

回忆 定理 2.4

设 A 是 $m \times n$ 阶实矩阵. 则

$$\dim V_A + \text{rank}(A) = n$$

例 求 (L)

$$\begin{cases} x_1 + x_2 - x_3 = 0 \\ x_1 - x_2 + x_4 = 0 \end{cases} \text{ 的解空间 } L$$

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

$$\text{rank}(A) = 2 \Rightarrow \dim V_A = 4 - 2 = 2$$

由 A 可得 $\begin{cases} x_3 = x_1 + x_2 \\ x_4 = -x_1 + x_2 \end{cases}$

取 $x_1 = 1, x_2 = 0$ 得解 $\vec{v}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \\ -1 \end{pmatrix}$

取 $x_1 = 0, x_2 = 1$ 得解 $\vec{v}_2 = \begin{pmatrix} 0 \\ 1 \\ 1 \\ 1 \end{pmatrix}$

$$V_A = \{ \alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 \mid \alpha_1, \alpha_2 \in \mathbb{R} \}.$$

记号 设 $\vec{v} \in \mathbb{R}^n, S \subseteq \mathbb{R}^n, S \neq \emptyset$ ①

把 $\{\vec{v}\} + S$ 简记为 $\vec{v} + S$. 称为

S 关于 $\vec{v}$ 的平移

定理 2.5 设 (L) 是方程组

$$\begin{cases} a_{11}x_1 + \dots + a_{1n}x_n = b_1 \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n = b_m \end{cases}$$

设 $A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix}$, 且 $\vec{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \in \mathbb{R}^n$ 是

(L) 的一个解. 则 (L) 的所有解是

$$\vec{v} + V_A$$

证: 设 $\vec{w} \in V_A$. 则 $\vec{v} + \vec{w}$ 是 (L) 的解

(见第 2 次作业)

反之. 设 $\vec{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}$ 是 (L) 的解.

$$\text{则 } \vec{u} - \vec{v} = \begin{pmatrix} u_1 - v_1 \\ \vdots \\ u_n - v_n \end{pmatrix} \in V_A \quad (\text{第 2 次作业})$$

$$\vec{u} = \vec{v} + \vec{u} - \vec{v} \in \vec{v} + V_A \quad \square$$

例 求 (L)
$$\begin{cases} x_1 + x_2 - x_3 = 1 \\ x_1 - x_2 + x_4 = 2 \end{cases}$$

的所有解:

解: 由上例可知, $\text{rank}(A) = 2$.

于是 $\text{rank}(A | \begin{smallmatrix} 1 \\ 2 \end{smallmatrix}) = 2$. (L) 有解

$\vec{v} = \begin{pmatrix} 0 \\ 0 \\ -1 \\ 2 \end{pmatrix}$ 是 (L) 的一个解. 于是 (L) 的

所有解是 $\vec{v} + V_A = \begin{pmatrix} 0 \\ 0 \\ -1 \\ 2 \end{pmatrix} + \left\{ \alpha_1 \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} + \alpha_2 \begin{pmatrix} 0 \\ 1 \\ 1 \\ 1 \end{pmatrix} \mid \alpha_1, \alpha_2 \in \mathbb{R} \right\}$

关于矩阵秩的若干等式和不等式

1. 设 A 是 $m \times n$ 阶矩阵. 则

$$\text{rank}(A) \leq \min(m, n)$$

证: $V_r(A) \subset \mathbb{R}^{1 \times n} \Rightarrow \text{rank}(A) \leq n$

$V_c(A) \subset \mathbb{R}^m \Rightarrow \text{rank}(A) \leq m$



②. 定义: 如果 $\text{rank}(A) = m$, 则称 A 行满秩.
 ... $\text{rank}(A) = n$, 则称 A 列满秩.
 如果 $\text{rank}(A) = \min(m, n)$, 则称 A 满秩.

定义: 设 $A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$

则矩阵 $\begin{pmatrix} a_{11} & a_{21} & \dots & a_{m1} \\ a_{12} & a_{22} & \dots & a_{m2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{mn} \end{pmatrix}_{n \times m}$ 称为 A 的转置

记为 A^t

注: 设 $A^t = (a'_{ji})_{\substack{j=1, \dots, n \\ i=1, \dots, m}} \in \mathbb{R}$

~~a_{ji}~~ , $a'_{ji} = a_{ij}$

2. $\text{rank}(A) = \text{rank}(A^t)$

证: $\dim V_r(A^t) = \dim V_c(A)$

例: (L) $\begin{cases} x+y=1 \\ 2x-y=2 \\ 5x+2y=5 \end{cases}$

$$B = \left(\begin{array}{cc|c} 1 & 1 & 1 \\ 2 & -1 & 2 \\ 5 & 2 & 5 \end{array} \right) \rightarrow \left(\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & -3 & 0 \\ 0 & -3 & 0 \end{array} \right)$$

$$\rightarrow \left(\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & -3 & 0 \\ 0 & 0 & 0 \end{array} \right) \quad \text{rank}(A) = \text{rank}(B) = 2$$

(L) 有解.

