(L_A)$
 $\Leftarrow: x'' \in \phi_r R(T)$, 有 $x' \in R(T)$ st. $\phi_r(x') = x''$, 有 $x \in V$ st. $\phi_r(T(x)) = x''$ $\therefore L_A \not\subset \phi_r(x) - x'$ \square . 2
 \therefore 有 $\phi_r(x)$ st. $L_A(\phi_r(x)) = x''$ 三) 类似, $\therefore \phi_r R(T) = R(L_A)$
 \therefore 由固例 1, $\text{rank } L_A = \dim(R(L_A)) = \dim(\phi_r R(T)) = \dim(R(T)) = \text{rank } T$ nullity 由 rank-thm 得

3. $\forall A, B$ m rows $N(A^T) \subseteq N(B^T) \Rightarrow R(B) \subseteq R(A)$
 \square 素数 $[A \cup B]$ (多 1 col), 由固例 3, $\dim R(A) + \dim N(A^T) = m$ $\dim R(B) + \dim N(B^T) = m$ $\dim R(A \cup B) + \dim N((A \cup B)^T) = m$ $\therefore [A \cup B]^T u = \begin{bmatrix} Au \\ Bu \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$
 $\Rightarrow \dim R(A) = \dim R(A \cup B)$ $\therefore V \not\subset R(A)$ 且 $V \subset R(B)$ $\therefore V \subset R(A)$

4. $\forall A_{m \times n}, B_{n \times p}$ $\dim R(AB) = \dim R(B) = \dim(N(A) \cap R(B))$

- \square θ map $T: R(B) \rightarrow R(AB)$ $y \mapsto Ay$ $\because T(cy_1 + cy_2) = T(c(B(x_1 + x_2))) = AB(c(x_1 + x_2))$
 $= cABx_1 + cABx_2$, linear. \therefore 由固例 3, $\dim R(B) = \text{nullity } T + \text{rk } T = \dim R(AB)$

5. $N(A) = N(A^T A)$ $\square \subseteq$ $\forall Ax = 0 \quad A^T A x = 0 \quad \therefore A^T A x = 0 = \|Ax\|^2$, 由 1). 0
 $\forall x \neq 0$

6. $V \subseteq \mathbb{R}^n$, \exists matrix UV st. $V = kN(V) = N(U)$

- \square $\exists \{v_1, \dots, v_r\}$ basis of $\dim = r$ V , $V = [v_1 \dots v_r]$, $R(V) = V$ 又令 $\{u_1, \dots, u_{n-r}\}$ basis of V^\perp
 $U = \begin{bmatrix} -u_1 & \dots & -u_{n-r} \end{bmatrix}$, $V^\perp = R(U^T) \Rightarrow V = R(U^T)^\perp = N(U)$

7. 1. L_A bijection $R(A^T) \rightarrow R(A)$; L_{AT} bijection $R(A) \rightarrow R(A^T)$

- \square 1° onto) $b \in R(A)$, $\exists R(A)$ 有 $Ax = b$, $x \in \mathbb{R}^n$, $\exists x = y + z$ $\therefore Ax = Ay + Az = b$.

- 1-1) $Ax = b$ 2° true, $\forall x \in \mathbb{R}^n$ 有 $x \neq 0 \in R(A^T)$, 即 $A^T y = x$, $A^T y = 0$ 有 $y = 0$

- 2° 由 1°, L_{AT} 为 $R(A) \rightarrow R(A^T)$ bijection 见后周, ***9.2

Linear Map 4

1. 由 2.2. 推出證明:

$$\text{若 } T: V \rightarrow W, U: W \rightarrow Z, \text{ let } A = [U]_{\beta}^{\gamma}, B = [T]_{\alpha}^{\beta}$$

$$\alpha = \{v_1, \dots, v_m\}, \beta = \{w_1, \dots, w_p\}, \gamma = \{z_1, \dots, z_n\}, \text{ consider } U \circ T: V \rightarrow Z, \text{ 为线性,}$$

$$\text{DEF } AB = [UT]_{\alpha}^{\gamma}, \text{ 若 } v_j: UT(v_j) = U(\sum_{k=1}^m B_{kj} w_k) = \sum_{k=1}^m B_{kj} U(w_k) \text{ (linearity)}$$

$$= \sum_{k=1}^m B_{kj} (\sum_{i=1}^p A_{ik} z_i) = \sum_{i=1}^p (\sum_{k=1}^m A_{ik} B_{kj}) z_i, \therefore \text{令 } C_{ij} = \sum_{k=1}^m A_{ik} B_{kj}, \text{ 为 } (AB)_{ij} = C_{ij}$$

NOTE, 因 fog - 1 ≠ gaf, 很容易的 $AB \neq BA$, $(\begin{smallmatrix} 1 & 0 \\ 0 & 0 \end{smallmatrix})(\begin{smallmatrix} 0 & 1 \\ 0 & 0 \end{smallmatrix}) = (\begin{smallmatrix} 0 & 0 \\ 0 & 0 \end{smallmatrix})$

易证 $[UT]_{\alpha}^{\gamma} = [U]_{\beta}^{\gamma} [T]_{\alpha}^{\beta}$. 且 $\boxed{[T]_{\alpha}^{\beta} [U]_{\beta}^{\gamma} = [U]_{\beta}^{\gamma} [T]_{\alpha}^{\beta}}$

2. P_1 : 正好 $V, V_2 \in L(V, W), T \in L(W, Z), T(U_1 + U_2) = T(U_1) + T(U_2)$ 等, matrix representation
let $A m \times n, B, C n \times p, D \in q \times n$

i) $A(B+C) = AB+AC$; $(D+E)A = D(A)+E(A) \quad \boxed{[A(B+C)]_{ij} = \sum_{k=1}^n A_{ik}(B+C)_{kj} = \sum_{k=1}^n A_{ik}B_{kj} + \sum_{k=1}^n A_{ik}C_{kj}}$

ii) 若 $k \in \mathbb{F}$, $k(AB) = (kA)B = A(kB)$ 从 Kronecker delta, $\delta_{ij} = \begin{cases} 1, & i=j \\ 0, & i \neq j \end{cases}$ 及 I_n 可由 δ_{ij} def.

iii) $I_m A = A = A I_n \quad \boxed{[I_m A]_{ij} = \sum_{k=1}^m I_{mk} A_{kj} = \sum_{k=1}^m \delta_{ik} A_{kj} = A_{ij} \text{ 且 } k=1 \Rightarrow \delta_{ik}=1} \quad (I_n)_{ij} = \delta_{ij}$

iv) 若 $\dim V = n$, $[I_V]_{\beta}^{\gamma} = I_n \quad \boxed{[I_V]_{\beta}^{\gamma} = [[I_V]_{\beta}^{\gamma}]_1, \dots, [I_V]_{\beta}^{\gamma}] = [[\beta]_{\beta}^{\gamma}, \dots] = [e_1, \dots, e_n] = I_n}$

NOTE Cancel law 不成立, eg $A = (\begin{smallmatrix} 0 & 0 \\ 0 & 0 \end{smallmatrix}), A^2 = 0 = (\begin{smallmatrix} 0 & 0 \\ 0 & 0 \end{smallmatrix})$, $\neq A \cdot A = \emptyset = A \cdot 0$, $A = 0$, 且

3. P_3 : 由定理 let $A m \times n, B n \times p$, i) $AB_{:,j}^{(1)} = A \cdot B_{:,j}$ ii) $B_{:,j} = B e_j$
 $\boxed{i) AB_{:,j} = \begin{pmatrix} AB_{1,j} \\ \vdots \\ AB_{n,j} \end{pmatrix} = \begin{pmatrix} \sum_{k=1}^n A_{ik} B_{kj} \\ \vdots \\ \sum_{k=1}^n A_{nk} B_{kj} \end{pmatrix} = A \begin{pmatrix} B_{1,j} \\ \vdots \\ B_{n,j} \end{pmatrix} = A B_{:,j} \quad ii) (B e_j)_{:,i} = \sum_{k=1}^p B_{i,k} e_k = B_{:,j}}$
虽然 $AB_{:,j} = A_{:,j} \cdot B$, 但用的不是 ()
 $\boxed{OR: \forall k \in \mathbb{F} (B I)_{:,j} = B I_{:,j} = B e_j}$
 $\boxed{ii) (AB)_{i,:} = (AB)_{i,1}, \dots, (AB)_{i,p} = (\sum_{k=1}^n A_{ik} B_{kj}, \dots, \sum_{k=1}^n A_{ik} B_{kp}) = [A_{i,1}, \dots, A_{i,n}] B = A_{i,:} B}$

4. 由 2. 线性方程组-证明: $[T(u)]_{\gamma} = [T]_{\beta}^{\gamma} [U]_{\beta}^{\gamma}$

$\boxed{\text{fix } u \in V, \text{ def } f: F \rightarrow V, a \mapsto au, \text{ let } f = \{1\} \text{ basis for } F. \text{ note } g = Tf}$

$\boxed{[T(u)]_{\gamma} = [g]_{\gamma} = [g]_{\gamma}^r = [Tf]_{\beta}^{\gamma} = [T]_{\beta}^{\gamma} [f]_{\beta}^{\gamma} = [T]_{\beta}^{\gamma} \cdot [f(u)]_{\beta} = [T]_{\beta}^{\gamma} [u]_{\beta}}$

5. DEF $L_A: \mathbb{F}^n \rightarrow \mathbb{F}^m$, $x \mapsto Ax$, $A m \times n$, L_A linear (\oplus P1, P2 & iii)) ≠ 3. ii)

P_1 i) $[L_A]_{\beta}^{\gamma} = A \quad \boxed{[[L_A(e_1)]_{\gamma}, \dots, [L_A(e_n)]_{\gamma}] = [[Ae_1]_{\gamma}, \dots, [Ae_n]_{\gamma}] = [A_{:,1}, \dots, A_{:,n}] = A}$

ii) $L_A = L_B \Leftrightarrow A = B \quad \boxed{A = [L_A]_{\beta}^{\gamma} = [L_B]_{\beta}^{\gamma} = B, \therefore A = B}$

iii) $L_A + B = L_A + L_B \quad L_k A = k L_A \quad \boxed{P_1 \text{ i) & ii)}}$

iv) (回 P1) $T: \mathbb{F}^n \rightarrow \mathbb{F}^m$ linear, \exists unique $m \times n$ C s.t. $T = L_C$, $\boxed{C = [T]_{\beta}^{\gamma}, [T(x)]_{\gamma} = [T]_{\beta}^{\gamma} [x]_{\beta}}$

v) Let $E n \times p$, $L_{AE} = L_A L_E$ (这证明 E 是矩阵且满足 $L_{A(BC)} = L_A L_{BC} = (L_A L_B) L_C$) 得力

$\boxed{L_{AE}(e_j) = AE(e_j) = A \cdot E(e_j) = A \cdot (E \cdot e_j) = L_A L_E(e_j) \therefore \text{两方程同} \quad \boxed{P_1 \text{ 适合)}}$

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例 1.1. $T: V \rightarrow W$ isomorphism, $V_0 \leq V$, i) $T(V_0) \leq W$, ii) $\dim V_0 = \dim T(V_0)$

\square i) $x, y \in T(V_0)$ $x+y \in T(V_0)$ $x+y = T(x+y) \in T(V_0)$ $\exists k \in \mathbb{R}, kx = T(kx) \in T(V_0)$ $\forall x \in T(V_0), \exists k \in \mathbb{R}, kx \in T(V_0)$ $x' = \sum a_i T(\beta_{v,i}) \text{ span } T(V_0)$

例 2. V basis $\beta = \{v_1, \dots, v_n\}$ W basis $\gamma = \{w_1, \dots, w_m\}$, $\exists! T_{ij}: V \rightarrow W, v_k \mapsto \sum_{j=1}^m a_{kj} w_j$

设 $\{T_{ij}: i=1 \dots m, j=1 \dots n\}$ basis for $L(V, W)$, $M^{ij} \in M_{m \times n}$, 令 $\alpha_{ik} = a_{ki}$, 则

$M^{ij} = [T_{ij}]_{\beta}^{\gamma}$, 且 $\exists! \Psi: L(V, W) \rightarrow M_{m \times n}(\mathbb{R}), T_{ij} \mapsto M^{ij}$, Ψ is isomorphism

\square i) $\sum \sum a_{ij} T_{ij} = 0$, $L(V, W) \neq 0 \Rightarrow f(x) = 0, \forall x \in V \Rightarrow \sum_j a_{ij} T_{ij}(v_k) = 0 \Rightarrow \sum_i a_{ik} T_{ik}(v_k) = 0$

$\forall k=1 \dots n, \sum_i a_{ik} w_i = 0 \Rightarrow a_{ik} = 0$ 且 k arbitrary $\therefore L$ ind.

ii) $[T_{ij}]_{\beta}^{\gamma} = [[T_{ij}(v_k)]_{k=1}^n]_{j=1}^m = M^{ij}$ (只有 $k=j$ 才有 w_j , 只有 $i=n$ 有 $[w_i]_j = e^i$)

iii) note $\Psi(T_{ij}) = M^{ij} \in M_{m \times n}(\mathbb{R})$ 的 basis, $\therefore \Psi$ is isomorphism

例 3. $S: P_3(\mathbb{R}) \rightarrow M_2(\mathbb{R})$, $f \mapsto \begin{pmatrix} f(1) & f(2) \\ f(3) & f(4) \end{pmatrix}$ is S invertible?

