

# 第八次作业

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

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

证: (1)  $\varphi$  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. (具体过程同学们可自行验证)

(2) 由定义可知:

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

(3) 利用初等行变换得

$$A_\varphi \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & -1 & -1 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\therefore \text{rank}(A_\varphi) = 2 \quad \therefore \dim(\text{Im}(\varphi)) = 2.$$

由秩零定理可得  $\dim(\text{Ker}(\varphi)) = 3 - 2 = 1$

(4)  $\text{Ker}(\varphi)$  对应的齐次线性方程组是

$$\begin{cases} x_1 + x_2 = 0 \\ x_2 + x_3 = 0 \end{cases} \Leftrightarrow \begin{cases} x_1 = -x_2 \\ x_3 = -x_2 \end{cases} \quad \text{程 } \text{Ker}(\varphi) \text{ 的一组基是 } \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$

$$\therefore \text{Im}(\varphi) = \text{Vc}(A_\varphi) \text{ 且 } \dim(\text{Im}(\varphi)) = 2.$$

$\therefore A_\varphi$  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  $\begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \\ 2 \end{pmatrix}$  或  $\begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 \\ 1 \\ 1 \end{pmatrix}$  或  $\begin{pmatrix} 1 \\ 0 \\ 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 \\ 1 \\ 1 \end{pmatrix}$

[任此题中任取  $A_\varphi$  的两个列向量都是  $\text{Im}(\varphi)$  的一组基]

注:

(1) 复习:

命题 6.6. 设  $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$  线性映射. 它在标准基下的矩阵是  $A \in \mathbb{R}^{m \times n}$

(i)  $\text{Im}(\varphi) = \text{Vc}(A)$ . 从而  $\dim(\text{Im}(\varphi)) = \text{rank}(A)$

特别地,  $\varphi$  满射  $\Leftrightarrow A$  行满秩

(ii)  $\text{Ker}(\varphi)$  是  $A$  对应的齐次线性方程组的解空间. 从而  $\dim(\text{Ker}(\varphi)) = n - \text{rank}(A)$

特别地,  $\varphi$  单射  $\Leftrightarrow A$  列满秩

(iii)  $\varphi$  双射  $\Leftrightarrow m=n$  且  $A$  满秩

(2) 求矩阵的秩可以做行变换或列变换  
求解空间只能做行变换.

}  $\Rightarrow$  第(3)问做行变换后第(4)问可直接用第(3)问做变换后的矩阵.

(3) 初等行变换不改变列秩, 但会改变 ~~解~~ 空间.  
列

$\therefore \text{Im}(\varphi) = \text{Vc}(A_\varphi)$ .  $\therefore \text{Im}(\varphi)$  的基要选  $A_\varphi$  的列向量, 而不是做变换后的矩阵的列向量.

但从变换后的矩阵判断  $A_\varphi$  的哪些列向量线性无关.

$$A_\varphi = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & -1 \\ 0 & 1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \cong B. \quad \begin{aligned} A_\varphi &= (\vec{A}_\varphi^{(1)}, \vec{A}_\varphi^{(2)}, \vec{A}_\varphi^{(3)}) \\ B &= (\vec{B}^{(1)}, \vec{B}^{(2)}, \vec{B}^{(3)}) \end{aligned}$$

$\vec{B}^{(1)}, \vec{B}^{(2)}$  线性无关  $\Rightarrow \vec{A}_\varphi^{(1)}, \vec{A}_\varphi^{(2)}$  线性无关

$$\text{这是因为 } \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 2 \end{pmatrix} \xrightarrow{\text{行变换}} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} = (\vec{B}^{(1)}, \vec{B}^{(2)})$$

因为  $\vec{B}^{(1)}, \vec{B}^{(2)}, \vec{B}^{(3)}$  任取两个线性无关  $\Rightarrow \vec{A}_\varphi^{(1)}, \vec{A}_\varphi^{(2)}, \vec{A}_\varphi^{(3)}$  任取两个线性无关