($\tilde{L}$) $\begin{cases} x+y=1 \\ 2x-y=2.01 \\ 5x+2.01y=5 \end{cases}$

$$\tilde{B} = \left(\begin{array}{cc|c} 1 & 1 & 1 \\ 2 & -1 & 2.01 \\ 5 & 2.01 & 5 \end{array} \right) \rightarrow \left(\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & -3 & 0.01 \\ 0 & -2.99 & 0 \end{array} \right)$$

$$\rightarrow \left(\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & -3 & 0.01 \\ 0 & 0 & -0.01 \times \frac{2.99}{3} \end{array} \right)$$

$\text{rank}(A)=2$
 $\text{rank}(\tilde{B})=3$

($\tilde{L}$) 无解

§3 线性映射初步

③

§3.1 定义与基本性质

定义: 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 如

$\forall \alpha \in \mathbb{R}, \vec{x}, \vec{y} \in \mathbb{R}^n$

(i) $\varphi(\vec{x} + \vec{y}) = \varphi(\vec{x}) + \varphi(\vec{y})$

(ii) $\varphi(\alpha \vec{x}) = \alpha \varphi(\vec{x})$.

则称 φ 是线性映射.

记号 $\vec{0}_n = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix} \in \mathbb{R}^n, \vec{0}_m = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix} \in \mathbb{R}^m$

注: 由 (ii) 中取 $\alpha=0$ 可知

$$\varphi(\vec{0}_n) = \vec{0}_m$$

命题 3.1 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性的

$\alpha_1, \dots, \alpha_k \in \mathbb{R}, \vec{v}_1, \dots, \vec{v}_k \in \mathbb{R}^n$

则 (i) $\varphi\left(\sum_{i=1}^k \alpha_i \vec{v}_i\right) = \sum_{i=1}^k \alpha_i \varphi(\vec{v}_i)$

(ii) 如果 $\vec{v}_1, \dots, \vec{v}_k$ 线性相关, 则

$\varphi(\vec{v}_1), \dots, \varphi(\vec{v}_k)$ 也线性相关

(iii) 如果 $\varphi(\vec{v}_1), \dots, \varphi(\vec{v}_k)$ 线性无关. 则

$\vec{v}_1, \dots, \vec{v}_k$ 也线性无关

证: (i) 对 k 归纳, $k=1$, 由定义中(ii)

$$\text{可知 } \varphi(\alpha_1 \vec{v}_1) = \alpha_1 \varphi(\vec{v}_1).$$

设 $k-1$ 时结论成立. 当 k 时

$$\begin{aligned} & \varphi(\alpha_1 \vec{v}_1 + \dots + \alpha_{k-1} \vec{v}_{k-1} + \alpha_k \vec{v}_k) \\ &= \varphi(\alpha_1 \vec{v}_1 + \dots + \alpha_{k-1} \vec{v}_{k-1}) + \varphi(\alpha_k \vec{v}_k) \quad [\text{定义中(i)}] \end{aligned}$$

$$= \alpha_1 \varphi(\vec{v}_1) + \dots + \alpha_{k-1} \varphi(\vec{v}_{k-1}) + \alpha_k \varphi(\vec{v}_k) \quad [\text{归纳法}]$$

(ii) 设 $\beta_1 \vec{v}_1 + \dots + \beta_k \vec{v}_k = \vec{0}_n$, 其中 $\beta_1, \dots, \beta_k \in \mathbb{R}$

不全为零. 由 (i) 可知

$$\vec{0}_m = \varphi(\beta_1 \vec{v}_1 + \dots + \beta_k \vec{v}_k) = \beta_1 \varphi(\vec{v}_1) + \dots + \beta_k \varphi(\vec{v}_k)$$

(iii) 是 (ii) 的逆否命题 图

例: 零映射: $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$
 $\vec{x} \mapsto \vec{0}_m$

$$\varphi(\vec{x} + \vec{y}) = \vec{0}_m = \vec{0}_m + \vec{0}_m = \varphi(\vec{x}) + \varphi(\vec{y})$$

$$\varphi(\alpha \vec{x}) = \vec{0}_m = \alpha \vec{0}_m = \alpha \varphi(\vec{x}). \quad (4)$$

恒同映射: $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^n$
 $\vec{x} \mapsto \vec{x}$

$$\varphi(\vec{x} + \vec{y}) = \vec{x} + \vec{y} = \varphi(\vec{x}) + \varphi(\vec{y})$$

$$\varphi(\alpha \vec{x}) = \alpha \vec{x} = \alpha \varphi(\vec{x})$$

定义: 设 $n \leq m$.