\square 法 1. $S(ax^3 + bx^2 + cx + d) = \begin{pmatrix} f(1) & f(2) \\ f(3) & f(4) \end{pmatrix} = 0 \Rightarrow a=0, b=0, c=0, d=0$, $\text{Nul}(S) = \{0\}$, $L-1$, $\boxed{\dim P_3(\mathbb{R}) = \dim M_2(\mathbb{R})}$, $\therefore S$ invertible

法 2. $[S]_{\beta}^{\gamma} = \dots$ 且 $[S]_{\beta}^{\gamma}$ 为矩阵可逆

法 3. 直接找出 $S^{-1}: \begin{pmatrix} a & b \\ c & d \end{pmatrix} \rightarrow P$ $\begin{cases} p(1) = a \\ p(2) = b \\ p(3) = c \\ p(4) = d \end{cases} \xrightarrow{\text{Lagrange}} p(x) = \sum_{i=0}^3 p(i+1)f_i, \text{ 且 } S(S^{-1}(x)) = x$

例 4. $A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$ let $\beta = \left[\begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix} \right] r = \left[\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \right]$ 找出 $[L_A]_{\beta}^{\beta} = P^{-1}AQ$ 且 PQ

\square 法 1. $[L_A]_{\beta}^{\beta} = [I][L_A][I]_{\beta}^{\alpha} = ([I]_{\beta}^{\alpha})^{-1} A ([I]_{\beta}^{\alpha}) = \begin{pmatrix} 1 & 3 \\ 2 & 4 \end{pmatrix}^{-1} A \begin{pmatrix} 5 & 7 \\ 6 & 8 \end{pmatrix}$

法 2. $[L_A]_{\beta}^{\beta} = [I]_{\beta}^{\beta} [L_A]_{\beta}^{\alpha} = \left[\begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix} \right]_{\beta}^{\alpha} Q^{-1} A Q \xrightarrow{\text{PQ}} PQ^{-1} A Q = Q([I]_{\beta}^{\beta})^{-1} = Q[I]_{\beta}^{\alpha}$

In fact, T invertible $\Leftrightarrow [T]_P^r$ invertible

(Linear Map 5)

$\square \Rightarrow$ let B $n \times n$, s.t. $B[T]_P^r = I_n$, 由 $\text{Ex. 2. 基本直和 } L(V, W) \cong M_{m \times n}(\mathbb{F})$, 由 Ex. 10. T_3 , $\exists! U \in L(W, V)$, s.t. $U: W \xrightarrow{\sim} \bigcup_{i=1}^n B_{ij} V_i$, $\therefore B = [U]_P^R$, 且 $U = T^{-1}$: $[UT]_P = [U]_P^R [T]_P^r = BA = I_n = [I_V]_P$, $\therefore UT = I_V$ (由 P 为 I)

(note $i=j$, dim $V_i = 1$)

2. 对 bijection T : $V \rightarrow W$, 考虑 $T^{-1} W \rightarrow V$ 为 linear, 由此 T 为 invertible

根据 T invertible, T^{-1} 也 invertible, $[T^{-1}]_Y^P = ([T]_P^r)^{-1}$ (Priority map, 由 $A \in \text{Ex. 10. P}_{\text{inv}}$ 由 I notation)

\square 因 T bijection, $\dim V = \dim W$, $[T]_P^r$ $n \times n$, 而 $T^{-1} T = I_W$, $\therefore I_n = [I_V]_P^R = [T^{-1} T]_P = [T^{-1}]_Y^P [T]_P^r$ 同理 $I_n = [I_W]_P^R = [T T^{-1}]_P = [T]_P^r [T^{-1}]_Y^P$

2. 事实上, 只有 $\dim V = \dim W \Leftrightarrow V \cong W$ (pp 3 bijection linear $T: V \rightarrow W$)

$\square \Rightarrow \sum p = \{v_1, \dots, v_n\}$ 由 Ex. 10. T_3 存 unique $T: V \rightarrow W$, $v_i \mapsto w_i$, !. $T^{-1} T = I_V$, 而 $\text{Rng}(T) = \text{span}\{T(v_1), \dots, T(v_n)\} = W$ \therefore onto \Rightarrow bijective

$\Leftarrow \text{Ex. 10. P}_i$

NOTE, 由此 $\dim V = n$, $V \cong \mathbb{F}^n$, 由 4-特征看 Ex. 2. 2

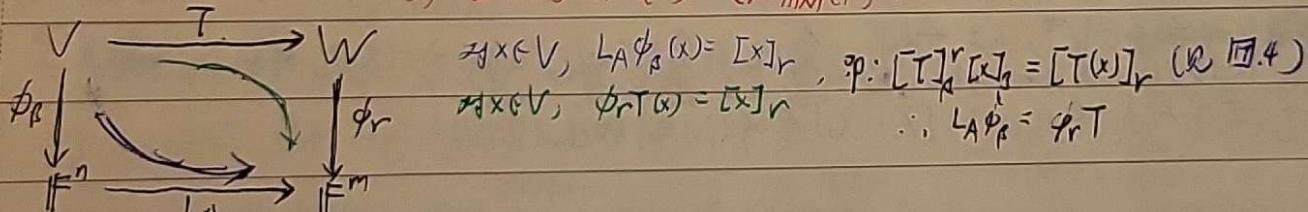
Ex. 2. 2

1. 由 $L(V, W) \rightarrow M_{\dim(W) \times \dim(V)}(\mathbb{F})$, $T \mapsto [T]_P^r$ 2. $\phi_P: V \rightarrow \mathbb{F}^{\dim(V)}$, $x \mapsto [x]_P$

两者皆 isomorphism

由 section: $\forall A \in M$, $\exists! T$ s.t. $\Phi(T) = [T]_P^r = A$, 由 $[T]_P^r$ 的唯一性

(由 $\text{Ex. 10. dim}(L(V, W)) = \dim(W) \times \dim(V)$ ($M_{m \times n}(\mathbb{F})$ dim mn))



NOTE $\text{Ex. 2. 2} \cong V$, 为 \mathbb{F}^n -isomorphism. 考 Ex. 4. 2 basis change

由 Ex. 4. 2 为 \mathbb{F}^n -isomorphism ($M_n(\mathbb{F})$)

3. 由 Ex. 5. T_2 , $\exists T: L_A: \mathbb{F}^n \rightarrow \mathbb{F}^m$, $[L_A]_P = Q^{-1} [L_A]_P Q = Q^{-1} A Q \xrightarrow{\text{由 Ex. 4. 2}} J \text{ col } \mathbb{F}^m$

e.g. $A = \begin{pmatrix} 2 & 1 & 0 \\ 0 & 1 & 3 \end{pmatrix} r = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \left(\begin{pmatrix} 2 & 1 & 0 \\ 0 & 1 & 3 \end{pmatrix} \right)$, $Q = \begin{pmatrix} -1 & 2 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{pmatrix}$, $\therefore [L_A]_P = Q^{-1} A Q = \begin{pmatrix} 0 & 2 & 1 \\ 0 & 1 & 3 \\ 0 & 0 & 1 \end{pmatrix}$

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(β, γ standard.)

Ex 1 1° let $\beta = \{(1), (2), (3)\}$ basis for \mathbb{R}^2 , β^* , 2° $T: P_1(\mathbb{R}) \rightarrow \mathbb{R}^2$, $p(x) \mapsto (p(0), p(2))$

$$\square 1^\circ \beta^* = \{f_1, f_2\} \text{ s.t. } f_1(1) = 1 = f_1(2 + (1)) = 2f_1(1) + f_1(0) \stackrel{T}{\mapsto} f^* \\ f_1(3) = 0 = 3f_1(0) + f_1(2) \quad \therefore f_1(e_1) = 1, f_1(e_2) = 3 \quad f_1(x+y) = x+3y$$

$$2^\circ [gT]_{\beta^*}, [gT]_{\beta^*} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$\text{这样 } g, T = af_1 + bf_2, g, T(1) = g, (1) = g, (e_1) + g, (e_2) = 1 \stackrel{=0}{=} ; g, T(0) = af_1(0) + bf_2(0) = a \quad \therefore a = 1 \\ g, T(x) = g, (2) = 2g, (e_2) = 0; g, T(2) = af_1(2) + bf_2(2) = b \quad \therefore b = 0 \quad \therefore \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

Ex 2 $V = \mathbb{R}^3$, $f_1\left(\begin{pmatrix} x \\ y \\ z \end{pmatrix}\right) = x-2y$, $f_2\left(\begin{pmatrix} x \\ y \\ z \end{pmatrix}\right) = x+y+z$, $f_3\left(\begin{pmatrix} x \\ y \\ z \end{pmatrix}\right) = y-3z$, 试 $\{f_1, f_2, f_3\}$ basis of V^*

$$\square i) \sum a_i f_i\left(\begin{pmatrix} x \\ y \\ z \end{pmatrix}\right) = 0 \stackrel{\text{zero func}}{\Rightarrow} \sum a_i \left(f_1\left(\begin{pmatrix} x \\ y \\ z \end{pmatrix}\right)\right) = 0 \stackrel{\text{if } h \neq 0}{\Rightarrow} \sum a_i x - 2 \sum a_i y = 0 \quad \therefore \begin{cases} a+b=0 \\ 2a+b=0 \end{cases} \Rightarrow a=0, b=0 \quad \text{且找出 } V^* \text{ 的 basis}$$

$$ii) f_1(v_1) = 1 \\ f_2(v_1) = 0 \quad f_3(v_1) = 0 \Rightarrow \begin{cases} x_1 - 2y_1 = 1 \\ x_1 + y_1 + z_1 = 0 \\ y_1 - 3z_1 = 0 \end{cases} \quad \therefore v_1 = \left(\frac{1}{5}, \frac{-1}{5}, \frac{1}{5}\right) \quad \text{同理 } v_2 = \left(\frac{3}{5}, \frac{1}{5}, \frac{1}{5}\right) \quad v_3 = \left(\frac{1}{5}, \frac{1}{5}, \frac{-3}{5}\right)$$

Ex 3. \forall basis in V^* \Rightarrow V -的 basis has dual basis, 即对 $x = \sum a_i x_i$, $f_j(\sum a_i x_i) = a_j$

\square let $\{f_1, \dots, f_n\}$ basis for V^* . 将 V^* 视为右向量的 PEF, 焉乎可由 V^* basis 得出的 V^* basis $\{x_1, \dots, x_n\}$

由 def 留出 对 $\forall t \in V^*$, $t = \sum a_i f_i$, $f_j(t) = f_j(\sum a_i f_i) = a_j = t(x_j)$, 特别的, 对 f_j ,

$\{x_i\}$ 是 V basis, 因 \cong isomorphism, 有 $t \mapsto x_i \mapsto f_j$, 且 $\{f_j\}$ 也是 V^* basis

$$f_j(x) = f_j(\sum a_i x_i) = a_j \quad \therefore \{f_1, \dots, f_n\}$$
 是 $\{x_1, \dots, x_n\}$ 的 dual basis

linear functional | Linear Map 6 |
 SQL
 (aggregate func 也是 linear functional)

$V \xrightarrow{T} W \xrightarrow{f} F$

Date.

1. $L: V \rightarrow F$, $\exists V^* = L(V, F)$ dual space $\xrightarrow{\text{def}} \dim V^* = \dim L(V, F) = \dim F$, $\dim V^* = \dim V$, $\therefore V^* \cong V$

DEF: $\forall x \in V, x = \sum a_i v_i, f_i(x) = a_i$ word function

T_{1a} $\Leftrightarrow \beta = \{f_1, \dots, f_m\}$, β^* basis for V^* $\xrightarrow{\text{def}} \forall f \in V^*, f \in \{f_1, \dots, f_m\}$ to kronecker delta
 $\square \exists g = \sum f_i w_i \in \text{span } \beta$, $w_i \in g = f_i \in \text{span } \beta$ $\therefore \text{span } g = \text{span } \beta$

T_{1b} $\forall T: V \rightarrow W$, DEF $T^*: W^* \rightarrow V^*$, $g \mapsto gT$, $[T^*]_{r^*}^{p^*} = [T]_p^r$ 见上图

\square i) $\forall g \in W^*, T^*(g) = gT \in L(V, F)$, $\xrightarrow{\text{def}} T^* \text{ map to } V^*$ $\therefore A = [T]^r_p$
 ii) $T(f+kg) = fT + kgT$, linearity iii) $\beta^* = \{f_1, \dots, f_m\}$ $r^* = \{g_1, \dots, g_m\}$ 考虑 $[T^*]_{r^*}^{p^*}$
 $\xrightarrow{\text{def}} [T^*(g_j)]_{p^*} \xrightarrow{\text{def}} \text{i-row, 由 } T_{1a}, T^*(g_j) = \sum_i g_i T(v_i) f_i$
 $\therefore \text{i-row} = g_j T(v_i) = g_j (\sum_k A_{ki} w_k) = \sum_k A_{ki} g_j(w_k) = \sum_k A_{ki} \delta_{jk} = A_{ji}$

2. 考虑 V^* , Def $\hat{x}: V^* \rightarrow F, f \mapsto f(x)$ 易证 \hat{x} L.T., $\exists \psi: V \rightarrow V^*$, $x \mapsto \hat{x}$ isomorphism