$\therefore$  任取两个都是  $\text{Im}(\varphi)$  的一组基

2.

$$(1) \begin{pmatrix} -1 & 1 \\ -2 & 3 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} 2 & 3 \\ 2 & 1 \end{pmatrix} = \begin{pmatrix} 0 & -2 \\ 2 & -3 \\ 2 & 5 \end{pmatrix}$$

$$(2) \begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix} \begin{pmatrix} \cos \beta & -\sin \beta \\ \sin \beta & \cos \beta \end{pmatrix} = \begin{pmatrix} \cos \alpha \cos \beta - \sin \alpha \sin \beta & -\cos \alpha \sin \beta - \sin \alpha \cos \beta \\ \sin \alpha \cos \beta + \cos \alpha \sin \beta & -\sin \beta \sin \alpha + \cos \alpha \cos \beta \end{pmatrix}$$

$$= \begin{pmatrix} \cos(\alpha + \beta) & -\sin(\alpha + \beta) \\ \sin(\alpha + \beta) & \cos(\alpha + \beta) \end{pmatrix}$$

$$(3) \begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix}^n = \begin{pmatrix} \cos n\alpha & -\sin n\alpha \\ \sin n\alpha & \cos n\alpha \end{pmatrix} \quad \text{用归纳法证}$$

$n=2$ 时, (2)中令  $\alpha = \beta$  即可.

假设  $n-1$  成立.

$$\begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix}^n = \begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix}^{n-1} \begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix}$$

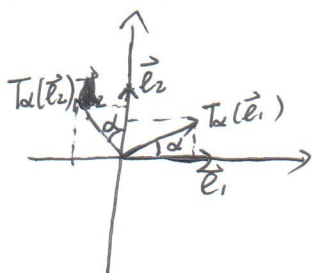
$$\stackrel{\text{归纳假设}}{=} \begin{pmatrix} \cos(n-1)\alpha & -\sin(n-1)\alpha \\ \sin(n-1)\alpha & \cos(n-1)\alpha \end{pmatrix} \begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix}$$

$$= \begin{pmatrix} \cos n\alpha & -\sin n\alpha \\ \sin n\alpha & \cos n\alpha \end{pmatrix}$$

注: 设  $T_\alpha: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  是旋转, 即

$$A = (T_\alpha(\vec{e}_1), T_\alpha(\vec{e}_2)) = \begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix} \quad \text{i.e. } \begin{aligned} T_\alpha(\vec{e}_1) &= \cos \alpha \vec{e}_1 + \sin \alpha \vec{e}_2 \\ T_\alpha(\vec{e}_2) &= -\sin \alpha \vec{e}_1 + \cos \alpha \vec{e}_2 \end{aligned}$$

$\vec{e}_1, \vec{e}_2$  逆时针旋转  $\alpha$  角



$$T_\alpha^k: \mathbb{R}^2 \rightarrow \mathbb{R}^2, \quad k \geq 1.$$

$$\vec{x} \rightarrow A^k \vec{x}$$

$\vec{e}_1, \vec{e}_2$  逆时针旋转  $k\alpha$  角.

3. (1) 令  $A = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{pmatrix} (y_1, y_2, \dots, y_n)$ . 求矩阵  $A$  的秩.

解: 令  $A_1 = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{pmatrix}$   $A_2 = (y_1, y_2, \dots, y_n)$   $A = A_1 A_2$ .

若  $A_1 = 0$  或  $A_2 = 0$ , 则  $A = 0$ ,  $\Rightarrow \text{rank}(A) = 0$ .