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

$$\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mapsto \begin{pmatrix} x_1 \\ \vdots \\ x_n \\ 0 \\ \vdots \end{pmatrix} \quad \left. \vphantom{\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}} \right\} m-n$$

$$\varphi\left(\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}\right) = \varphi\left(\begin{pmatrix} x_1 + y_1 \\ \vdots \\ x_n + y_n \end{pmatrix}\right) = \begin{pmatrix} x_1 + y_1 \\ \vdots \\ x_n + y_n \\ 0 \\ \vdots \end{pmatrix}$$

$$= \begin{pmatrix} x_1 \\ \vdots \\ x_n \\ 0 \\ \vdots \end{pmatrix} + \begin{pmatrix} y_1 \\ \vdots \\ y_n \\ 0 \\ \vdots \end{pmatrix} = \varphi\left(\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}\right) + \varphi\left(\begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}\right).$$

$$\varphi\left(\alpha \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}\right) = \varphi\left(\begin{pmatrix} \alpha x_1 \\ \vdots \\ \alpha x_n \end{pmatrix}\right) = \begin{pmatrix} \alpha x_1 \\ \vdots \\ \alpha x_n \\ 0 \\ \vdots \end{pmatrix} = \alpha \begin{pmatrix} x_1 \\ \vdots \\ x_n \\ 0 \\ \vdots \end{pmatrix}$$

投影: $n \geq m$

$$\varphi: \mathbb{R}^n \longrightarrow \mathbb{R}^m$$

$$\begin{pmatrix} x_1 \\ \vdots \\ x_m \\ \vdots \\ x_n \end{pmatrix} \mapsto \begin{pmatrix} x_1 \\ \vdots \\ x_m \end{pmatrix}$$

$$\varphi \left(\begin{pmatrix} x_1 \\ \vdots \\ x_m \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} y_1 \\ \vdots \\ y_m \\ \vdots \\ y_n \end{pmatrix} \right) = \varphi \left(\begin{pmatrix} x_1+y_1 \\ \vdots \\ x_m+y_m \\ \vdots \\ x_n+y_n \end{pmatrix} \right) = \begin{pmatrix} x_1+y_1 \\ \vdots \\ x_m+y_m \end{pmatrix}$$

$$= \begin{pmatrix} x_1 \\ \vdots \\ x_m \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} y_1 \\ \vdots \\ y_m \\ \vdots \\ y_n \end{pmatrix} = \varphi \left(\begin{pmatrix} x_1 \\ \vdots \\ x_m \\ \vdots \\ x_n \end{pmatrix} \right) + \varphi \left(\begin{pmatrix} y_1 \\ \vdots \\ y_m \\ \vdots \\ y_n \end{pmatrix} \right)$$

$$\varphi \left(\alpha \begin{pmatrix} x_1 \\ \vdots \\ x_m \\ \vdots \\ x_n \end{pmatrix} \right) = \varphi \left(\begin{pmatrix} \alpha x_1 \\ \vdots \\ \alpha x_m \\ \vdots \\ \alpha x_n \end{pmatrix} \right) = \begin{pmatrix} \alpha x_1 \\ \vdots \\ \alpha x_m \end{pmatrix} = \alpha \begin{pmatrix} x_1 \\ \vdots \\ x_m \end{pmatrix}$$

$$= \alpha \varphi \left(\begin{pmatrix} x_1 \\ \vdots \\ x_m \\ \vdots \\ x_n \end{pmatrix} \right)$$

线性函数. 设 $\alpha_1, \dots, \alpha_n \in \mathbb{R}$

$$f: \mathbb{R}^n \longrightarrow \mathbb{R}$$

$$\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mapsto \alpha_1 x_1 + \dots + \alpha_n x_n$$

$$\text{设 } \vec{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \vec{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

(5)

$$f(\vec{x} + \vec{y}) = f \left(\begin{pmatrix} x_1+y_1 \\ \vdots \\ x_n+y_n \end{pmatrix} \right) = \alpha_1(x_1+y_1) + \dots + \alpha_n(x_n+y_n)$$

$$= (\alpha_1 x_1 + \dots + \alpha_n x_n) + (\alpha_1 y_1 + \dots + \alpha_n y_n)$$

$$= f(\vec{x}) + f(\vec{y})$$

设 $\lambda \in \mathbb{R}$

$$f(\lambda \vec{x}) = f \left(\begin{pmatrix} \lambda x_1 \\ \vdots \\ \lambda x_n \end{pmatrix} \right) = \alpha_1 \lambda x_1 + \dots + \alpha_n \lambda x_n$$

$$= \lambda (\alpha_1 x_1 + \dots + \alpha_n x_n) = \lambda f(\vec{x})$$

命题 3.2. 设 $\varphi: \mathbb{R}^n \longrightarrow \mathbb{R}^m$

$$\vec{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mapsto \begin{pmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{pmatrix}$$