$\square \cup L.T. \text{ 考虑 } \hat{x} + ky \mapsto \hat{x} + kf(y) \Rightarrow \hat{x} + f(-V^*)$, $\hat{x}(f) = f(\hat{x}) = f(x) + kf(y) \quad (V^* \text{ 为 set of L.T.})$

ii) 1-1. 若 $\psi(x) = 0$, $\xrightarrow{\text{def}} \hat{x} = 0$, $\xrightarrow{\text{def}} \forall f \in V^*, \hat{x}(f) = 0$ $\xrightarrow{\text{def}} f(x) = 0$. 而 $x \in V, x = \sum a_i v_i, \sum a_i f(v_i) = 0$
 因子任意, 对各 $f_j, \sum a_i f_j(v_i) = 0, \sum a_i \delta_{ij} = 0, \therefore a_j = 0 \quad \therefore x = 0$ (由 T_{1a})

iii) onto: $\therefore \dim V^* = \dim L(V, F) = \dim V \cdot \dim F = 2 \cdot \dim V \quad \therefore \text{onto}$

3. proper

T_{2a} 令 $W \subsetneq V$, \exists nonzero linear functional $f \in V^*$ st. $f(x) = 0, \forall x \in W$

\square 对 W basis $\{v_1, \dots, v_m\}$ 伸至 V basis $\{v_1, \dots, v_n\}, W \subsetneq V \therefore m < n$. 取 $f_{m+1} \in V^*$ basis, $\forall x \in W, f_{m+1}(\sum_i a_i v_i) = 0$

而 $f_{m+1}(V) = 1 \quad \therefore f_{m+1}$ nonzero

T_{2b} 令 $T: V \rightarrow W$, i) T onto $\Leftrightarrow T^*$ 1-1
 ii) T 1-1 $\Leftrightarrow T^*$ onto

\square i) \Rightarrow let $g \in \text{Nul}(T^*)$, $g \in W^*$, $\forall w \in W$, consider $g(w) = g(T(w)) = T^*(g)(w) = 0$
 $\therefore g$ zero func. $\therefore \text{Nul}(T^*) = \{0\}$ \Leftrightarrow 1-1. (由 T_{1a}) assume $R(T) \neq W$, 而 $R(T) \subset W$, 由 T_{2a} ,
 有 nonzero $g \in W^*$, 且 $g(w) = 0, \forall w \in R(T)$. 但 $\forall x \in V, T^*(g)(x) = g(T(x)) = 0$ 于质

ii) \Rightarrow 由 $R(T) \subseteq W$, 且 $\{T(v_i)\}_{i=1}^n$ 为 $R(T)$ basis, $\forall f \in V^*$, Def $g \in W^*$ as: $g(T(v_i)) = f(v_i)$
 note 由 $T^*(g)(v_i) = gT(v_i) = g(f(v_i))$ $\therefore T^*(g) = f$ (由 T_{1b})

\Leftarrow assume 1-1, $\exists x \neq 0 \in V$, st. $T(x) = 0$, def $f \in V^*$ as $f(x) \neq 0$, 而 T onto \therefore 有 $g \in W^*$ st. $gT = f$

consider $gT(x) = g(0) = 0$ 但 $f(x) \neq 0$
 $(g \in L(W, F), g \text{ L.T.}, \text{send } 0 \rightarrow 0)$

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例 1. product of 2 matrix, rank \leq min of both rank \square If $AB=0$, $A, B \neq 0$... X

例 2. If $A = \begin{pmatrix} 1 & 2 \\ 1 & 0 \\ 1 & 1 \end{pmatrix}$ \Rightarrow elementary los \neq 0

\square A 可逆 由 P_{11} $A = \begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 0 \\ 1 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 0 \\ 0 & -1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 0 & -1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = I_3 = E_0 \cdot E_1 \cdot A$

$$E_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, E_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix}, E_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix}, E_4 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, E_5 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, E_6 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} A = E_1^{-1} \cdots E_6^{-1}$$

例 3. $\forall T, U: V \rightarrow W$, a) $R(T+U) \subseteq R(T) + R(U)$ b) $\text{rank}(T+U) \leq \text{rank } T + \text{rank } U$

\square a) $y = T(x) + U(x)$ $\xrightarrow{\text{fusing } x}$ b) $\text{rank } T+U \leq \dim(R(T)+R(U)) = \dim \overbrace{R(T)}^{P_1, T_{20}} + \dim \overbrace{R(U)}^{P_2, T_{20}} - \dim(R(T) \cap R(U)) \leq \text{rank } T + \text{rank } U$

NOTE 這個定理 $\text{rank}(A+B) = \text{rank}(L_{A+B}) = \text{rank}(L_A + L_B) \stackrel{\text{由 } P_2, T_{20}}{\leq} \text{rank } A + \text{rank } B$

例 4. $A_{m \times n}, B_{n \times p}$, $\text{rank } A = m$, $\text{rank } B = n$, $\# \text{rank}(AB)$

\square $L_A(\mathbb{F}^n) = \mathbb{F}^m$, $L_B(\mathbb{F}^p) = \mathbb{F}^n$, $\# L_{AB}(\mathbb{F}^p) = L_A L_B(\mathbb{F}^p) = L_A(\mathbb{F}^n) = \mathbb{F}^m$, $\therefore \text{rank } = m$

例 5. $\exists A_{m \times n}$ rank m , $\exists B_{n \times m}$ st. $AB = I_m$; $\exists B_{n \times m}$ rank m , $\exists A_{m \times n}$ st. $AB = I_m$

\square i) $RPR(L_A) = \mathbb{F}^m$, onto. $\# \mathbb{F}^m$ basis $\{e_1 \dots e_m\}$, $\forall e_i$, $\exists L_A(e_i) = e_i$, $\therefore B = (v_1 \dots v_m)$

$$AB = (Av_1 \dots Av_m) = (L_A(v_1) \dots L_A(v_m)) = (e_1 \dots e_m) = I_m$$

ii) $\# \text{rank}(B^T) = m$, $\therefore B^T$ $m \times n$, rank $= m$, by i), $\exists C$ st. $B^T C = I_m = (C^T B)^T$, $\therefore C^T = A$

例 6. i) $A_{3 \times 1}, B_{1 \times 3}$, $\# AB$ at most rank 1

ii) $C_{3 \times 3}$ rank 1 $\Rightarrow A_{3 \times 1}, B_{1 \times 3}$, st. $C = AB$

\square i) $\begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} \cdot (b_1, b_2, b_3) = \begin{pmatrix} a_1 b_1 & a_1 b_2 & a_1 b_3 \\ a_2 b_1 & a_2 b_2 & a_2 b_3 \\ a_3 b_1 & a_3 b_2 & a_3 b_3 \end{pmatrix}$, note col 同行关系 (eg. $2 \times 1 \mapsto \frac{a_2}{a_1}$) $\therefore \text{rank } \leq 1$

ii) $\# C_{1 \times m}$ Lhd. $\# C = \begin{pmatrix} -c_1 \\ -a_{11} c_1 \\ -b_{11} c_1 \end{pmatrix}$, $\therefore A = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, B = c_1$

例 7. $\# A = \begin{pmatrix} 1 & 0 & 1 & 2 & 6 \\ -1 & 1 & 3 & -1 & 0 \\ 0 & 1 & 4 & -1 & 3 \\ 0 & -1 & 5 & 1 & -6 \end{pmatrix}$, i) $M_{5 \times 5}$ rank 2, $AM = 0$

ii) $B_{5 \times 3}$, $AB = 0$, $\# B$ rank ≤ 2

\square i) $\# M = (m_1, m_2, 0, 0, 0)$, $AM = 0$ BP $M_{11} = 0 \in \mathbb{F}^4$, $A_m M_{21} = 0 \in \mathbb{F}^4$, 快方解法 $\# (A|0) \neq \# Ax = 0$

null space $\left\{ \begin{pmatrix} x_3 + 3x_5 \\ -2x_3 + x_5 \\ x_3 \\ -2x_5 \\ x_5 \end{pmatrix} \right\}$, basis $\left\{ \begin{pmatrix} 1 \\ -2 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 3 \\ 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} \right\}$, $\therefore \# M = \begin{pmatrix} 1 & 3 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 \end{pmatrix}$

ii) $\# B = (b_1, b_2, b_3, b_4, b_5)$ BP b_i , $\# Ax = 0$ 的解空间, $\# N(L_A) = \dim(\mathbb{F}^5) = \frac{3}{\# \text{nullity}(L_A)}$

定理 1. $\# A^T A = \# AA^T = \# A = \# A^T$ (Gram matrix theorem)

\square 由 例 5 & 例 3

研究 \mathbb{F} matrix representation 与 L.T. 有怎样的关系, 即 $\text{rank}(A) = \text{rank}(LA)$

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1. T_{1a} A $m \times n$, 有 P 为 $n \times n$ 可逆, i) $\text{rank}(AQ) = \text{rank } A$ ii) $\text{rank}(PA) = \text{rank } A$ 且 $A = [L_A \quad I_n]$ 且 L_A 可逆
 \square 由 $\text{Range}(LAQ) = \text{Range}(LAQ) = L_A \text{Range}(P^T) = L_A(L_Q(P^T)) = L_A(P^{-1}) = R(L_A)$ (四.1)

$$\therefore \text{rank}(AQ) = \dim(\text{Range}(LAQ)) = \dim(\text{Range}(L_A)) = \text{rank } A \quad \text{i)} \quad \text{ii)} \quad R(L_A) = L_A \text{Range}(P^T) = L_A(P^{-1}) = R(L_A)$$

\star 这即说 elementary row/col operation 不改变 rank (四.2, E.J.) (L_p isomorphism, 四.1, $T \neq L_p$)

2. T_{1b} rank of matrix \leq max # of L.hol. col (四.3 的推导)

$$\square \text{rank } A = \frac{\# \text{det}}{\# \text{det}} = \dim(R(L_A)) = \dim(L_A(P)) = \dim(\text{span}\{A(e_i)\}_{i=1}^n) = \dim(\text{span}\{A_{\cdot j}\}_{j=1}^r)$$

3. T_{1c} let A $m \times n$, A rank = $r \Rightarrow \exists$ 可逆 B, C s.t. $(I_r \quad 0) \sim \begin{pmatrix} B & \\ 0 & C \end{pmatrix} = D = BAC$

\square 由初等变换且 $A \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ 且 $D = E_p \cdots E_1 A F_1 \cdots F_q$, 而 E_i, F_j 可逆 且有 $r+1$ (operation 为 r , & T_{1b})

$$\square \text{由 } T_{1c}, \text{ 有 } D = BAC, \therefore D^T = (BAC)^T = C^T A^T B^T, \text{ 由 } C^T, B^T \text{ 可逆, 由 } T_{1a} \text{ iii) } \text{rank}(C^T A^T B^T) = \text{rank } A^T = \text{rank } DT = r$$

i) (T_{1b} 拓展) rank of matrix \leq max # of L.hol. row

$$\square \text{max # of L.hol. row} \leq \text{max # of L.hol. col of } A^T, \text{ 而 } \text{rank } A^T = \text{rank } A$$

IV.2e T_{1b} 未提及说明 $= \text{row/col generate subspace dim} \neq \text{rank}$ (col space)

iii) 可逆矩阵的 elementary matrix 衍生

$$\square A \text{ } m \times n, \text{ if } \text{rank } A = n \therefore \text{rank}(L_A) = n = \text{rank } A (\# \text{det})$$

$$\square \text{由 } T_{1c}, \text{ 且 } D = I_n = BAC = E_p \cdots E_1 A F_1 \cdots F_q \therefore A = E_1^{-1} \cdots E_p F_q^{-1} \cdots F_1^{-1}$$

$$\square \text{由 } T_{1a}, \text{ 有 } \text{rank } A = \text{rank } kA, k \text{ non-zero scalar.} \quad \square \text{WTS } \text{Range}(LA) = R(L_A), \text{ 而 } R(L_{kA}) = R(L_A(P^T)) = kL_A(P^T) = L_A(P^T) = R(L_A)$$

IV.2f T_{1b} 未提及说明 $= \text{rank } A \leq \text{rank } (UT) \leq \text{rank } U$

$$\square \text{a) } R(UT) = UT(V) = U(R(T)) \subseteq V(W) = R(U), \text{ 由 } \text{rank } UT \leq \text{rank } U$$

$$\square \text{c) } \text{rank}(LA_B) = \text{rank}(L_A \circ L_B) \leq \text{rank}(L_A) \quad \text{d) } \text{rank } AB = \text{rank } AB^T = \text{rank } B^T A^T \leq \text{rank } B^T = \text{rank } B$$

$$\square \text{b) } \text{令 } A = [U]_B^r, B = [T]_A^s, \text{ 则 } AB = [UT]_A^r \quad \square \text{d) } \text{rank } AB = \text{rank } UT \leq \text{rank } B = \text{rank } T = \text{rank } A$$

$$\square \text{PP } \text{rank}(AB) \leq \min\{\text{rank}(A), \text{rank}(B)\}$$

T (Fröbenius) $\text{rk } ABC + \text{rk } BC \leq \text{rk } B + \text{rk } ABC$

$$\square \text{rk } ABC = \text{rk } B - \dim(R(B) \cap N(A)) \text{ 且令 } B \in BC, \text{ 则 } \text{rk } ABC = \text{rk } BC - \dim(R(BC) \cap N(A))$$

$$\therefore \text{rk } ABC + \text{rk } BC = \text{rk } B + \text{rk } ABC + \dim(R(BC) \cap N(A)) - \dim(R(B) \cap N(A))$$

$$\text{Mif } B \in I_n \text{ (Sylvester)} \quad \text{rk } ABC = \text{rank } A + \text{rank } B - n$$

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即有无 eigenvalue L.T.

