

第七次习题课

一. 线性映射

1. $\phi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 线性映射与子空间

2. ϕ 单射 $\Leftrightarrow \ker(\phi) = \{\vec{0}_n\}$

hw1. 设 ϕ 是 $\mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性映射且双射, ϕ^{-1} 是 ϕ 的逆映射.

证: (1) ϕ^{-1} 是线性映射;

(2) $n = m$.

证明: (1) $\phi^{-1}: \mathbb{R}^m \rightarrow \mathbb{R}^n$

$\forall \vec{y}_1, \vec{y}_2 \in \mathbb{R}^m, \forall \alpha, \beta \in \mathbb{R} \quad \exists! \vec{x}_1, \vec{x}_2 \in \mathbb{R}^n$

s.t. $\phi(\vec{x}_1) = \vec{y}_1, \phi(\vec{x}_2) = \vec{y}_2$.

$\therefore \vec{x}_1 = \phi^{-1}(\vec{y}_1), \vec{x}_2 = \phi^{-1}(\vec{y}_2)$

$\therefore \phi(\alpha\vec{x}_1 + \beta\vec{x}_2) = \alpha\phi(\vec{x}_1) + \beta\phi(\vec{x}_2) = \alpha\vec{y}_1 + \beta\vec{y}_2$

$\therefore \phi^{-1}(\alpha\vec{y}_1 + \beta\vec{y}_2) = \alpha\vec{x}_1 + \beta\vec{x}_2$

$\alpha\vec{x}_1 + \beta\vec{x}_2 = \alpha\phi^{-1}(\vec{y}_1) + \beta\phi^{-1}(\vec{y}_2)$

$\therefore \phi^{-1}(\alpha\vec{y}_1 + \beta\vec{y}_2) = \alpha\phi^{-1}(\vec{y}_1) + \beta\phi^{-1}(\vec{y}_2)$

$\therefore \phi^{-1}$ 是线性映射.

(2). 由对偶定理可知: $\dim(\ker(\phi)) + \dim(\text{im}(\phi)) = n$.

ϕ 是单射 $\Rightarrow \ker \phi = \{\vec{0}_n\} \Rightarrow \dim(\ker(\phi)) = 0$.

ϕ 是满射 $\Rightarrow \text{im } \phi = \mathbb{R}^m \Rightarrow \dim(\text{im}(\phi)) = m$

因此 $m = n$.

注: $\dim(\ker(\phi)) + \dim(\text{im}(\phi)) = n \quad \phi: \mathbb{R}^n \rightarrow \mathbb{R}^m$

① ϕ 是单射, 则 $n \leq m$

ϕ 是单射 $\Leftrightarrow \ker(\phi) = \{\vec{0}_n\} \Leftrightarrow \dim(\ker(\phi)) = 0$

$\Leftrightarrow \dim(\text{im}(\phi)) = n$

$\text{im}(\phi) \subseteq \mathbb{R}^m \Rightarrow n \leq m$.

② φ 是满射, 则 $n \geq m$.

φ 是满射 $\Leftrightarrow \text{im}(\varphi) = \mathbb{R}^m \Leftrightarrow \dim(\text{im}(\varphi)) = m$

3. $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$\sum_{j=1}^n \alpha_j \vec{v}_j \mapsto \sum_{j=1}^n \alpha_j \vec{w}_j \quad \alpha_1, \dots, \alpha_n \in \mathbb{R}.$$

若 $A \in \mathbb{R}^{m \times n}$, $\varphi_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 满足 $\varphi_A(\vec{e}_j) = \vec{A}^{(j)}$

hw 3. $\varphi: \mathbb{R}^3 \rightarrow \mathbb{R}^4$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \mapsto \begin{pmatrix} x_1 + x_2 \\ x_1 - x_3 \\ x_2 + x_3 \\ x_1 + 2x_2 + x_3 \end{pmatrix}$$

证: (1) φ 是线性映射:

$$\forall \vec{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}, \vec{y} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \in \mathbb{R}^3, \alpha, \beta \in \mathbb{R}$$

验证 $\varphi(\alpha\vec{x} + \beta\vec{y}) = \alpha\varphi(\vec{x}) + \beta\varphi(\vec{y})$ 即可.