法一: 若  $A_1 \neq 0$  且  $A_2 \neq 0$ . 则  $\text{rank}(A_1) = \text{rank}(A_2) = 1$ .

$$\therefore \text{rank}(A_1) + \text{rank}(A_2) - 1 \leq \text{rank}(A) \leq \min(\text{rank}(A_1), \text{rank}(A_2))$$

$$\text{i.e. } 1 \leq \text{rank}(A) \leq 1 \Rightarrow \text{rank}(A) = 1.$$

法二: 若  $A_1 \neq 0, A_2 \neq 0$ .

$$A = \begin{pmatrix} x_1 y_1 & x_1 y_2 & \dots & x_1 y_n \\ x_2 y_1 & x_2 y_2 & \dots & x_2 y_n \\ \vdots & \vdots & \ddots & \vdots \\ x_m y_1 & x_m y_2 & \dots & x_m y_n \end{pmatrix} \xrightarrow[r_i - \frac{x_i}{x_1} r_1]{\text{不妨设 } x_1 \neq 0} \begin{pmatrix} x_1 y_1 & x_1 y_2 & \dots & x_1 y_n \\ 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 \end{pmatrix}$$

$\therefore x_1 \neq 0, y_1, \dots, y_n$  不全为 0.  $\therefore$  第一行不是零向量  $\Rightarrow \text{rank}(A) = 1$ .

$$\text{综上: } \text{rank}(A) = \begin{cases} 0, & A_1 = 0 \text{ 或 } A_2 = 0. \\ 1, & A_1 \neq 0, A_2 \neq 0. \end{cases}$$

(2)  $A \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(A) = 1$ . 则  $A$  可表示成  $\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{pmatrix} (y_1, y_2, \dots, y_n)$ .  $x_i, y_j$  不全为 0.

方法一: 用行向量看.

设  $A = \begin{pmatrix} \vec{A}_1 \\ \vec{A}_2 \\ \vdots \\ \vec{A}_m \end{pmatrix}$ .  $\therefore \text{rank}(A) = 1 \therefore A$  任意非零行向量都是  $\text{rk}(A)$  的一组基. 不妨设  $\vec{A}_1$  是一组基 i.e.  $\text{rk}(A) = \langle \vec{A}_1 \rangle$ .

$$\text{则 } \exists! k_2, \dots, k_m \in \mathbb{R} \text{ s.t. } \vec{A}_i = k_i \vec{A}_1, \quad i = 2, \dots, m$$

$$\Rightarrow A = \begin{pmatrix} \vec{A}_1 \\ k_2 \vec{A}_1 \\ \vdots \\ k_m \vec{A}_1 \end{pmatrix} = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ k_2 a_{11} & \dots & k_2 a_{1n} \\ \vdots & \ddots & \vdots \\ k_m a_{11} & \dots & k_m a_{1n} \end{pmatrix} = \begin{pmatrix} 1 \\ k_2 \\ \vdots \\ k_m \end{pmatrix} (a_{11} \ a_{12} \ \dots \ a_{1n})$$

方法二：用列向量看

$A = (\vec{A}^{(1)}, \dots, \vec{A}^{(n)})$  不妨设  $V_c(A) = \langle \vec{A}^{(1)} \rangle$  则  $\exists l_2, \dots, l_n \in \mathbb{R}$

$$\vec{A}^{(j)} = l_j \vec{A}^{(1)}, j=2, \dots, n.$$

$$A = \begin{pmatrix} a_{11} & l_2 a_{11} & \dots & l_n a_{11} \\ a_{21} & l_2 a_{21} & \dots & l_n a_{21} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & l_2 a_{m1} & \dots & l_n a_{m1} \end{pmatrix} = \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} (1 \cdot l_2 \dots l_n)$$

注：由(1)(2)可知  $\text{rank}(A)=1 \Leftrightarrow A = \begin{pmatrix} x_1 \\ \vdots \\ x_m \end{pmatrix} (y_1, \dots, y_n)$   $x_i$  不全为0  
 $y_j$  不全为0