则 φ 是线性映射 $\Leftrightarrow f_1, \dots, f_m$ 是线性函数

证: " $\Rightarrow$ " 设 $\vec{z}, \vec{y} \in \mathbb{R}^n$

$$\varphi(\vec{z} + \vec{y}) = \begin{pmatrix} f_1(\vec{z} + \vec{y}) \\ \vdots \\ f_m(\vec{z} + \vec{y}) \end{pmatrix} \Rightarrow f_i(\vec{z} + \vec{y})$$

||

$$\varphi(\vec{z}) + \varphi(\vec{y}) = \begin{pmatrix} f_1(\vec{z}) + f_1(\vec{y}) \\ \vdots \\ f_m(\vec{z}) + f_m(\vec{y}) \end{pmatrix} = \begin{pmatrix} f_i(\vec{z}) + f_i(\vec{y}) \\ \vdots \\ f_i(\vec{z}) + f_i(\vec{y}) \end{pmatrix} = f_i(\vec{z}) + f_i(\vec{y})$$

$i=1, \dots, m$

$$\varphi(\alpha \vec{x}) = \begin{pmatrix} f_1(\alpha \vec{x}) \\ \vdots \\ f_m(\alpha \vec{x}) \end{pmatrix} \Rightarrow f_i(\alpha \vec{x}) = \alpha f_i(\vec{x})$$

$$\parallel \\ \alpha \varphi(\vec{x}) = \begin{pmatrix} \alpha f_1(\vec{x}) \\ \vdots \\ \alpha f_m(\vec{x}) \end{pmatrix}$$

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

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

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

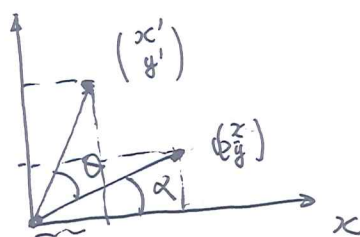
$$= \alpha \begin{pmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{pmatrix} = \alpha \varphi(\vec{x}). \quad \square$$

例: ~~旋转~~ $\varphi: \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$\begin{pmatrix} x \\ y \end{pmatrix} \mapsto \begin{pmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{pmatrix}$$

~~不是线性映射~~

y



$$\begin{aligned} x' &= x \cos \theta - y \sin \theta \\ y' &= x \sin \theta + y \cos \theta \end{aligned}$$

⑥

$$r = \sqrt{x^2 + y^2}$$

$$\begin{aligned} x' &= r \cos(\theta + \alpha) = r \cos \theta \cos \alpha - r \sin \theta \sin \alpha \\ &= x \cos \theta - y \sin \theta \end{aligned}$$

$$y' = r \sin(\theta + \alpha) = x \sin \theta + y \cos \theta.$$

例: 平移 设 $\vec{v} \in \mathbb{R}^n \setminus \{\vec{0}\}$

$$\begin{aligned} \varphi: \mathbb{R}^n &\rightarrow \mathbb{R}^n \\ \vec{x} &\mapsto \vec{x} + \vec{v} \end{aligned}$$

不是线性映射 因为 $\varphi(\vec{0}_n) = \vec{v} \neq \vec{0}_n$

例: 二次函数 $f: \mathbb{R} \rightarrow \mathbb{R}$

$$x \mapsto x^2$$

$$f(x+y) = (x+y)^2 = x^2 + 2xy + y^2$$

$$f(x) + f(y) = x^2 + y^2.$$

$\therefore 2 \neq 0 \therefore f$ 不是线性的.