(2). 由定义

$A = (\varphi(\vec{e}_1), \varphi(\vec{e}_2), \varphi(\vec{e}_3))$ 计算出

$$\varphi(\vec{e}_1) = \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix}, \quad \varphi(\vec{e}_2) = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 2 \end{pmatrix}, \quad \varphi(\vec{e}_3) = \begin{pmatrix} 0 \\ -1 \\ 1 \\ 1 \end{pmatrix}.$$

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

回顾: 旋转 $R_\theta: \mathbb{R}^2 \rightarrow \mathbb{R}^2$
 $R_\theta(\vec{e}_1, \vec{e}_2) = (\vec{e}_1, \vec{e}_2) \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix}$

$$\varphi(\vec{e}_1, \vec{e}_2, \vec{e}_3) = (\vec{e}_1, \vec{e}_2, \vec{e}_3, \vec{e}_4) \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & -1 \\ 0 & 1 & 1 \\ 1 & 2 & 1 \end{pmatrix}$$

$\begin{matrix} \uparrow & & \uparrow \\ \mathbb{R}^3 & & \mathbb{R}^4 \end{matrix}$

(3). 利用初等行变换得

$$A \rightarrow \dots \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\text{rank}(A) = 2 \quad \therefore \dim(\text{im } \varphi) = 2.$$

$$\text{由秩定理得: } \dim(\ker \varphi) = 3 - 2 = 1.$$

注: 此时是一个未知量.

(4) $\ker(\varphi)$ 对应的齐次线性方程组是

$$\begin{cases} x_1 + x_2 = 0 \\ x_2 + x_3 = 0 \end{cases} \iff \begin{cases} x_1 = -x_2 \\ x_3 = -x_2 \end{cases}$$

于是, $\ker(\varphi)$ 的一组基是 $\begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$.

$$\therefore \text{im}(\varphi) = V_c(A) \quad \text{且} \quad \dim(\text{im } \varphi) = 2.$$

$\therefore A$ 中任意两个线性无关的列向量是 $\text{im}(\varphi)$ 的一组基.

例: $\text{im}(\varphi)$ 的一组基是 $\begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \\ 2 \end{pmatrix}$

$$4. \ker(\varphi_A) = \text{sol}(H) \text{ 且 } L \text{ 相容} \iff \vec{b} \in \text{im}(\varphi).$$

$$5. a. \varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$\dim(\ker(\varphi)) + \dim(\text{im}(\varphi)) = n$$

b. $\phi: \mathbb{R}^n \rightarrow \mathbb{R}^n$ 中单 \Leftrightarrow ϕ 满

hw 2. 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^n$ 线性映射, 且 $\forall \vec{x} \in \mathbb{R}^n, \varphi(\vec{x}) \in \langle \vec{x} \rangle$.

证明: $\exists \lambda \in \mathbb{R}$ s.t. $\forall \vec{x} \in \mathbb{R}^n \varphi(\vec{x}) = \lambda \vec{x}$.

证: 设 $\vec{v}_1, \dots, \vec{v}_n$ 是 \mathbb{R}^n 的一组基. 且 $\varphi(\vec{v}_i) = \lambda_i \vec{v}_i, \lambda_i \in \mathbb{R}$
 $i=1, 2, \dots, n$.