(3)  $A \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(A)=r$ .  $A$  可以写成  $r$  个秩为1的矩阵的和  
 但不能写成  $r-1$  个秩为1的矩阵的和

方法一：用行向量看

$$\text{设 } A = \begin{pmatrix} \vec{A}_1 \\ \vdots \\ \vec{A}_r \\ \vec{A}_{r+1} \\ \vdots \\ \vec{A}_m \end{pmatrix}$$

不妨设  $\vec{A}_1, \dots, \vec{A}_r$  是  $V_r(A)$  的一组基且

$$\vec{A}_i = \alpha_{i1} \vec{A}_1 + \dots + \alpha_{ir} \vec{A}_r, i=r+1, \dots, m, \alpha_{ij} \in \mathbb{R}$$

$$\Rightarrow A = \begin{pmatrix} \vec{A}_1 \\ \vdots \\ \vec{A}_r \\ \alpha_{r+1,1} \vec{A}_1 + \dots + \alpha_{r+1,r} \vec{A}_r \\ \vdots \\ \alpha_{m,1} \vec{A}_1 + \dots + \alpha_{m,r} \vec{A}_r \end{pmatrix} = \begin{pmatrix} \vec{A}_1 \\ \vec{0} \\ \vec{0} \\ \alpha_{m,1} \vec{A}_1 \\ \vdots \\ \alpha_{m,1} \vec{A}_1 \end{pmatrix} + \begin{pmatrix} \vec{0} \\ \vec{A}_2 \\ \vec{0} \\ \alpha_{m,2} \vec{A}_2 \\ \vdots \\ \alpha_{m,2} \vec{A}_2 \end{pmatrix} + \dots + \begin{pmatrix} \vec{0} \\ \vec{0} \\ \vec{A}_r \\ \alpha_{m,r} \vec{A}_r \\ \vdots \\ \alpha_{m,r} \vec{A}_r \end{pmatrix}$$

$\parallel$   $B_1$                        $\parallel$   $B_2$                        $\parallel$   $B_r$

i.e.  $A = B_1 + B_2 + \dots + B_r$  且  $\text{rank}(B_i)=1, i=1, 2, \dots, r$

[注： $\because \vec{A}_1, \dots, \vec{A}_r$  是基  $\therefore \vec{A}_1, \dots, \vec{A}_r$  都是非0向量]

方法二：用列向量看

类似设  $\vec{A}^{(1)}, \dots, \vec{A}^{(r)}$  是  $V_c(A)$  的一组基,  $\vec{A}^{(j)} = \beta_{j,1} \vec{A}^{(1)} + \dots + \beta_{j,r} \vec{A}^{(r)}, j=r+1, \dots, n, \beta_{j,i} \in \mathbb{R}$

$$A = (\vec{A}^{(1)}, \dots, \vec{A}^{(r)}, \beta_{r+1,1} \vec{A}^{(1)} + \dots + \beta_{r+1,r} \vec{A}^{(r)}, \dots, \beta_{n,1} \vec{A}^{(1)} + \dots + \beta_{n,r} \vec{A}^{(r)})$$

$$= (\vec{A}^{(1)}, \vec{0}, \dots, \vec{0}, \beta_{r+1,1} \vec{A}^{(1)}, \dots, \beta_{r+1,r} \vec{A}^{(r)}) + \dots + (\vec{0}, \vec{0}, \dots, \vec{A}^{(r)}, \beta_{n,r} \vec{A}^{(r)}, \dots, \beta_{n,r} \vec{A}^{(r)})$$

如果  $A = B_1 + \dots + B_{r-1}$ ,  $\text{rank}(B_i) = 1$ .

$$\square) \text{rank}(A) \leq \text{rank}(B_1) + \dots + \text{rank}(B_{r-1}) = r-1.$$

$\therefore \text{rank}(A) = r$  矛盾.

注: 例 6.13.  $A, B \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(A+B) \leq \text{rank}(A) + \text{rank}(B)$ .