例 1 a) T : state \mathbb{R}^2 by $\pi/2$ b) 令 $C^\infty(\mathbb{R}) = \{\text{set of func: } \mathbb{R} \rightarrow \mathbb{R} \mid \text{有任意阶导数}\} \subseteq F(\mathbb{R}, \mathbb{R})$

$T: C^\infty(\mathbb{R}) \rightarrow C^\infty(\mathbb{R}), f \mapsto f'$ 抽出. orthogonal

天此β

□ a) 几何高维, 无 $\Gamma(v)$ 为 v 的伸展(k_v), ∴ 无 α -vector, not oligonized

b) 若 f eigenvect. 有 $T(f) = f' = \lambda f$ 1-order DE: $f(t) = ce^{\lambda t}$. $\therefore f$ 在 $t=0$ 時為 eigenvect. λ 為 A real ($\lambda=0$, f 为 zero func.)

Frqz: V V.a. span. K scalar. KI_V diagonalizable?

note \neq bases β , $[kI_V]_A = kI$, 对称矩阵, \therefore 无论向量都是. (由 $f(\lambda)$, $\lambda - \text{eigenval} = k$)

$\exists A \sim kI$, i.e. $A = kI$; \Rightarrow diagonalizable & eigenvalues are scalar matrix

$$\boxed{A = P^{-1}kI_P = kP^{-1}I_P = kI} ; \text{ Given } D = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} \text{ has } -\text{eigenval, } \lambda_1 = \dots = \lambda_n = \lambda$$

由上知 $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ 不是 diagonalizable $\square f(\lambda) = (\lambda-1)(\lambda-1)$, - eigenvalue, 是可 diagonalizable,
scalar matrix, False

例 3. 求无 basis 在 \mathbb{R}^2 st. $[T]_B = A = \begin{pmatrix} 1 & 2 \\ 2 & 6 \end{pmatrix}$, $[T]_F = B = \begin{pmatrix} 2 & 1 \\ 1 & 3 \end{pmatrix}$

$$\square \quad \text{令 } B = Q^{-1} A Q, \quad \det(C - \lambda I) = \det(Q^{-1} A Q - \lambda Q^{-1} I Q) = \det(Q^{-1}(A - \lambda I)Q)$$

$\Rightarrow \det Q^{-1} \det((A - \lambda I)) \det Q = \det A \neq 0$, 即 characteristic poly $\neq 0$, 且 $f_A(t) = t^3 - 7t + 2 \neq f_B(t) = t^3 - 5t + 5$

例4. 若 $T: M_2(\mathbb{R}) \rightarrow M_2(\mathbb{R})$, $A \mapsto A^T$. 则 T 是线性映射.

$\square \quad \text{令 } T(A) = \lambda A, \quad A^T = \lambda A, \quad \text{而 } T(A^T) = T(\lambda A) = A = \lambda^2 A \therefore \lambda = \pm 1.$

例 5 若 real symmetric 矩阵的特征值是 real

$$\boxed{\text{证 } \forall A x = \lambda X, \quad \langle Ax, x \rangle = \langle \lambda x, x \rangle = \lambda \langle x, x \rangle} \quad \text{对 } x=0 \text{ 时 } \neq 0, \text{ 故 } \forall x \neq 0$$

Fig 6, real symmetric \mathbf{A} distinct eigenvalues, \mathbf{A} mutually orthogonal

$$\boxed{\text{If } \lambda_1 \neq \lambda_2, \quad A v_1 = \lambda_1 v_1, \quad \langle A v_1, v_2 \rangle = \langle \lambda_1 v_1, v_2 \rangle = \lambda_1 \langle v_1, v_2 \rangle}$$

$$\lambda_1 \neq \lambda_2 \Rightarrow A v_2 = \lambda_2 v_2 \quad \text{also} \quad \sqrt{\lambda_1 \lambda_2} v_1 = \langle v_1, A^T v_2 \rangle = \langle v_1, \lambda_2 v_2 \rangle$$

$$\langle v_1, v_2 \rangle = \lambda_2 \langle v_1, v_2 \rangle \quad \therefore \langle v_1, v_2 \rangle = 0$$

$$v_1^T (A^T v_2) \quad \text{symmetric} = A$$

$$\therefore \text{tr}(A) = \text{tr}(P^{-1}JP) = \text{tr}(PP^{-1}J) = \text{tr}(J) = \sum \lambda_i(J) = \sum \lambda_i(A) \quad \text{法2 由右} \sum_{i=1}^n \lambda_i + \lambda_0 = \lambda_1 + \dots + \lambda_n$$

Ex 2 skew-symmetric $A^T = -A$ has eigenvalues or purely imaginary

$$\boxed{\lambda \in \mathbb{C} \quad x^T A x \leq \lambda x^T x = \lambda \|x\|^2 \Rightarrow \lambda = -\bar{\lambda}} \\ = (A^T \bar{x})^T x = \bar{\lambda} \|x\|^2$$

diagonalization & eigenvalues

$\lambda \in A$, \exists change basis B , s.t. A diagonal, 由定理 2

若 \det

例 1. $T: V \rightarrow V$ diagonalizable iff \exists ordered basis β s.t. $[T]_\beta$ diagonal

令 $\text{矩阵 } A$ diagonalizable iff L_A diagonalizable

Note, 若 $D = [T]_\beta$ diagonal, $\forall \lambda \in \mathbb{C}$, $T(v_j) = \sum_i D_{ij} v_i = D_{jj} v_j = \lambda_j v_j$ } T diagonalizable
而若 $T(v_j) = \lambda_j v_j$, $D[T]_\beta = D = (\lambda_1, 0, \dots, 0, \lambda_n)$ $[T(v_1)]_\beta, [T(v_2)]_\beta, \dots$

故 $v \in V$ eigenvector of $T \Leftrightarrow \exists \lambda \in \mathbb{C}$, $T(v) = \lambda v$ eigenvector

$\therefore \lambda$ eigenvalue/eigenvec 矩阵, \exists ordered basis $\beta = \{v_1, \dots, v_n\}$ set of eigenvectors, 且因 β basis, 使 $\forall \lambda \in \mathbb{C}$ $(L_A - \lambda I_n)$ 有 n 个 λ 的 eigenvalue

$f(\lambda)$, characteristic polynomial

2. λ eigenvalue $\Leftrightarrow \det(A - \lambda I_n) = 0$ 因 2 note: $A + \lambda$ 全非零, $\det(A + \lambda) \neq 0$

$\square \Leftrightarrow \exists v \in V, Av = \lambda v \Leftrightarrow (A - \lambda I_n)(v) = 0 \Leftrightarrow A - \lambda I_n$ not invertible $\Leftrightarrow \det(A - \lambda I_n) = 0$

对 T , characteristic polynomial of $T \Rightarrow f(\lambda) = \det([T]_\beta - \lambda I_n)$

且 $f(\lambda)$ indep. of chosen basis

T_{1a} characteristic polynomial of degree n , leading coeff $\neq (-1)^n$ 由定理 3 2P2Z1.7 basis, 且 A matrix representation similar, $\det(A)$ 为 $f(\lambda)$ 的 WTS $f(\lambda) = (-1)^n \lambda^n + k_{n-1} \lambda^{n-1} + \dots + k_1 \lambda + k_0$. In fact, by induction $f(\lambda) = (A_{11} - \lambda) \dots (A_{nn} - \lambda) + q(\lambda)$,
 $q(\lambda)$ degree $\leq n-2$: \because 对 $n-1$ 成立, 对 n , $\begin{vmatrix} A_{11} - \lambda & & & \\ & \ddots & & \\ & & A_{nn} - \lambda & \\ & & & \ddots & A_{11} - \lambda \end{vmatrix} = (A_{nn} - \lambda) \begin{vmatrix} A_{11} - \lambda & & & \\ & \ddots & & \\ & & A_{nn-1} - \lambda & \\ & & & \ddots & A_{11} - \lambda \end{vmatrix} + \sum_{i=1}^{n-1} (-1)^{i+1} A_{ni} \det \widetilde{A}_{ni}$
 $= (A_{nn} - \lambda)(A_{11} - \lambda) \dots (A_{n-1, n-1} - \lambda) + (A_{nn} - \lambda)q(\lambda) + q'(\lambda)$ 带入 $A_{nn} - \lambda, A_{ii} - \lambda, \dots$ degree at most $n-2$
 $\therefore f(\lambda) = (A_{11} - \lambda) \dots (A_{nn} - \lambda), \lambda$ coeff $\neq (-1)^n$ NOTE, $f(0) = k_0 = \det(A - 0 \cdot I_n) = \det(A), \lambda \in \mathbb{C}$

T_{1b} $A_{n,n}$ 有 n distinct eigenvalues \square 由 T_{1a} , eigenvalue $\neq 0$ 的 zero, 是 n 不同解

T_{1c} $\forall T: V \rightarrow V$, λ eigenvalue of T , $\forall v \in V$ eigenvector $\Leftrightarrow v \neq 0 \wedge v \in \text{Null}(T - \lambda I)$
 $\square \Rightarrow v \neq 0$, 且有 $T(v) = \lambda v$. $\therefore T(v) - \lambda v = (T - \lambda I)(v) = 0 \Leftrightarrow \text{BP}(T - \lambda I)(v) = 0$, $T(v) = \lambda v$ BP E_λ (回 4)

3. T_{2a} let $\lambda_1, \dots, \lambda_k$ distinct eigenvalues of T , $\{v_1, \dots, v_n\}$ L.ind.

\square induction on k , $k=1$ T, $k \geq 2$ k, let $\sum_{i=1}^k a_i v_i = 0$, 而这 apply fine: $T - \lambda_k I : a_i [T(v_i) - \lambda_k I(v_i)] + \dots = 0$

$a_1 (\lambda_1 - \lambda_k) v_1 + \dots + a_k 0 = 0 \quad \because \{v_1, \dots, v_{k-1}\}$ L.ind. $\therefore a_1 (\lambda_1 - \lambda_k) = \dots = a_{k-1} (\lambda_1 - \lambda_{k-1}) = 0$ 且 λ distinct

2) ①: $a_k v_k = 0$ 且 v_k eigenvector, $\therefore v_k \neq 0 \therefore a_k = 0$

T_{2a}

T_{2b} T 有 n distinct eigenvalue $\Rightarrow T$ diagonalizable \square $\{v_1, \dots, v_n\}$ L.ind. \therefore 由 1. 之, 有 n 个 diff eigenvec. $\therefore T$ diagonalizable

Note T_{2b} 为 diff true. I 有 eigenval 只有 1, 但 diagonalizable 例 1. 之, 1. 例 2.

2. 1. 例 3.

4. diagonalization 为 diag 矩阵 power, 若 $A = PDP^{-1}$, $A^k = PD^kP^{-1}$, 例 1. 例 3

$D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} P = \begin{pmatrix} v_1 & v_2 \end{pmatrix}$ 由 P 为 eigen decomposition $\Leftrightarrow \{v_i\}$ 为 basis

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$\therefore \det A^k = \det P \det D \det P^{-1} = \det D = \prod \lambda_i$ 例 1. 例 2.

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diagonal & triangular

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1. diagonal matrix $\begin{pmatrix} v_1 & 0 \\ 0 & v_n \end{pmatrix} \rightsquigarrow \text{diag}(v)$, || triangular matrix $\begin{pmatrix} v_1 & & \\ 0 & \ddots & \\ & & v_n \end{pmatrix}$ 易階代換
det = $v_1 v_2 \dots v_n$
if $x \in \mathbb{F}^n$, $\text{diag}(v) \cdot x = \sum x_i v_i = v \odot x$ || strict triangular $\begin{pmatrix} v_1 & & \\ 0 & \ddots & \\ 0 & & v_n \end{pmatrix}$ main diagonal = 0
det = 0
且 $\text{diag}(v)^{-1} = \text{diag}\left(\frac{1}{v_1}, \dots, \frac{1}{v_n}\right)^T$ || tri + triangular

2. a) square A nilpotent iff $\exists m \geq 1$ s.t. $A^m = 0$ char. f(x) = (-1)^n x^n
strict triangular nilpotent: $A^0 = 0$ 由 Cayley-Hamilton f(A) = (-1)^n A^n = 0

b) triangular A inverse \Leftrightarrow triangular 且 $A^{-1} = \frac{1}{\det A} \text{adj} A$ 若 A upper triangular, $C_{ij} = 0$ i > j
且 $C_{ij} \in \mathbb{F}$ $\Rightarrow C_{ij}(A)$ upper $\Rightarrow A \# D(I+B)$, $A^{-1} = (I+B)^{-1} D^{-1}$ 由 $(I+B)^{-1} = I - B + B^2 - \dots + (-1)^{n-1} B^{n-1}$ (且 $B^n = 0$)
且 $A^{-1} = [a'_1 \dots a'_n]$ $AA^{-1} = [Aa'_1 \dots Aa'_n] = I \quad \therefore Aa'_k = e_k = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ k \\ 1 \\ 0 \end{bmatrix}$ the a_k k-th row of A 为 0
Note $\text{det } A \neq 0 \Leftrightarrow \text{diag}_i \neq 0$ $\therefore A^{-1}$ lower triangular

$\|x\|_1 \leq 2$

measure of $\|x\|_1$ is unit ball vector transform of ellipsoid to size

Date: $\|Ax\|_p = \sqrt{\lambda_{\max} \frac{\|x\|_1^2}{\|x\|_1}} \dots \|Ax\|_p$ induced norm by.