命题 3.3 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性映射

U 是 $\mathbb{R}^n$ ~~线性子空间~~ 子空间, V 是 $\mathbb{R}^m$ 的子空间 则

(i) $\varphi(U)$ 是 $\mathbb{R}^m$ 的子空间, $\dim U \geq \dim \varphi(U)$

(ii) $\varphi^{-1}(V)$ 是 $\mathbb{R}^n$ 的子空间

证: (i) 设 $\vec{v}_1, \vec{v}_2 \in \varphi(U), \alpha \in \mathbb{R}$

$\exists \vec{u}_1, \vec{u}_2 \in U$ 使得

$$\varphi(\vec{u}_1) = \vec{v}_1, \varphi(\vec{u}_2) = \vec{v}_2$$

$$\Rightarrow \vec{v}_1 + \vec{v}_2 = \varphi(\vec{u}_1) + \varphi(\vec{u}_2) = \varphi(\vec{u}_1 + \vec{u}_2)$$

$$\Rightarrow \vec{v}_1 + \vec{v}_2 \in \varphi(U)$$

$$\alpha \vec{v}_1 = \alpha \varphi(\vec{u}_1) = \varphi(\alpha \vec{u}_1)$$

$$\Rightarrow \alpha \vec{v}_1 \in \varphi(U)$$

由此可知 $\varphi(U)$ 是子空间

~~验证~~

设 $\vec{v}_1, \dots, \vec{v}_d$ 是 $\varphi(U)$ 的一组基

且 $\varphi(\vec{u}_i) = \vec{v}_i, i=1, 2, \dots, d, \vec{u}_i \in U$ ⑦

由命题 3.1 (iii)

$\vec{u}_1, \dots, \vec{u}_d \in U$ 线性无关

则 $\dim U \geq d$.

(ii) 期中考试题

§ 3.2 核与像

定义: 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性映射

$\text{im}(\varphi)$ 即 $\varphi(\mathbb{R}^n)$ 称为 φ 的像空间

(证: $\text{im}(\varphi)$ 是子空间)

$\varphi^{-1}(\{\vec{0}_m\})$ 称为 φ 的核空间, 记为

$\ker(\varphi)$

验证: $\ker(\varphi)$ 是 $\mathbb{R}^n$ 的子空间

设 $\vec{v}_1, \vec{v}_2 \in \ker(\varphi)$

$$\varphi(\vec{v}_1 + \vec{v}_2) = \varphi(\vec{v}_1) + \varphi(\vec{v}_2) = \vec{0}_m + \vec{0}_m = \vec{0}_m$$

$$\varphi(\alpha \vec{v}_1) = \alpha \varphi(\vec{v}_1) = \vec{0}_m$$

$$\Rightarrow \vec{v}_1 + \vec{v}_2, \alpha \vec{v}_1 \in \ker(\varphi).$$

例: 零映射 φ : $\ker(\varphi) = \mathbb{R}^n$, $\text{im}(\varphi) = \{\vec{0}_m\}$

恒同映射: $\ker(\varphi) = \{\vec{0}_n\}$, $\text{im}(\varphi) = \mathbb{R}^m$

嵌入: $\ker(\varphi) = \{\vec{0}_n\}$, $\text{im}(\varphi) = \left\{ \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \\ 0 \\ \vdots \\ 0 \end{pmatrix} \mid \alpha_1, \dots, \alpha_n \in \mathbb{R} \right\}$

投影: $\ker(\varphi) = \left\{ \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \alpha_{m+1} \\ \vdots \\ \alpha_n \end{pmatrix} \right\}^n \mid \alpha_{m+1}, \dots, \alpha_n \in \mathbb{R}$

$$\text{im}(\varphi) = \mathbb{R}^m$$

命题 3.3 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性映射

则 φ 是单射 $\Leftrightarrow \ker(\varphi) = \{\vec{0}_n\}$

证: $\Rightarrow$ $\because \varphi$ 是单射且 $\varphi(\vec{0}_n) = \vec{0}_m$
 $\therefore \ker(\varphi) = \{\vec{0}_n\}$

$\Leftarrow$ 设 $\vec{u}, \vec{v} \in \mathbb{R}^n$ 且 $\varphi(\vec{u}) = \varphi(\vec{v})$

$$\text{则 } \varphi(\vec{u}) - \varphi(\vec{v}) = \vec{0}_m$$

$$\Rightarrow \varphi(\vec{u} - \vec{v}) = \vec{0}_m$$

$$\Rightarrow \vec{u} - \vec{v} \in \ker(\varphi) = \{\vec{0}_n\}$$

$$\Rightarrow \vec{u} - \vec{v} = \vec{0}_n$$

$$\Rightarrow \vec{u} = \vec{v}$$



例: 设 $\varphi: \mathbb{R}^2 \rightarrow \mathbb{R}^n$ 是线性映射. (8)

证 $\mathbb{R}^2$ 中 φ 是单射

$$\text{证: } \varphi\left(\begin{pmatrix} x \\ y \end{pmatrix}\right) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\text{则 } \begin{cases} x \cos \theta - y \sin \theta = 0 & \times \cos \theta \\ x \sin \theta - y \cos \theta = 0 & \times \sin \theta \end{cases}$$

$$\Rightarrow x(\cos^2 \theta + \sin^2 \theta) = 0$$

$$\Rightarrow x = 0 \Rightarrow y = 0$$

$$\Rightarrow \ker(\varphi) = \left\{ \begin{pmatrix} 0 \\ 0 \end{pmatrix} \right\} \Rightarrow \varphi \text{ 是单射.}$$