下证 $\lambda_1 = \lambda_2 = \dots = \lambda_n$

取 $\vec{x} = \vec{v}_1 + \dots + \vec{v}_n$, 则 $\exists \lambda \in \mathbb{R}$ s.t. $\varphi(\vec{x}) = \lambda \vec{x}$

$$\begin{aligned} \text{则 } \varphi(\vec{v}_1 + \dots + \vec{v}_n) &= \varphi(\vec{v}_1) + \dots + \varphi(\vec{v}_n) = \lambda_1 \vec{v}_1 + \dots + \lambda_n \vec{v}_n \\ &= \lambda (\vec{v}_1 + \dots + \vec{v}_n) \end{aligned}$$

$$\Rightarrow (\lambda_1 - \lambda) \vec{v}_1 + \dots + (\lambda_n - \lambda) \vec{v}_n = \vec{0}$$

$\because \vec{v}_1, \dots, \vec{v}_n$ 线性无关 $\therefore \lambda_1 = \lambda_2 = \dots = \lambda_n = \lambda$

$\forall \vec{y} \in \mathbb{R}^n \exists! \alpha_1, \dots, \alpha_n \in \mathbb{R}$ s.t. $\vec{y} = \alpha_1 \vec{v}_1 + \dots + \alpha_n \vec{v}_n$

$$\begin{aligned} \Rightarrow \varphi(\vec{y}) &= \varphi(\alpha_1 \vec{v}_1 + \dots + \alpha_n \vec{v}_n) \\ &= \alpha_1 \varphi(\vec{v}_1) + \dots + \alpha_n \varphi(\vec{v}_n) \\ &= \lambda (\alpha_1 \vec{v}_1 + \dots + \alpha_n \vec{v}_n) \\ &= \lambda \vec{y}. \end{aligned}$$

□

hw 5. 证: " \Leftarrow "

$$\varphi(\vec{x}) = \begin{pmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{pmatrix}, \quad \varphi(\vec{y}) = \begin{pmatrix} f_1(\vec{y}) \\ \vdots \\ f_m(\vec{y}) \end{pmatrix}$$

$$\begin{aligned} \alpha \varphi(\vec{x}) + \beta \varphi(\vec{y}) &= \begin{pmatrix} \alpha f_1(\vec{x}) + \beta f_1(\vec{y}) \\ \vdots \\ \alpha f_m(\vec{x}) + \beta f_m(\vec{y}) \end{pmatrix} = \begin{pmatrix} f_1(\alpha \vec{x} + \beta \vec{y}) \\ \vdots \\ f_m(\alpha \vec{x} + \beta \vec{y}) \end{pmatrix} \\ &= \varphi(\alpha \vec{x} + \beta \vec{y}) \end{aligned}$$

$\therefore \varphi$ 是线性映射.

$$" \Rightarrow " \quad \alpha \varphi(\vec{x}) + \beta \varphi(\vec{y}) = \varphi(\alpha \vec{x} + \beta \vec{y})$$

$$\therefore \alpha f_i(\vec{x}) + \beta f_i(\vec{y}) = f_i(\alpha \vec{x} + \beta \vec{y}).$$

f_1, \dots, f_m 为线性函数

二. 矩阵的运算

6.1 线性映射在标准基下的矩阵表示

$$\varphi(\vec{e}_j) = \sum_{i=1}^m a_{ij} \vec{e}_i$$

$$A = (\varphi(\vec{e}_1), \dots, \varphi(\vec{e}_n)) = (a_{ij})_{m \times n}$$

$$\underline{\Phi}: \text{Hom}(\mathbb{R}^n, \mathbb{R}^m) \longrightarrow \mathbb{R}^{m \times n} \quad \underline{\Psi}: \mathbb{R}^{m \times n} \longrightarrow \text{Hom}(\mathbb{R}^n, \mathbb{R}^m)$$

$$\varphi \longmapsto A_\varphi \qquad A_\varphi \longmapsto \varphi_A$$

$$\underline{\Psi} \circ \underline{\Phi} = \text{id}_{\text{Hom}(\mathbb{R}^n, \mathbb{R}^m)} \quad \underline{\Phi} \circ \underline{\Psi} = \text{id}_{\mathbb{R}^{m \times n}}$$

a. $\text{im}(\varphi) = V_c(A)$, $\dim(\text{im}(\varphi)) = \text{rank}(A)$.