1. (operator norm) $\|A\|_p = \sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} = \inf \{c > 0 \mid \|Ax\|_1 \leq c\|x\|_1, \forall x\} = \|A\|_{op} = \sup_{\|x\|_1 \leq 1} \|Ax\|_1$

记忆为 norm \square property i) $\forall A, x \in \mathbb{R}^n, \|Ax\|_1 > 0$, $\forall A \sup_{\|x\|_1 \leq 1} \|Ax\|_1 > 0$, $\forall A \sup_{\|x\|_1 \leq 1} \|Ax\|_1 \geq 0$

当 $A=0, \|A\|_p = \sup_{\|x\|_1 \leq 1} \|Ax\|_1 = 0$, 而是 $\|A\|_{op} = 0$, $\sup_{\|x\|_1 \leq 1} \|Ax\|_1 = 0$, $\forall \|x\|_1 \leq 1 \Rightarrow \|Ax\|_1 = 0 \Rightarrow Ax = 0 \Rightarrow A=0$

ii) $\sup_{\|x\|_1 \leq 1} \|Ax\|_1 = \sup_{\|x\|_1 \leq 1} \|\lambda_1 \|_1 \|x\|_1 = \lambda_1 \sup_{\|x\|_1 \leq 1} \|x\|_1$ iii) MRTS $\|A\|_{op} + \|B\|_{op} = \sup_{\|x\|_1 \leq 1} \|Ax + Bx\|_1$

$\sup_{\|x\|_1 \leq 1} \|Ax + Bx\|_1 \geq \|Ax\|_1 + \|Bx\|_1, \forall \|x\|_1 \leq 1$ \square xy 代换, 取 $u = x - y, \|u\|_1 \leq 1$, 有上式 $= \|A\|_{op} + \|B\|_{op}$

$\geq \|A + B\|_{op}, \dots \text{by def } \|A\|_{op} + \|B\|_{op} = \sup_{\|x\|_1 \leq 1} \|Ax + Bx\|_1 = \|A + B\|_{op}$

2. $\forall \|.\|_a$ cont. on V.S. V under topology induced by $\|.\|_1$ (unit. cont.)

\square WTS $\forall \varepsilon > 0 \exists \delta > 0$ s.t. $\|x - x'\|_1 < \delta \Rightarrow |\|x\|_a - \|x'\|_a| < \varepsilon$ or $\sup_{x \in V}$

\square 由 property iii) on $\|.\|_a$ $\left| \begin{array}{l} \|x\|_a - \|x'\|_a = \|x' + (x-x')\|_a - \|x'\|_a \leq \|x-x'\|_a \\ \|x\|_a - \|x'\|_a = \|x - (x-x')\|_a - \|x'\|_a \leq \|x-x'\|_a \end{array} \right. \therefore \text{RHS} \leq \|x-x'\|_a \max_{x \in V} \|x\|_a$

\square x under V to basis $\{e_i\}$ sc. $x = \sum d_i e_i, x' = \sum d'_i e_i, \|x-x'\|_a \leq \sum |d_i - d'_i| \cdot \|e_i\|_a \leq \|x-x'\|_1$

$\therefore \text{存在 } \delta = \frac{\varepsilon}{M} \quad \|x-x'\|_1 < \delta \text{ 有 } |\|x\|_a - \|x'\|_a| \leq \|x-x'\|_a \leq \|x-x'\|_1 \cdot M < \varepsilon$

iii) (non-equivalent) $\forall 2 \|\cdot\|_a, \|\cdot\|_b$ on V.S. V eqn, $\exists \exists \text{ real } 0 < C_1 \leq C_2 \text{ st. } \forall x \in V, C_1 \|x\|_b \leq \|x\|_a \leq C_2 \|x\|_b$

Step 1) (norm transitivity) 若 $\|\cdot\|_a, \|\cdot\|_b$ 等价, $\|\cdot\|_1$, $\|\cdot\|_a$ 等价 $\|\cdot\|_b$

\square 有 $C_1 \|x\|_1 \leq \|x\|_a \leq C_2 \|x\|_1, \therefore \frac{C_1}{C_2} \|x\|_1 \leq \|x\|_b \leq \frac{C_2}{C_1} \|x\|_1 \text{ 由 eqn.}$

Step 2) 只须证 $C_1 \|x\|_1 \leq \|x\|_a \leq C_2 \|x\|_1$. $\# x=0$ 为 trivial, $\# x \neq 0$, 取 $u = \frac{x}{\|x\|_1}$ \square $C_1 \leq \|u\|_a \leq C_2$

Note $\text{Hull}_1 = 1$. 由 Extreme Value Thm, cont. $\|\cdot\|_a$ on compact set $\{u \mid \|u\|_1 = 1\}$ 必有 $\max_{\text{closed \& bounded}} \text{min}$ in set

$\therefore \text{令 } C_1 = \min_{\|u\|_1=1} \|u\|_a \quad \square u \neq 0, C_2 > C_1 > 0, \text{ 且 } C_1 \leq \|u\|_a \leq C_2$

3. 讨 $\|A\|$ induced norm of matrix norm

ver norm def

\square i) $\# A \neq 0$ 有 $\|x\|=1$, st. $Ax \neq 0$, $\therefore \|Ax\| \neq 0$, 又若 $A=0$ 则 $\forall x \neq 0 \Rightarrow Ax=0$, $\therefore \|Ax\|=0$

ii) $\|cA\| = \max \{cAx \mid x \in \mathbb{R}^n\} = |c| \max \{Ax \mid x \in \mathbb{R}^n\}$ (induced) $\forall g \neq 0$ 有 $\bar{x} = \frac{g}{\|g\|_1}$ 使 $\|x\|=1$

iii) $\|Ax\|_1 = \|x\|_1 \& \|Ax\|_1 = \|(A+B)x\|_1, \therefore \|Ax + Bx\|_1 \leq \|Ax\|_1 + \|Bx\|_1 \therefore \|Ay\|_1 = \|A\| \|y\|_1 \times 1 = \|y\|_1 \cdot \|Ax\|_1 \leq \|y\|_1 \cdot \|A\|_1$

$\leq \|A\|_1 \|x\|_1 + \|B\|_1 \|x\|_1$

(induced) of matrix norm

iv) $\|Ax\|_1 = 1$, 有 $\|ABx\|_1 = \|AB\|_1$

$\therefore \|ABx\|_1 \leq \|A\|_1 \cdot \|Bx\|_1 \leq \|A\|_1 \|B\|_1 \|x\|_1$

recent update $A^T A = UDU^T$ spectral decomp (A^TA)x = [A^TA]_x = Q^T[UDU^T]Qx = U⁻¹UDU^TQx = U⁻¹Ux = $\sum c_i \lambda_i x_i$

2 PSEJF note U = $\begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & 1 & \\ & & & \ddots \end{pmatrix}$ note $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n \geq 0$ & orthogonal eigenvectors $\{u_i\}_{i=1}^n$

$\square \|A\|^2 = \langle x, A^T Ax \rangle$ 因 $A^T A$ opt. P. P_{def} , 有 $\lambda_1 \geq \dots \geq \lambda_n \geq 0$ & orthogonal eigenvectors $\{u_i\}_{i=1}^n$

$\therefore \|x\|=1, \# c_1 x_1 + \dots + c_n x_n, c_1 x_1 + \dots + c_n x_n \rangle$

$\therefore \|x\|=1, \# c_1 x_1 + \dots + c_n x_n \geq c_1 \lambda_1 x_1 + \dots + c_n \lambda_n x_n \geq c_1 \lambda_1 x_1 = \lambda_1$

$\therefore \|A\|^2 = \max \{c^T A^T A c \mid c \in \mathbb{R}^n\} = \lambda_1$

$\max_{\text{def}} \lambda_i(A)$

5. (Rayleigh) $\# A P$ pd 有 $\lambda_{\min} \|x\|^2 \leq x^T Px \leq \lambda_{\max} \|x\|^2$ R PCT.

Inner product & norm

def norm $\|x\|, \mathbb{R}^n \rightarrow \mathbb{R}$

- $\|x\| \geq 0$ if $x \neq 0$, $\|x\| = 0 \Leftrightarrow x = 0$
- $\|ax\| = |a|\|x\|$
- (triangle) $\|x+y\| \leq \|x\| + \|y\|$

$$L_1 \|x\|_1 = \sum_i |x_i|$$

$$L_2 \sqrt{\sum_i |x_i|^2}$$

$$\text{Determinate } \sqrt{\sum_i |x_i|^2}$$

$$\max_i |x_i|$$

1. $\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ inner product iff

a. positivity $\langle x, x \rangle \geq 0$ ($x = 0 \Leftrightarrow \langle x, x \rangle = 0$)

b. symmetry $\langle x, y \rangle = \langle y, x \rangle$

c. additivity

$$\langle x+y, z \rangle = \langle x, z \rangle + \langle y, z \rangle$$

d. homogeneity $\langle rx, y \rangle = r \langle x, y \rangle, r \in \mathbb{R}$

P. i) $\langle x, y+z \rangle = \langle y+z, x \rangle = \langle y, x \rangle + \langle z, x \rangle = \langle x, y \rangle + \langle x, z \rangle$

ii) $\langle x, ry \rangle = \langle ry, x \rangle = r \langle y, x \rangle = r \langle x, y \rangle$

2. def Euclidean inner product $\langle x, y \rangle = \sum_i x_i y_i = x^T y$

$$\text{norm } \|x\| = \sqrt{\langle x, x \rangle} = \sqrt{x^T x}$$

$$\text{PP} (\sum_i x_i y_i)^2 \leq (\sum_i x_i^2)(\sum_i y_i^2)$$

P. i) (Cauchy-Schwarz) $|\langle x, y \rangle| \leq \|x\| \|y\|, x \neq 0$ 且 $y \neq 0$, $x, y \in \mathbb{R}^n$

□ $\|x\| = \|y\| = 1, \underbrace{\langle x-y \rangle^2}_{1. a} = \|x\|^2 - 2\langle x, y \rangle + \|y\|^2 = 2 - 2\langle x, y \rangle, \therefore \langle x, y \rangle \leq 1$

由 $x, y \neq 0$, 且 $x \neq y$. 令 $x' = x/\|x\|, y' = y/\|y\|$, now $\|x'\| = \|y'\| = 1 \therefore \langle x', y' \rangle = \frac{\langle x, y \rangle}{\|x\| \|y\|} \leq 1$

且令 $x' = -x$, $\langle x', y \rangle = -\langle x, y \rangle \leq \|x\| \|y\|$ 上行未 $\therefore |\langle x, y \rangle| \leq \|x\| \|y\| \frac{x}{\|x\|} = \frac{y}{\|y\|}$

ii) $\|x\| \geq 0$, 且 $x = 0$ 1.a

norm $\|x\| = \sqrt{\langle x, x \rangle} = \sqrt{x^T x}$ 1.d

iii) (homogeneity) $\forall r \in \mathbb{R}, \|rx\| = |r| \|x\|$

$$\square \sqrt{\langle rx, rx \rangle} = \sqrt{r^2 \langle x, x \rangle}$$

iv) (triangle) $\|x+y\| \leq \|x\| + \|y\|$

□ $\|x+y\|^2 = \|x\|^2 + 2\langle x, y \rangle + \|y\|^2 \leq (\|x\| + \|y\|)^2$

v) $\|x\| - \|y\| \leq \|x-y\|$ 1.v

□ $\|x\| = \|x-y+y\| \leq \|x-y\| + \|y\|, \therefore \|x\| - \|y\| \leq \|x-y\| \quad \begin{cases} \text{if } \|y\| - \|x\| \leq \|y-x\| \\ = \|x-y\| \end{cases}$

Note $\|\cdot\|$ 为 unit. dist: $\forall \varepsilon > 0$, $\exists \delta = \delta(\varepsilon)$, 使 $\|x-y\| < \delta \Rightarrow \delta \leq \|x\| - \|y\| \leq \varepsilon$ 及左 T_i

3. 在 \mathbb{C}^n , inner product $\langle x, y \rangle := \sum_i x_i \bar{y}_i$

P. i) $\langle x, y \rangle = \overline{\langle y, x \rangle}$

□ 由左: $\sum_i x_i \bar{y}_i = \sum_i \bar{y}_i \bar{x}_i = \sum_i \bar{y}_i x_i$

ii) $\langle x, r_1 y + r_2 z \rangle = \bar{r}_1 \langle x, y \rangle + \bar{r}_2 \langle x, z \rangle$ □ $\langle r_1 y + r_2 z, x \rangle = r_1 \langle y, x \rangle + r_2 \langle z, x \rangle = \bar{r}_1 \langle y, x \rangle + \bar{r}_2 \langle z, x \rangle$

4. matrix norm 为向量的推广了 property, 应满足 $\|AB\| \leq \|A\| \|B\|$ eg. Frobenius $\|A\|_F = \sqrt{\sum_{i,j} A_{ij}^2}$
 为 connect matrix norm & vector norm, ∴ 为 induced norm (且 Lp norm equiv)
 $\| \cdot \|_{(m)} \| \cdot \|_{(n)}$ vec norm on $\mathbb{R}^{m \times n}$ matrix norm $\| \cdot \|$ induced by $\| \cdot \|_{(m)}$ $\| \cdot \|_{(n)}$ $\|Ax\|_{(m)} \leq \|A\| \|x\|_{(n)}$

eg. $\|A\| = \max_{\|x\|_{(n)}=1} \|Ax\|_{(m)}$ 为 induced norm (\mathbb{R}^m)

NOTE 由左 T_i $\forall A \exists \|x_0\| \geq 1$ st. $\|Ax_0\| = \|A\|$ (ip $\|A\|$ 为 attainable)

5. spectral radius $r(A) = \max \{|\lambda(A)|\}$ note $\|A\| \geq \|Ax\| = |\lambda| \|x\| = r(A)$, induce norm 为 radius. 由左 T_i $\|A\|_2 \leq \max_{\|x\|=1} \|Ax\|$ 得力

Date.