~~定理 3.1 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性映射~~
~~则 $\dim(\ker(\varphi)) + \dim(\text{im}(\varphi)) = n$~~
~~证: 设 $\vec{v}_1, \dots, \vec{v}_k$ 是 $\ker(\varphi)$ 的一组基~~
 ~~$\vec{v}_1, \dots, \vec{v}_k, \vec{v}_{k+1}, \dots, \vec{v}_n$ 是 $\mathbb{R}^n$ 的一组基~~
~~(注意若 $\ker(\varphi) = \{\vec{0}_n\}$ 则 $k=0$)~~
~~则 $\varphi(\vec{v}_{k+1}), \dots, \varphi(\vec{v}_n)$ 线性无关~~

回忆: $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性映射

如果 $\forall \vec{x}, \vec{y} \in \mathbb{R}^n, \alpha \in \mathbb{R}$

$$\varphi(\vec{x} + \vec{y}) = \varphi(\vec{x}) + \varphi(\vec{y}), \quad \varphi(\alpha \vec{x}) = \alpha \varphi(\vec{x})$$

证1. φ 线性 $\Rightarrow \varphi(\vec{0}_n) = \vec{0}_m$

证法1. 令 $\alpha = 0$.

$$\varphi(0 \cdot \vec{x}) = 0 \cdot \varphi(\vec{x}) \Rightarrow \varphi(\vec{0}_n) = \vec{0}_m$$

证法2. $\varphi(\vec{0}_n + \vec{0}_n) = \varphi(\vec{0}_n) + \varphi(\vec{0}_n)$

$$\begin{aligned} & \parallel \\ & \varphi(\vec{0}_n) \\ \Rightarrow & \varphi(\vec{0}_n) = \vec{0}_m \end{aligned}$$

证2 线性映射把“真的”映成“真的”，“平的”映成“平的”。且零向量映成零向量

证3 命题3.3 (i) 补充:

设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 线性, U 是 $\mathbb{R}^n$ 的子空间, 则

$$\dim \varphi(U) \leq \dim U.$$

证: 设 $\vec{v}_1, \dots, \vec{v}_d$ 是 $\varphi(U)$ 的一组基. ⑨

因为 $\vec{v}_1, \dots, \vec{v}_d \in \varphi(U)$, 所以存在

$\vec{u}_1, \dots, \vec{u}_d \in U$, 使得

$$\varphi(\vec{u}_1) = \vec{v}_1, \dots, \varphi(\vec{u}_d) = \vec{v}_d$$

由命题3.1 (iii)

$\vec{u}_1, \dots, \vec{u}_d$ 线性无关.

则由基扩充定理, U 有基底

$\vec{u}_1, \dots, \vec{u}_d, \vec{u}_{d+1}, \dots, \vec{u}_k, \quad k \geq d$

$$\Rightarrow \dim U = k \geq d = \dim \varphi(U). \quad \square$$

定理3.1 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 线性映射

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

证: 设 $\vec{v}_1, \dots, \vec{v}_d$ 是 $\ker(\varphi)$ 的一组基

由基扩充定理, $\mathbb{R}^n$ 有基底

$\vec{v}_1, \dots, \vec{v}_d, \vec{v}_{d+1}, \dots, \vec{v}_n$

(证: 若 $\ker(\varphi) = \{\vec{0}_n\}$, 则取 $d=0$)

断言1: $\varphi(\vec{v}_{d+1}), \dots, \varphi(\vec{v}_n)$ 线性无关.

设 $\alpha_{d+1}, \dots, \alpha_n \in \mathbb{R}$ 使得

$$\alpha_{d+1} \varphi(\vec{v}_{d+1}) + \dots + \alpha_n \varphi(\vec{v}_n) = \vec{0}_m$$

$$\text{则 } \varphi(\alpha_{d+1} \vec{v}_{d+1} + \dots + \alpha_n \vec{v}_n) = \vec{0}_m$$

$$\text{即 } \alpha_{d+1} \vec{v}_{d+1} + \dots + \alpha_n \vec{v}_n \in \ker(\varphi)$$