$$\varphi \text{ 满射} \iff A \text{ 行满秩.}$$

b. $\dim(\ker(\varphi)) = n - \text{rank}(A)$ φ 单射 $\iff A$ 列满秩

c. φ 双射 $\iff m=n$ A 满秩

6.2 线性映射的运算

映射的加法与数乘

6.3 矩阵的线性运算

推论: $\text{Hom}(\mathbb{R}^n, \mathbb{R}^m) = \{ \varphi \mid \varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m \}$.

$$\underline{\Phi}: \text{Hom}(\mathbb{R}^n, \mathbb{R}^m) \longrightarrow \mathbb{R}^{m \times n}$$

$$\varphi \longmapsto A_\varphi$$

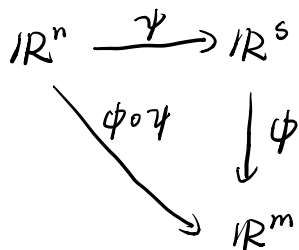
$$\underline{\Psi}: \mathbb{R}^{m \times n} \longrightarrow \text{Hom}(\mathbb{R}^n, \mathbb{R}^m)$$

$$A \longmapsto \varphi_A$$

$$\underline{\Phi}(\lambda \varphi + \mu \psi) = \lambda \underline{\Phi}(\varphi) + \mu \underline{\Phi}(\psi)$$

$$\underline{\Psi}(\lambda A + \mu B) = \lambda \underline{\Psi}(A) + \mu \underline{\Psi}(B)$$

6.4 矩阵的乘法



1. 矩阵的乘法不满足交换律, 满足结合律

2. $(AB)^t = B^t A^t$

3. $\text{rank}(A) = \text{rank}(BA) = \text{rank}(AC)$

$A \in \mathbb{R}^{m \times n}$, B 是 m 阶满秩, C 是 n 阶满秩.

7.1 $M_n(\mathbb{R})$ 上的运算

$A^t = A$: A 为对称矩阵

$A^t = -A$: A 为斜对称矩阵

$A^2 = A$: A 为幂等矩阵

$A^k = 0$: A 为幂零矩阵

7.2. 交换不限量与中心元

$$\text{tr}(AB) = \text{tr}(BA).$$

中心元: 数乘矩阵

换位元引理

eg1. 设矩阵 $A = (a_{ij}) \in \mathbb{R}^{m \times n}$, 其中 $i = 1, 2, \dots, m, j = 1, 2, \dots, n$

令

$$J_m = \begin{pmatrix} 0 & 1 & 0 & \dots & 0 & 0 \\ 0 & 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 0 & 1 \\ 0 & 0 & 0 & \dots & 0 & 0 \end{pmatrix}_{m \times m}$$

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

求矩阵乘法 $J_m A$ 和 $A J_n$.

解: $A = (a_{ij})_{m \times n}$

$$J_m A = \begin{pmatrix} a_{21} & a_{22} & \dots & a_{2n} \\ a_{31} & a_{32} & \dots & a_{3n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \\ 0 & 0 & \dots & 0 \end{pmatrix}, \quad A J_n = \begin{pmatrix} 0 & a_{11} & a_{12} & \dots & a_{1,n-1} \\ 0 & a_{21} & a_{22} & \dots & a_{2,n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & a_{m1} & a_{m2} & \dots & a_{m,n-1} \end{pmatrix}$$