THE 1. real symmetric A diagonal G.R

$$\square \lambda(\bar{x}^T x) = \bar{x}^T (\lambda x) = (\bar{x}^T A)x \stackrel{\text{Asymmetric}}{=} (\bar{A}\bar{x})^T x = (\bar{\lambda}\bar{x})^T x = \bar{\lambda}(\bar{x}^T x)$$

$$\therefore (\lambda - \bar{\lambda})(\bar{x}^T x) = 0 \quad \lambda = \bar{\lambda}$$

$\langle x, x \rangle > 0, \bar{x} = 0 \Rightarrow x = 0$ i.e. x is zero

1. (Spectral Thm for symmetric A)

i) λ eigenvalue $\lambda \in \mathbb{R}$ iff \exists eigenv. real (T.E. 2)

ii) distinct evol in even orthogonal ($\mathbf{U}_P \mathbf{U}$)

iii) \exists diag $D \in \mathbb{R}^{n \times n}$ & orthogonal $\mathbf{U} \in \mathbb{R}^{n \times n}$ s.t. $A = \mathbf{U}D\mathbf{U}^T$, $D = \text{diag}(\lambda_1, \dots, \lambda_n)$ \mathbf{U} orthogonal

schw. Jmp \rightarrow spectral decomposition \square 4th defn. 4th diag. \square $\mathbf{U}^T \mathbf{U} = \mathbf{I}$, $\therefore A = \mathbf{U}D\mathbf{U}^T = \mathbf{U}D\mathbf{U}^{-1}$

iv) $\{v_1, \dots, v_n\}$ eigenv. orthogonal basis ($\forall i, j \neq i, j$)

if $\lambda_i \neq \lambda_j$ then v_i & v_j distinct $\square \mathbf{U}$ full-rank, L.I.M. $\{v_i\}$, \mathbf{U} ortho-normal

T.E. 2 diagonalizable A to orthogonal evol $\Rightarrow A$ symmetric

$$\square A \xrightarrow{\text{defn. 4}} A = P D P^{-1} \quad A^T = P^{-1} D^T P^T = \underbrace{P^T D P^{-1}}_{P^T = P} = A$$

then $\mathbf{U} = \mathbf{P}$ $\mathbf{U}^T = \mathbf{P}^T$ (defn. 1)

2. (Gram-Schmidt Process)

\square L.I.M. $\{v_1, \dots, v_k\} \rightarrow$ orthonormal L.I.M. $\{e_1, \dots, e_k\}$. b/c $\text{Proj}_u(v) = \frac{\langle u, v \rangle}{\langle u, u \rangle} u$, $\text{Proj}_u(u) = u$

$$\square \begin{aligned} u_1 &= v_1, & e_1 &= \text{norm}(u_1) = \frac{u_1}{\|u_1\|} \\ u_2 &= v_2 - \text{proj}_{u_1}(v_2), & e_2 &= \text{norm}(u_2) \\ &\vdots & & \text{Induction hypothesis: } u_i \perp u_j \quad \text{for all } i < j \\ u_k &= v_k - \sum_{i=1}^{k-1} \text{proj}_{u_i}(v_k), & e_k &= \text{norm}(u_k) \end{aligned}$$

if v_j is 2nd L.I.M. in v then $\langle u_i, u_j \rangle = 0$

if $u_j = 0$, break Alg

if $u_k = v_k - \sum_{i=1}^{k-1} \text{proj}_{u_i}(v_k)$ where proj_i is projection onto u_i then $u_k = v_k - \sum_{i=1}^{k-1} \langle u_i, v_k \rangle \cdot e_i$

证 3 A symmetric A^{-1} symmetric $\square (A^{-1})^T (A^{-1})^{-1} = A^{-1}$

2. 对 $(0, 0)$ symmetric B 的 spectral cleavage by \mathbf{U} (eg. natural basis I), 因为由 process on A

• λ, v_i evol, find orthogonal $\{v_1, v_2, \dots, v_n\}$ basis, $\mathbf{U} = (v_1 | v_2 | \dots | v_n)$, note $\mathbf{U}^T A \mathbf{U}$ symmetric

$$\bullet \text{1st col of } \mathbf{U}^T A \mathbf{U}, (U_1^T A)_{11} = U_1^T \lambda_1 v_1 = \begin{pmatrix} -v_1 \\ -v_2 \\ \vdots \\ -v_n \end{pmatrix} \begin{pmatrix} \lambda_1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} \lambda_1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \therefore (U_1^T A \mathbf{U})_{11} = \begin{pmatrix} \lambda_1 & 0 \\ 0 & B \end{pmatrix}, \text{由 T.E. 3 symmetric}$$

• apply process on B .. until find orthogonal basis

证 4. $\{x_1, \dots, x_n\}$ mutual orthogonal, $x_i \neq 0$, 且 $\{x_1, \dots, x_n\}$ L.I.M. $\square \text{if } \sum \alpha_i x_i = 0 \Rightarrow (\sum \alpha_i x_i)^T x_j = 0 \quad \sum \alpha_i x_i^T x_j = 0$

证 5 A symmetric A $\lambda_{\min}(A^{-1}) = \frac{1}{\lambda_{\max}(A)}$ & $\lambda_{\max}(A^{-1}) = \frac{1}{\lambda_{\min}(A)}$

\square A real IR \square T.E. 1

$$\sum \alpha_i x_i^T x_j = 0 \quad \alpha_j = 0$$

orthogonal & eigen

$$T_0: x \perp y \Leftrightarrow \|x+y\|^2 = \|x\|^2 + \|y\|^2 \quad \text{if } \|x+y\| \neq 0$$

Date:

Generalized orthogonal basis

1. x, y orthogonal iff $\langle x, y \rangle = 0$ // A orthogonal matrix iff $A^T = A^{-1}$ ($A^T A = I$)
 Generalization of relation P.E. offdiag. & permutation Pst // if all of 1's are orthogonal then $\sum_{i,j} A_{ij} A_{ij}^T = I$ (orthonormal)
 $\square A_{ij} (A^T A)_{ij} = I_{ij} = 0 = \sum_{k} A_{ki} A_{kj}^T$ col dot product
 $i \in \text{BSS } P_{ij} = 1$ then it's row j orthonormal

2. ~~$A \neq 0 \Rightarrow \lambda_i, i=1, \dots, n$ are eigen. P. Th.~~ (Ex 2)

\square If A is symmetric A , T is orthogonal \Rightarrow all eigenvectors U, V associate with eval $\lambda \neq 0$, U, V orthogonal

$$\square C^T A V = \lambda C^T V, X = (A^T D)^T V = \lambda V^T V, \therefore R \nmid V^T V = 0 \Rightarrow \lambda V^T V = \lambda C^T V$$

$$(b) A = \sum_{i=1}^n \lambda_i U_i U_i^T, \therefore \{(\lambda_i, U_i)\}_{i=1}^n \text{ are eigen pairs}, U_i \text{ normalized}$$

$$\square A I = A \equiv A \sum U_i U_i^T, \text{ since } A U_i = \lambda_i U_i, \text{ but } U_i^T \text{ sum over } i, \sum A U_i U_i^T = \sum \lambda_i U_i U_i^T = A I = A$$

$$(c) I = \sum U_i U_i^T \{U_i\} \text{ eigenvectors}$$

$$\square \{U_i\} \text{ orthogonal basis, } \therefore A H X = \frac{1}{K} \partial_K V_K, \text{ since } \sum_i U_i U_i^T X = \sum_i \sum_k U_i U_i^T \partial_k V_k = \sum_k \partial_k V_k = X = \text{Dir. Krasner's lemma, orthogonality sum}$$

3. $\{V\}$ is orthogonal component $V^\perp = \{X \mid V^T X = 0, \forall V \in V\}$ ($\forall V \oplus V^\perp = \mathbb{R}^n$, $\forall V$ subspace)

\square transform T orthogonal projector onto V , if $\forall x \in \mathbb{R}^n$, $T(x) \in V \quad \forall x \in V^\perp$

~~P orthogonal projector onto subspace $V = R(P)$~~ $\Leftrightarrow P^2 = P = P^T$ (symmetric)

$$\square \Rightarrow R(I-P) \subseteq R(P)^\perp, \therefore \forall x \in V \quad \text{if } P^T(I-P)x = 0, \therefore P^T(I-P) = 0 = P^T - P^T P$$

$$\therefore P = P^T = (P^T P)^T = P^T P = P^T \quad \Leftrightarrow \forall x \in V, (P^T)^T(I-P)x = P^T P(I-P)x = 0 \quad \therefore (I-P)x \in R(P)^\perp$$

$$2a) R(A)^\perp = N(A^T) \quad \text{and} \quad N(A)^\perp = R(A^T)$$

$$\square 1^o. \subseteq \{x \in R(A)\}^\perp, \forall v \in R(A), \exists y \in V \quad v = Ay, \therefore (Ay)^T x = 0, y^T A^T x = 0, \text{ but } A^T x = 0$$

$$\supseteq \{x \in N(A^T)\} \quad \forall v \in R(A) \quad \text{since } v^T x = y^T A^T x = 0 \quad \text{LA: } \mathbb{R}^n \rightarrow \mathbb{R}^m$$

2' apply L on both sides $\square 2^o$ $\text{op } N(A) \text{ vec } \perp A^T \text{ col span vector}$

$$2b) V^\perp = V \quad \Leftrightarrow \begin{cases} a_1, \dots, a_n \in V \\ b_1, \dots, b_n \in V^\perp \text{ basis, by def } V = R(A) = [a_1 \dots a_n] \end{cases} \quad \therefore \text{WTS } R(P) \subseteq R(A)$$

$$\not\subseteq \{x \in N(A^T)\} \quad \therefore \forall v \in V \quad \forall x \in V, (Av)^T x = 0 = v^T (A^T x), \text{ but } v \text{ arbitrary, } A^T x = 0.$$

$$2c) V \subseteq W \Rightarrow W^\perp \subseteq V^\perp \quad \square \forall w \in W \quad \forall w^\perp \in W^\perp \quad \text{if } w^\perp w = 0 \Rightarrow w \in V^\perp \Rightarrow w \in W^\perp$$

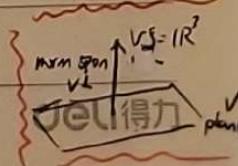
$$2d) P \text{ orthogonal decomposition VS. } \forall x \in \mathbb{R}^n, P x + (I-P)x \in V^\perp \quad \text{but unique}$$

$$2e) P \text{ orthogonal on } V, \quad \text{if } P x = x, \forall x \in V \quad \text{then } P(P) = V \quad \text{and } N(P) = V^\perp$$

$$\square 1' \times \text{orthogonal decompose } R x + (I-P)x, \quad \text{if } x = x + 0 \in V \times V^\perp, \therefore P x = x \quad \square \text{by def } R(P) \subseteq V$$

$$3^o \quad \text{if } R \perp N(P)^\perp = V^\perp \perp V = R(P^T) = R(I-P)$$

$$2f) \forall x \in V, \|P x\| \leq \|x\| \quad \square \quad \text{if } \|x\| = \|P x + (I-P)x\| = \sqrt{\|P x\|^2 + \dots} \leq \|P x\| + \dots$$



Orthogonal 2.

Date:

1. matrix of projection matrix if $A^2 = A$ (idempotent)

若 $\langle P\mathbf{x}, \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle$ 为 orthogonal projection, 若 \mathbf{x} orthogonal to \mathbf{y}

$\mathbf{x}, \mathbf{y} \in xy\text{-plane}$ $\Rightarrow P = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ 有 $P^2(\mathbf{x}) = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = P(\mathbf{x})$

P_i : projection of eval = 0/1 $\square A^2 = \lambda \mathbf{v} = A\mathbf{v} = \lambda \mathbf{v}$, $\therefore (\mathbf{A}^{-1}\mathbf{A})\mathbf{v} = 0$ if $\mathbf{v} \neq 0$

i) proj $A\mathbf{B}$ 不等于 proj \mathbf{B} $\Rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 \\ 0 & 0 \end{pmatrix}$

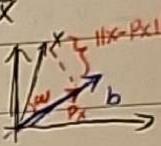
$P_{AB}^2 \neq P_{AB} \Rightarrow P_{AB}^2 = P_{BA}$ $(AB)^2 = ABAB = AAAB = A^2B^2 = AB$

2. construct projection on 1-D line span by \vec{b} of vector \mathbf{x}

• project \mathbf{x} on line, $\therefore P\mathbf{x} = \lambda \vec{b}$

• projected $P\mathbf{x}$ 与 \mathbf{x} closed, $P\mathbf{x}$ min $\|\mathbf{x} - P\mathbf{x}\|$