$\Rightarrow \exists \alpha_1, \dots, \alpha_d \in \mathbb{R}$ 使得

$$\alpha_{d+1} \vec{v}_{d+1} + \dots + \alpha_n \vec{v}_n = \alpha_1 \vec{v}_1 + \dots + \alpha_d \vec{v}_d$$

$$\Rightarrow \alpha_1 \vec{v}_1 + \dots + \alpha_d \vec{v}_d + (-\alpha_{d+1}) \vec{v}_{d+1} + \dots + (-\alpha_n) \vec{v}_n = \vec{0}_n$$

$$\Rightarrow \alpha_{d+1} = \dots = \alpha_n = 0$$

于是对任意

$$\text{对 } \varphi: \text{im}(\varphi) = \langle \varphi(\vec{v}_{d+1}), \dots, \varphi(\vec{v}_n) \rangle$$

设 $\vec{w} \in \text{im}(\varphi)$. 则 $\exists \vec{v} \in \mathbb{R}^n$ 使得

$$\varphi(\vec{v}) = \vec{w}$$

$\therefore \vec{v}_1, \dots, \vec{v}_n$ 是 $\mathbb{R}^n$ 的一组基

$\therefore \exists \beta_1, \dots, \beta_n$ 使得

$$\vec{v} = \beta_1 \vec{v}_1 + \dots + \beta_d \vec{v}_d + \beta_{d+1} \vec{v}_{d+1} + \dots + \beta_n \vec{v}_n$$

$$\begin{aligned} \vec{w} = \varphi(\vec{v}) &= \varphi(\beta_1 \vec{v}_1 + \dots + \beta_d \vec{v}_d) \\ &\quad + \varphi(\beta_{d+1} \vec{v}_{d+1} + \dots + \beta_n \vec{v}_n) \end{aligned}$$

$$= \beta_{d+1} \varphi(\vec{v}_{d+1}) + \dots + \beta_n \varphi(\vec{v}_n)$$

$$\in \langle \varphi(\vec{v}_{d+1}), \dots, \varphi(\vec{v}_n) \rangle$$

由上述两式可知

$\varphi(\vec{v}_{d+1}), \dots, \varphi(\vec{v}_n)$ 是 $\text{im}(\varphi)$ 的一组基

$$\text{从而 } \dim(\text{im}(\varphi)) = n - d.$$

例: 证明旋转 φ 是单射

$\therefore \ker(\varphi)$ 是子集

$$\therefore \dim(\ker(\varphi)) = 0$$

$$0 + \dim(\text{im}(\varphi)) = 2$$

再由: $\text{im}(\varphi) \subset \mathbb{R}^2$. 所以 $\text{im}(\varphi) = \mathbb{R}^2$.

推论 3.1 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 线性

则 (i) φ 是单射 $\Leftrightarrow \dim \operatorname{im}(\varphi) = n$

(ii) 当 $m=n$ 时

φ 是单射 $\Leftrightarrow \varphi$ 是满射

证: (i) φ 是单射 $\Leftrightarrow \ker(\varphi) = \{\vec{0}_n\}$
命题 3.3

$$\Leftrightarrow \dim(\ker(\varphi)) = 0$$

$$\Leftrightarrow \dim \operatorname{im}(\varphi) = n$$

定理 3.1

(ii) ~~由 (i)~~ $\Rightarrow$ 此时 $\operatorname{im}(\varphi) \subset \mathbb{R}^n$

由 (i) φ 是单射 $\Leftrightarrow \operatorname{im}(\varphi) = \mathbb{R}^n$ $\square$

例: 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是单射或满射

则 (i) φ 是双射

(ii) φ^{-1} 也是线性的

证: 由推论 3.1 (ii)

(14)

φ 是单射或满射 $\Rightarrow \varphi$ 是双射.

设 $\vec{x}, \vec{y} \in \mathbb{R}^n$. 则 $\exists \vec{u}, \vec{v} \in \mathbb{R}^n$
使得 $\vec{x} = \varphi(\vec{u}), \vec{y} = \varphi(\vec{v}) \Rightarrow \begin{cases} \varphi^{-1}(\vec{x}) = \vec{u} \\ \varphi^{-1}(\vec{y}) = \vec{v} \end{cases}$

$$\begin{aligned} \varphi(\vec{u} + \vec{v}) &= \varphi(\vec{u}) + \varphi(\vec{v}) = \vec{x} + \vec{y} \\ \varphi^{-1}(\vec{x} + \vec{y}) &= \varphi^{-1}(\varphi(\vec{u} + \vec{v})) = \vec{u} + \vec{v} \\ &= \varphi^{-1}(\vec{x}) + \varphi^{-1}(\vec{y}) \end{aligned}$$