搬移原理:

$$\begin{aligned} J_m A &= (E_{12}^{(m)} + E_{23}^{(m)} + \dots + E_{m-1,m}^{(m)}) A \\ &= E_{12}^{(m)} A + E_{23}^{(m)} A + \dots + E_{m-1,m}^{(m)} A \\ &= \begin{pmatrix} \vec{A}_2 \\ \vec{0} \\ \vdots \\ \vec{0} \\ \vec{0} \end{pmatrix} + \begin{pmatrix} \vec{0} \\ \vec{A}_3 \\ \vdots \\ \vec{0} \\ \vec{0} \end{pmatrix} + \dots + \begin{pmatrix} \vec{0} \\ \vec{0} \\ \vdots \\ \vec{A}_m \\ \vec{0} \end{pmatrix} = \begin{pmatrix} \vec{A}_2 \\ \vec{A}_3 \\ \vdots \\ \vec{A}_m \\ \vec{0} \end{pmatrix} \end{aligned}$$

$$\begin{aligned} A J_n &= A (E_{12}^{(n)} + E_{23}^{(n)} + \dots + E_{n-1,n}^{(n)}) \\ &= A E_{12}^{(n)} + A E_{23}^{(n)} + \dots + A E_{n-1,n}^{(n)} \\ &= (\vec{0}, \vec{A}^{(1)}, \vec{0}, \dots, \vec{0}, \vec{0}) + (\vec{0}, \vec{0}, \vec{A}^{(2)}, \dots, \vec{0}, \vec{0}) + \dots + (\vec{0}, \dots, \vec{0}, \vec{A}^{(m-1)}) \\ &= (\vec{0}, \vec{A}^{(1)}, \vec{A}^{(2)}, \dots, \vec{A}^{(n-2)}, \vec{A}^{(n-1)}) \end{aligned}$$

eg2. 设 $a, b, c \in \mathbb{R}$, $m \in \mathbb{Z}^+$, 证明

$$\begin{pmatrix} 1 & a & c \\ 0 & 1 & b \\ 0 & 0 & 1 \end{pmatrix}^m = \begin{pmatrix} 1 & ma & \frac{m(m-1)}{2} ab + mc \\ 0 & 1 & mb \\ 0 & 0 & 1 \end{pmatrix}$$

证: \therefore 可以用数学归纳法.

$$\text{二. } A = \begin{pmatrix} 0 & a & c \\ 0 & 0 & b \\ 0 & 0 & 0 \end{pmatrix}, \text{ 则 } \begin{pmatrix} 1 & a & c \\ 0 & 1 & b \\ 0 & 0 & 1 \end{pmatrix}^m = (E_3 + A)^m$$

$$= \sum_{k=0}^m \binom{m}{k} E_3^{m-k} A^k$$

注: $AE_3 = E_3A = A$.

$$A^0 = E_3, A^1 = A, A^2 = \begin{pmatrix} 0 & a & c \\ 0 & 0 & b \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & a & c \\ 0 & 0 & b \\ 0 & 0 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 0 & 0 & ab \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\dots A^k = O_3 \quad (k \geq 3).$$

$$\therefore (E_3 + A)^m = E_3 + \binom{m}{1} A + \binom{m}{2} A^2$$

$$= \begin{pmatrix} 1 & ma & \frac{m(m-1)}{2} ab + mc \\ 0 & 1 & mb \\ 0 & 0 & 1 \end{pmatrix}$$

eg 3. $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m, \psi: \mathbb{R}^s \rightarrow \mathbb{R}^n, \beta: \mathbb{R}^t \rightarrow \mathbb{R}^s$

分别对应于标准基下的矩阵 A, B, C .

$$\text{rank}(ABC) = \dim(\text{Im}(\varphi \circ \psi \circ \beta)) = \dim(\text{Im}(\varphi) / \text{Im}(\psi \circ \beta))$$

$$= \dim(\text{Im}(\psi \circ \beta)) - \dim(\text{ker}(\varphi) / \text{Im}(\psi \circ \beta))$$

$$\geq \text{rank}(BC) - \dim(\text{ker}(\varphi) / \text{Im}(\psi))$$

$$= \text{rank}(BC) - \dim(\text{Im}(\varphi)) + \dim(\text{Im}(\varphi) / \text{Im}(\psi))$$

$$= \text{rank}(BC) - \text{rank}(B) + \text{rank}(AB) \quad \overset{\text{dim}(\text{Im}(\varphi \circ \psi))}{\text{''}}$$