只当垂直, $\because \langle P\mathbf{x} - \mathbf{x}, \vec{b} \rangle = 0 = \langle \mathbf{x} - \mathbf{x}, \vec{b} \rangle \Leftrightarrow \lambda = \frac{\vec{b}^T \mathbf{x}}{\|\vec{b}\|^2}$
 • \therefore projected $P\mathbf{x} = \frac{\vec{b}^T \mathbf{x}}{\|\vec{b}\|^2} \vec{b}$ Note 由 $\vec{b}^T \vec{b} = 1$ $\Rightarrow \|P\mathbf{x}\| = \|\mathbf{x}\| = \|\mathbf{x} - P\mathbf{x}\| = \sqrt{\|\mathbf{x}\|^2 - \|\mathbf{x} - P\mathbf{x}\|^2} = \sqrt{1 - \left(\frac{\vec{b}^T \mathbf{x}}{\|\vec{b}\|^2}\right)^2 \|\vec{b}\|^2} = \sqrt{1 - \left(\frac{\vec{b}^T \mathbf{x}}{\|\vec{b}\|}\right)^2} = \|\mathbf{x}\| \cdot \frac{\|\vec{b}\|}{\|\mathbf{x}\|}$



$$\therefore P\mathbf{x} = \frac{\vec{b}^T \mathbf{x}}{\|\vec{b}\|^2} \vec{b} = \frac{\vec{b} \vec{b}^T}{\|\vec{b}\|^2} \mathbf{x} \quad P_i: P\mathbf{x} \neq P\mathbf{x} \text{ vec, eval = 1}$$

2° Project on $V \subseteq \mathbb{R}^n$, dim $V = m \geq 1$ 有 basis $\{\vec{b}_1, \dots, \vec{b}_m\}$

• project \mathbf{x} on V , $P\mathbf{x} = \sum \lambda_i \vec{b}_i = [\vec{b}_1 \dots \vec{b}_m] \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_m \end{bmatrix} = B\lambda$

• 垂直/orthogonal to all V 's basis $\{\vec{b}_1, \dots, \vec{b}_m\}$, $\therefore \langle P\mathbf{x} - \mathbf{x}, \vec{b}_i \rangle = 0 = \vec{b}_i^T (\mathbf{x} - B\lambda) \Leftrightarrow \begin{pmatrix} -b_1^T & & \\ & \ddots & \\ & & -b_m^T \end{pmatrix} (\mathbf{x} - B\lambda) = 0$

$$\text{pseudo-inverse} \quad \langle P\mathbf{x} - \mathbf{x}, \vec{b}_m \rangle = 0 = \vec{b}_m^T (\mathbf{x} - B\lambda) \Leftrightarrow B^T (\mathbf{x} - B\lambda) = 0 \Leftrightarrow B^T B\lambda = B^T \mathbf{x}$$

• \therefore projected $P\mathbf{x} = B\lambda$, $P = B(B^T B)^{-1} B^T$ ($\because m=1$, $B^T B$ scalar, reduced 1°)

例 1. 1° \exists line $\langle \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} \rangle$, projection $P = \frac{1}{9} \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix} \begin{pmatrix} 1 & 2 & 2 \end{pmatrix}^T = \frac{1}{9} \begin{pmatrix} 1 & 2 & 2 \\ 2 & 4 & 4 \\ 2 & 4 & 4 \end{pmatrix}$, 且对任意 $\mathbf{x} \in \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ $P(\mathbf{x}) = \frac{1}{9} \begin{pmatrix} 3 \\ 10 \\ 10 \end{pmatrix}$

2° 2D plane $V = \langle \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix} \rangle$, $P = B(B^T B)^{-1} B^T = \begin{pmatrix} 1 & 0 \\ 1 & 2 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} 5 & 2 & -1 \\ 2 & 2 & 2 \\ 2 & 2 & 5 \end{pmatrix}^{-1} \begin{pmatrix} 1 & 0 \\ 1 & 2 \\ 2 & 2 \end{pmatrix}^T = \frac{1}{10} \begin{pmatrix} 5 & 2 & -1 \\ 2 & 2 & 2 \\ 2 & 2 & 5 \end{pmatrix}^T$, $P(\mathbf{x}) = B\lambda = B \begin{pmatrix} 5 \\ 2 \\ 2 \end{pmatrix}^T$

应用 least square PP find $A\mathbf{x} = \mathbf{b}$ 无解时的解: find $\tilde{\mathbf{x}} \in G(A)$ 使 $\tilde{\mathbf{x}}$ 与 \mathbf{b} 最近

(orthogonal projection of \mathbf{b} onto $R(A)$)

quadratic form & norm 2 $\rightarrow P_{24}$

1. quadratic form is $f: \mathbb{R}^n \rightarrow \mathbb{R}$, $x \rightarrow x^T Q x$, $Q \in \mathbb{R}^{n \times n}$

$$\sum_{i,j} Q_{ij} x_i x_j \text{ 是高 } 2 \times 2 \text{ 的 quadratric Date.}$$

NOTE Q ~~is~~ symmetric, 若非, $JTF Q < \frac{1}{2}(Q + Q^T)$ 然後 Q new symmetric

and quadratic form $x^T Q x \left\{ \begin{array}{ll} \text{positive definite} & > 0 \text{ for } x \neq 0 \\ \text{positive semi-definite} & \geq 0 \quad \forall x \\ \text{negative definite} & < 0 \quad \forall x \neq 0 \\ \text{negative semi-definite} & \leq 0 \quad \forall x \end{array} \right.$

2. (Sylvester Criterion) $x^T Q x$ ps definte iff all leading principal minor of $Q > 0$

\square 1. Δ_i i-th leading p.m. $\because x = \sum_{i=1}^n x_i \vec{e}_i = \sum_{i=1}^n x_i \vec{v}_i$ $\therefore x = [v_1 \ v_2 \ \dots \ v_n] \tilde{x}$ $\therefore x^T Q x =$

$$d_{ii} =$$

$$x^T V^T Q V \tilde{x} = \tilde{x}^T \bar{Q} \tilde{x} \text{ when } \bar{Q} = V^T Q V \text{ note } \bar{Q}_{ij} = \langle v_i, Q v_j \rangle \text{ 希望 } \bar{Q}_{ij} = 0, i \neq j$$

要 build bases v_i os $\begin{cases} v_1 = d_{11} e_1 \\ v_2 = d_{21} e_1 + d_{22} e_2 \\ \vdots \\ v_n = d_{n1} e_1 + \dots + d_{nn} e_n \end{cases}$ 希望 p.m. Δ_i st. $\forall i$ 有 $\tilde{x}^T \bar{Q} \tilde{x} > 0$

$$\Rightarrow \begin{bmatrix} q_{11} & q_{12} & \dots & q_{1n} \\ q_{21} & q_{22} & \dots & q_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ q_{n1} & q_{n2} & \dots & q_{nn} \end{bmatrix} \begin{bmatrix} d_{11} & 0 & \dots & 0 \\ 0 & d_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_{nn} \end{bmatrix} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

$$\text{由 } \Delta_{ii} = \frac{1}{d_{ii}} \det \begin{pmatrix} q_{11} & \dots & q_{1i} \\ q_{21} & \dots & q_{2i} \\ \vdots & \vdots & \vdots \\ q_{i1} & \dots & q_{ii} \end{pmatrix}$$

$$\therefore \text{sys on } \begin{cases} d_{11} q_{11} + d_{22} q_{21} + \dots + d_{nn} q_{n1} = 0 \\ d_{11} q_{12} + d_{22} q_{22} + \dots + d_{nn} q_{n2} = 0 \\ \vdots \\ d_{11} q_{1n} + d_{22} q_{2n} + \dots + d_{nn} q_{nn} = 0 \end{cases}$$

$$\therefore \begin{cases} d_{11} q_{11} + d_{22} q_{21} + \dots + d_{nn} q_{n1} = 0 \\ d_{11} q_{12} + d_{22} q_{22} + \dots + d_{nn} q_{n2} = 0 \\ \vdots \\ d_{11} q_{1n} + d_{22} q_{2n} + \dots + d_{nn} q_{nn} = 1 \end{cases}$$

$$\text{由 } \bar{Q} = \begin{bmatrix} d_{11} & 0 & \dots & 0 \\ 0 & d_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_{nn} \end{bmatrix} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

$$\therefore \Delta_{ii} > 0 \Leftrightarrow \text{p.d.} \Rightarrow \text{sum square, } > 0$$

$$\Leftrightarrow \text{若 } \exists k \text{ st. } \Delta_k = 0 = \det Q_k = \det \begin{pmatrix} q_{11} & \dots & q_{1k} \\ \vdots & \ddots & \vdots \\ q_{k1} & \dots & q_{kk} \end{pmatrix}$$

$$\therefore \Delta_k = 0 \Rightarrow \text{p.s.} \Rightarrow \text{sum square, } > 0$$

Note \exists p.s. \Leftrightarrow all p.m. ≥ 0 , e.g. $\begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{pmatrix}$ 为 eval 1, 4, 4, 14, 14, 14. Indefinite (Optimal P, P₂₄)

3. Rayleigh quotient $R_A(x) = \frac{x^T A x}{x^T x}$, note i) $R_A(x) = R_A(\alpha x)$, $\alpha \neq 0, \vec{x} \neq \vec{0}$
A symmetric ii) $\exists x \in A$ eigvec with eval, $R_A(x) = \lambda$

\mathcal{T}_1 $\|x\|_2 = 1$ 有 $\lambda_{\min} \leq x^T A x \leq \lambda_{\max}$ λ w.r.t A

\square 证 λ_{\max} $A = U D U^T$, $\therefore y = U^T x$, $\max_{\|x\|=1} x^T A x = \max_{\|y\|=1} y^T D y = \max_{\|y\|=1} \sum_{i=1}^n \lambda_i y_i^2$

由 $\forall i \in I = \{ \text{index of } \lambda_{\max} \} \quad \sum_{i \in I} y_i^2 = 1$, $\text{and } \|y\|_2^2 = \sum_{i \in I} \lambda_i y_i^2 = \lambda_{\max} \sum_{i \in I} y_i^2 = \lambda_{\max}$

由 i) $\forall x \in (m - \max)$ $\|x\|_2 \neq 0$ $\lambda_{\min} \leq R_A(x) \leq \lambda_{\max}$

4. 2D matrix induced $\|A\|_2$, 取 A 有 $\|A^{-1}\| \geq (\|A\|)^{-1}$ $\square \frac{\|AA^{-1}\|}{\|A\| \cdot \|A^{-1}\|} \leq 1$ $\therefore \frac{\|AA^{-1}\|}{\|A\| \cdot \|A^{-1}\|} \leq 1$ $\therefore \frac{\|AA^{-1}\|}{\|A\| \cdot \|A^{-1}\|} \leq 1$

Perturbation Matrix

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(col representation)

I. ~~if~~ permutation $\pi : \{1 \dots m\} \rightarrow \{1 \dots m\}$: $(\underbrace{1 \dots m}_{\pi(1) \dots \pi(m)})$ 有 permutation P_{mm} , $P_{ij} = \begin{cases} 1 & \pi(i)=j \\ 0 & \text{otherwise} \end{cases}$

$$P = \begin{bmatrix} -e_{\pi(1)} \\ \vdots \\ -e_{\pi(m)} \end{bmatrix}$$

$$\text{e.g. } \pi = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 \\ 1 & 4 & 2 & 5 & 3 \end{pmatrix} \quad P = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 \end{pmatrix}$$

ii) $P \circ \sigma$ permutes v_i , $P \circ \sigma = \begin{bmatrix} v_{\pi(\sigma(1))} \\ \vdots \\ v_{\pi(\sigma(n))} \end{bmatrix}$ $\square v_i = \sum_k p_{ik} v_k = v_{\pi(\sigma(i))}$

~~Orthogonal if perm~~ iii) ~~Orthogonal~~ $P^{-1} = P^T = P_{\pi^{-1}}$ $\square P_{\pi} P_{\pi}^T v_{\sigma(i)} = \sum_k p_{ik} p_{k\sigma(i)}^T = \sum_k p_{ik} p_{kj}^T = \sum_k p_{ik} p_{jk} = \delta_{ij} \therefore P_{\pi} P_{\pi}^T = I_n$

iv) $P_{\sigma} P_{\pi} v = P_{\pi \circ \sigma} v$, $\pi \circ \sigma(i) = \pi(\sigma(i))$; $v^T P_{\sigma} P_{\pi} = v^T P_{\pi \circ \sigma}$

v) $v^T P$ permutes v_i , $[v_1 \dots v_n] P = [v_{\pi(1)} \dots v_{\pi(n)}]$

vi) $\exists k \text{ s.t. } P_{\pi}^k = I_n$ \square ~~每~~ $\pi \circ \pi \circ \dots$ ~~为全排列~~ \Rightarrow ~~perm~~ k 为 perm. order

vii) eval to unity $z^n = e^{i2\pi n/m}$ (in unit complex circle) \square Orthogonal, \therefore preserve $\|Pv\|$

$\|Pv\| = \|v\| \Rightarrow \|v\| = 1$, 又有 $P^k = I \therefore \lambda^k = 1 \therefore \lambda = e^{i2\pi k/m}, m=123..$ Algebra P

Unitary / Orthogonal 3

P51.1

Date: / /

1. \mathbb{C}^n matrix U , \mathcal{U} Unitary if $U^H U = I_n = U U^H$

$\mathcal{P}_i)$ $| \det U | = 1 \Leftrightarrow I_2$ ii) $\langle Ux, Uy \rangle = \langle x, y \rangle$ (preserve complex inner product) $\begin{aligned} & (Ux)^H Uy \\ & = x^H U^H Uy \\ & = x^H y \end{aligned}$

iii) preserve \mathbb{R} vector norm, $\forall x, \|Ux\|_2 = \|x\|_2$ $x^T U^T Ux = x^T x$ $\begin{aligned} & \text{Op angle } \theta \\ & \cos \theta = \frac{x^T y}{\|x\|_2 \|y\|_2} \in \text{young} \end{aligned}$

iv) (Takagi Factorization) \forall symmetric A , Unitary U , real ≥ 0 diagonal Σ s.t. $A = U \Sigma U^T$