类似地可证 $\forall \alpha \in \mathbb{R}$

$$\varphi^{-1}(\alpha \vec{x}) = \alpha \varphi^{-1}(\vec{x}) \quad \square$$

§ 3.3. 线性映射的矩阵表示

设 $\vec{e}^{(n)}, \dots, \vec{e}^{(n)}$ 是 $\mathbb{R}^n$ 的标准基

$\vec{e}_1^{(m)}, \dots, \vec{e}_r^{(m)}$ 是 $\mathbb{R}^m$ 的标准基

定义: 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 是线性映射

$A = (\varphi(\vec{e}^{(1)}), \dots, \varphi(\vec{e}^{(n)}))$ 称为
 φ 在 $\vec{e}^{(1)}, \dots, \vec{e}^{(n)}, \vec{\varepsilon}^{(1)}, \dots, \vec{\varepsilon}^{(m)}$ 下的矩阵
 简称为 φ 的矩阵. 记为 A_φ .

注: A 是 $m \times n$ 阶的矩阵.

$$\text{设 } \vec{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

$$\begin{aligned} \varphi(\vec{x}) &= \varphi(x_1 \vec{e}^{(1)} + \dots + x_n \vec{e}^{(n)}) \\ &= x_1 \varphi(\vec{e}^{(1)}) + \dots + x_n \varphi(\vec{e}^{(n)}) \\ &= x_1 \vec{A}^{(1)} + \dots + x_n \vec{A}^{(n)} \end{aligned}$$

$$\triangleq A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

$$\begin{aligned} \text{证} \quad \varphi(\vec{x}) &= x_1 \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} + x_2 \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} + \dots + x_n \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix} \\ &= \begin{pmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \end{pmatrix}. \end{aligned}$$

例: 零映射 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ ①

$$A_\varphi = (\underbrace{\vec{0}_m, \dots, \vec{0}_m}_n) = \mathbf{O}_{m \times n}$$

4. 恒同映射 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^n$

$$A_\varphi = (\vec{e}^{(1)}, \dots, \vec{e}^{(n)}) = E_n = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix}$$

嵌入 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m \quad n \leq m$

$$A_\varphi = (\vec{\varepsilon}^{(1)}, \dots, \vec{\varepsilon}^{(n)}) = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \\ 0 & 0 & \dots & 0 \end{pmatrix}$$

投影 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m \quad n \geq m$

$$A_\varphi = [\vec{e}^{(1)}] \begin{pmatrix} 1 & 0 & \dots & 0 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & 0 & \dots & 0 \end{pmatrix} = (E_n, \mathbf{O}_{m \times (n-m)})$$

旋转 $\varphi: \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$A_\varphi = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

例：利用定理2.4证明定理3.1

证： $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 线性

记 A_φ 为 A . 设 $\vec{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n$

$$\text{则 } \varphi(\vec{x}) = x_1 \vec{A}^{(1)} + \dots + x_n \vec{A}^{(n)}$$

$$\text{于是 } \ker(\varphi) = V_A \quad \text{im}(\varphi) = V_c(A)$$

由定理2.4

$$\dim V_A + \text{rank}(A) = n$$

$$\begin{array}{ccc} \parallel & & \parallel \\ \dim \ker(\varphi) & & \dim V_c(A) \\ & \parallel & \\ & \dim \text{im}(\varphi) & \end{array}$$

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

定理3.2 设 $\vec{v}_1, \dots, \vec{v}_n \in \mathbb{R}^m$. 则存在 $\textcircled{3}$

唯一的线性映射 φ 使得

$$\varphi(\vec{e}^{(j)}) = \vec{v}_j, \quad j=1, \dots, n$$

$$\text{证 } \varphi: \mathbb{R}^n \longrightarrow \mathbb{R}^m$$

$$\vec{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mapsto x_1 \vec{v}_1 + \dots + x_n \vec{v}_n$$

$$\text{设 } \vec{v}_j = \begin{pmatrix} v_{1j} \\ \vdots \\ v_{mj} \end{pmatrix}.$$

$$\varphi(\vec{x}) = \begin{pmatrix} x_{11}x_1 + \dots + v_{1n}x_n \\ \vdots \\ v_{m1}x_1 + \dots + v_{mn}x_n \end{pmatrix}$$

由命题3.3. φ 是线性的

设 φ 是线性映射. 满足 $\varphi(\vec{e}^{(j)}) = \vec{v}_j, j=1, \dots, n$

$$\begin{aligned} \forall \varphi(\vec{x}) &= \varphi(x_1 \vec{e}^{(1)} + \dots + x_n \vec{e}^{(n)}) \\ &= x_1 \varphi(\vec{e}^{(1)}) + \dots + x_n \varphi(\vec{e}^{(n)}) \\ &= x_1 \vec{v}_1 + \dots + x_n \vec{v}_n \\ &= \varphi(\vec{x}) \quad \square \end{aligned}$$