U cols \Rightarrow orthonormal eigenvectors for $A^T A$, Σ diagonal entry \Rightarrow square root of $A^T A$ eigenval

v) U unitary \Leftrightarrow col orthonormal set $\square U^H U = [U^H u_1 \dots U^H u_n] = I_n$ $\forall j (U^H u_j)_{ij} = \delta_{ij}$

NOTE $\mathcal{P}_vi)$ Unitary \Leftrightarrow row orthonormal $(U^T U)$

vi) AB unitary $\Leftrightarrow A^T, B^T$ unitary $\square A^T B^T A^T = A A^T = I$; $B^T A^T A B = I$

T U unitary, \mathcal{U} triangular \Leftrightarrow diag entry > 0 , $U = I$

\square U_i has n_i columns $\times U_i$ has n_i rows: $u_1 = e_1, \dots, u_{n_i} = e_{n_i} \quad \|\lambda\|_1 = 1 \Rightarrow u_1 = e_1, \dots$

2. (non-square orthonormal set) $\mathcal{U} A_{mn}$ has col orthonormal, s.t. $A^H A = I_n$

$\square (A^H A)_{ij} = \sum_k A_{ik}^H A_{kj} = \sum_k A_{ki} A_{kj} = \langle a_i, a_j \rangle = \delta_{ij}$

3. eval of unitary U in complex unit circle. $\square \mathcal{P}_{iii}) \quad \|Ux\|_2 = |\lambda| \|x\|_2$ thus $|\lambda| = 1$

4. (Q_{mn} col orthonormal note $\& m \geq n$, SVD $V = I_{mn}, \Sigma = I_m, U = (Q_1 \dots)$)

$\square m < n$ 时 可能有比 Q 大的 $Q_m \in \mathbb{R}^{m \times m}$ 但不会比 m 多的 U_{mn}

$\therefore Q^T Q = I_n, Q_i = 1, \therefore \Sigma = I_{mn}, V = I_n \quad \text{so } Q_{mn} = U_{mn} I_{mn} I_n = (Q_{mn} | U_{mn} \dots U_m)$

where U_i orthonormal. note all U_i s.t. $Q^T Q = I_n$ (i.e. $\{U_{mn} \dots U_m\}$ L.I.) \Rightarrow Gram Schmidt.

5. $\|U\|_2 = 1 \quad \square \|U\|_2 = \sup_{x \neq 0} \frac{\|Ux\|_2}{\|x\|_2} = 1 \quad (\mathcal{P}_{iii}))$

6. (preserve metric L_2 norm) $\|AU\|_2 = \sup_{\|x\|=1} \|AUx\|_2 = \sup_{\|x\|=1} \|AUx\|_2 = \sup_{\|y\|=1} \|Ay\|_2 = \|A\|_2$

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RF:

$$a_3 = 2a_1 + 4a_2$$

例 1 Rank 分解 $A = \begin{pmatrix} 1 & 3 & 1 & 4 \\ 2 & 7 & 3 & 9 \\ 1 & 5 & 3 & 1 \\ 1 & 2 & 0 & 8 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & -2 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$ $\square C = \begin{pmatrix} 1 & 3 & 4 \\ 2 & 2 & 1 \\ 1 & 2 & 8 \end{pmatrix}$, $F = \begin{pmatrix} 1 & 0 & -2 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

例 2 QR 分解 $A = \begin{pmatrix} 12 & -51 & 4 \\ 6 & 167 & -68 \\ -4 & 24 & -41 \end{pmatrix}$ \square col. L.H.A. [A] PGS. 2 Gram process $\rightarrow \begin{bmatrix} 12 & -69 & -4/5 \\ 6 & 158 & 6/5 \\ -4 & 30 & -33 \end{bmatrix} \rightarrow \begin{bmatrix} 6/7 & 4/15 & -4/35 \\ 3/7 & 14/15 & 6/35 \\ -2/7 & 8/15 & -28/35 \end{bmatrix}$
 $Q^T A = Q^T Q R = R = \begin{bmatrix} 14 & 21 & -14 \\ 0 & 175 & -70 \\ 0 & 0 & 35 \end{bmatrix}$ \square ~~Q = QT~~ $Q'' = Q \cdot L \cdot Q$

例 3 用 QR 解 sys eq. $A\vec{x} = \vec{b}$. $Q^T R\vec{x} = \vec{b} \Leftrightarrow R\vec{x} = Q^T \vec{b}$, 因 R upper \Delta, 用 backward sub
 答 $\begin{pmatrix} r_{11}x_1 + \dots + r_{1n}x_n \\ r_{21}x_1 + \dots + r_{2n}x_n \\ \vdots \\ r_{m1}x_1 + \dots + r_{mn}x_n \end{pmatrix} = Q^T \vec{b}$

Decomposition

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1. $\forall A_{m \times n} \text{ rk} = r$ [rank decomposition] $\Rightarrow A_{mn} = C_{mr} F_{rn}$ full-rank
 \Leftrightarrow rk $C = r$, $F = r \times n$ $C = [\vec{c}_1 \dots \vec{c}_r]$, $F = [\vec{f}_1 \dots \vec{f}_r]$, $\vec{c}_i \in \mathbb{C}^m$, $\vec{f}_j \in \mathbb{C}^n$, $C^T F = A$
 $\vec{c}_i = \sum_{j=1}^r c_{ij} \vec{f}_j$

$P_1a)$ not unique. $\exists V \in \mathbb{R}^{r \times r}$, $C \in \mathbb{C}^{r \times r}$, $F \in \mathbb{R}^{r \times n}$ rank decomposition

$$\begin{cases} m < n \\ \text{rk } A = m \end{cases} \Rightarrow A = I_m A$$

$P_1b)$ $\begin{cases} m < n \\ \text{rk } A = m \end{cases} \Rightarrow A = A I_n$

2. $\forall A_{m \times n} \text{ rk} = n$ [QR decomposition] $\Rightarrow A = Q R$ unitary $\rightarrow [R_{nn}]$ 且 = $\bar{Q} R$

Q 为 A 的转置列向量 Gram-Schmidt Process 得正交矩阵 $[q_1 \dots q_n]$, $\forall k \in [n]$ 有

$$Q = (q_1 \dots q_n) \quad R = \begin{pmatrix} r_{11} & \dots & r_{1n} \\ 0 & \ddots & 0 \end{pmatrix} \quad \text{且 } Q^T Q = I_n, \quad R \text{ 为 } \geq 0 \quad q_i^T v_k = q_i^T c_k$$

$P_2a)$ Q 为 A 的转置列向量 (then \Rightarrow invertible), R 为 ≥ 0 , $\det A = \det Q \det R$ $\therefore |\det A| = |\det R| \neq 0$

$P_2b)$ QR 为唯一

3. LU decomposition $\xrightarrow{\text{S. 算法}}$

4. UV decomposition $\xrightarrow{\text{S. *}}$

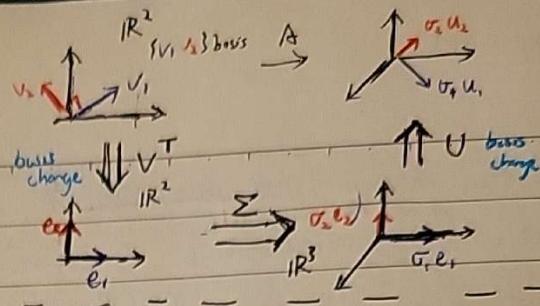
5. \forall pos. chollesky / $L U^T$ up to P

5G

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例1. SVD 几何意义

$A \in \mathbb{R}^{3 \times 2}$
 $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$
 $B \subset \mathbb{R}^2, \mathbb{R}^m$ standard basis
 $\bar{B}, \bar{C} \subset \mathbb{R}^2, \mathbb{R}^m$ 任意 basis



- $V \xrightarrow[B \rightarrow \bar{B}]{} Q \quad \therefore V^T = V^{-1} \xrightarrow[B \rightarrow \bar{B}]{} Q^T$, 将 $[X]_{\bar{B}}$ 转换为 \bar{B} 表示
- Σ scale each dim by σ_i 将 A 的 dim 变成 \mathbb{R}^m (\bar{C} 表示)
- $U \xrightarrow[\bar{C} \rightarrow C]{} Q$, 令 output of $T(x)$ 处于 $[T(x)]_C$

应用1

A movie rating

若 i) all user rate use 1 base

ii) rating 无 noise

iii) U_i typical movie

U_i typical viewer

例2. SVD $A = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ 0 & \frac{1}{\sqrt{3}} & \frac{2}{\sqrt{6}} \end{pmatrix} \begin{pmatrix} 4 & 2 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \end{pmatrix}$ 基础 for $N(A), R(A)$

1° U orthogonal, $\therefore U^T$ 为 bijective $\therefore Ux = v \Leftrightarrow x = U^{-1}v \therefore$ 只用考虑 x st. $\sum U^T x = v$ (由 3.ii)

易知, 故 y st. $\sum y = 0 - y$ 应 $\begin{pmatrix} 0 \\ * \\ * \end{pmatrix}$ 时 应令 x 满足 $U^T x = \begin{pmatrix} 0 \\ y \\ 0 \end{pmatrix}$ (由 3.ii) $\Rightarrow x = U \begin{pmatrix} 0 \\ y \\ 0 \end{pmatrix} = U$ 3rd col / 4rd row

2° U^T bijective \therefore 只用考虑 $U \sum \sqrt{\lambda} \text{mp input } x = \begin{pmatrix} a \\ b \\ 0 \end{pmatrix}$ 因 $\sum x = \begin{pmatrix} a \\ b \\ 0 \end{pmatrix} \therefore U \sum x = \begin{pmatrix} \frac{4}{\sqrt{2}} & -\frac{2}{\sqrt{2}} & 0 \\ \frac{4}{\sqrt{2}} & \frac{2}{\sqrt{2}} & 0 \\ 0 & 0 & 0 \end{pmatrix}$

$$= \begin{pmatrix} \frac{4}{\sqrt{2}} \\ \frac{4}{\sqrt{2}} \\ 0 \end{pmatrix} a + \begin{pmatrix} -\frac{2}{\sqrt{2}} \\ \frac{2}{\sqrt{2}} \\ 0 \end{pmatrix} b, R = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Decomposition 2

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1. SVD

$$A_{mn} = \boxed{U} \begin{matrix} \xrightarrow{\text{orthogonal}} \\ \Sigma \\ \xleftarrow{\text{orthogonal}} V^T \end{matrix}$$

$$\boxed{m} A = \boxed{U} \boxed{\Sigma} \boxed{V^T}$$

$$\Sigma \text{ 为 } \Delta \begin{pmatrix} \sigma_1 & 0 & & \\ 0 & \sigma_2 & & \\ & & \ddots & \\ 0 & 0 & \dots & \sigma_n \end{pmatrix}, \text{ or } R \begin{pmatrix} \sigma_1 & 0 & & \\ 0 & \sigma_2 & & \\ & & \ddots & \\ 0 & 0 & \dots & 0 \end{pmatrix}$$

$$\sum_i \sigma_i A_{2r} \quad \sigma_1 \geq \dots \geq \sigma_r > \sigma_{r+1} = \dots = \sigma_{\min(n,m)} = 0$$

由 SVD 定理, $A^T A$ 有 singular val σ_i , $\sigma_i^2(A) = \lambda_i(A^T A) \geq 0$ 且 $A^T A$ 为 optim. P.

$$A = \sum_{i=1}^r \sigma_i u_i v_i^T \quad \text{# sum of } r \times 1 \text{ matrix } \Rightarrow J_s$$

2. 特征 SVD:

$$1^\circ A^T A \text{ symmetric psd, } \therefore \text{eigen diagonalize } \Rightarrow P \# P^{-1} = P \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} P^T, \quad \lambda_i \geq 0 \quad \text{psd}$$

$$\text{但若 SVD } \exists, \quad A^T A = V \Sigma^T V^T V \Sigma V^T = V \begin{bmatrix} \sigma_1^2 & & \\ & \ddots & \\ & & \sigma_n^2 \end{bmatrix} V^T \Rightarrow \begin{cases} V^T = P^T \\ \sigma_i^2 = \lambda_i \end{cases}$$

$\therefore A^T A$ 为 eigenvectors (orthonormal by Spectral Thm) $\Rightarrow V^T \# W / V \text{ col}$

$$2^\circ \{V_i\}_{i=1}^r \text{ orthonormal, } \{AV_i\}_{i=1}^r \text{ 在 } m \geq r \text{ 的 basis } \Rightarrow (AV_i)^T (AV_j) = V_i^T (A^T A) V_j = V_i^T \lambda_j V_j = 0$$

$$3^\circ \text{ 因此用 } \{AV_i\}_{i=1}^r \text{ 为 } W, \text{ 为 orthonormal } \quad u_i = \frac{AV_i}{\|AV_i\|} = \frac{AV_i}{\sqrt{\lambda_i \|V_i\|^2}} = \frac{AV_i}{\sqrt{\lambda_i \|V_i\|^2}} = \frac{AV_i}{\sigma_i} \quad i=1 \dots r$$

$$4^\circ \text{ 对 remaining } m-r \text{ 用 Lhd. vec w.r.t } u_{i=r+1} \text{ # Gram-Schmidt orthonormalize } \quad (\text{ singular val eq})$$

$$u_i \# A A^T \text{ vec: } A A^T u_i = A \cdot (A^T u_i)$$

3. P_i orthogonal & 有