4.  $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^n$  线性映射. 证:

$$(1) \ker(\varphi) \subset \ker(\varphi^2) \subset \dots$$

$$(2) \text{im}(\varphi) \supset \text{im}(\varphi^2) \supset \dots$$

$$(3) \exists k \in \mathbb{Z}^+ \text{ s.t. } \ker(\varphi^k) = \ker(\varphi^{k+1}) = \dots \\ \text{im}(\varphi^k) = \text{im}(\varphi^{k+1}) = \dots$$

证: (1). 证  $\ker(\varphi^i) \subset \ker(\varphi^{i+1}) \quad \forall i \geq 1$ .

$$\forall \vec{x} \in \ker(\varphi^i), \text{有 } \varphi^i(\vec{x}) = \vec{0}, \square) \varphi^{i+1}(\vec{x}) = \varphi(\varphi^i(\vec{x})) = \varphi(\vec{0}) = \vec{0}.$$

$$\Rightarrow \vec{x} \in \ker(\varphi^{i+1}) \quad \text{i.e. } \ker(\varphi^i) \subset \ker(\varphi^{i+1}).$$

(2). 证  $\text{im}(\varphi^i) \supset \text{im}(\varphi^{i+1}) \quad \forall i \geq 1$ .

$$\forall \vec{y} \in \text{im}(\varphi^{i+1}), \exists \vec{x} \in \mathbb{R}^n \text{ s.t. } \varphi^{i+1}(\vec{x}) = \vec{y}, \text{ i.e. } \varphi^i(\varphi(\vec{x})) = \vec{y},$$

$$\text{其中 } \varphi(\vec{x}) \in \mathbb{R}^n. \Rightarrow \vec{y} \in \text{im}(\varphi^i), \text{ i.e. } \text{im}(\varphi^{i+1}) \subset \text{im}(\varphi^i).$$

(3). 分析:  $\ker(\varphi^i)$  是  $\mathbb{R}^n$  的子空间,  $\dim(\ker(\varphi^i)) \leq n$ .

$$\text{im}(\varphi^i) \supset \{0\}, \therefore \dim(\text{im}(\varphi^i)) \geq 0.$$

注: (2) 方法 = : 证  $\text{im}(\varphi^i) \supset \text{im}(\varphi^{i+1}) \quad \forall i \geq 1$ .

$$\text{im}(\varphi^{i+1}) = \varphi^i(\text{im}(\varphi)) \subset \varphi^i(\mathbb{R}^n) = \text{im}(\varphi^i)$$

(反证法). 若不存在  $k \in \mathbb{Z}^+$  s.t.  $\ker(\varphi^k) = \ker(\varphi^{k+1}) = \dots$

则可以得到一个严格包含的序列:

$$\ker(\varphi^{i_0}) \subsetneq \ker(\varphi^{i_1}) \subsetneq \dots \subsetneq \ker(\varphi^{i_n}) \subsetneq \ker(\varphi^{i_{n+1}}) \subsetneq \dots$$

其中  $\dim(\ker(\varphi^{i_s})) \geq s$ .  $s=0, 1, 2, \dots$

则  $\dim(\ker(\varphi^{i_{n+1}})) \geq n+1$ , 与  $\ker(\varphi^{i_{n+1}}) \subset \mathbb{R}^n$  矛盾.

$\therefore \exists k \in \mathbb{Z}^+$ , s.t.  $\ker(\varphi^k) = \ker(\varphi^{k+1}) = \dots$

由对偶定理可知  $\dim(\ker(\varphi^i)) + \dim(\text{im}(\varphi^i)) = n$ .  $\forall i \geq 1$

$$\Rightarrow \dim(\text{im}(\varphi^k)) = \dim(\text{im}(\varphi^{k+1})) = \dots = n - \dim(\ker(\varphi^k))$$

又  $\because \text{im}(\varphi^k) \supset \text{im}(\varphi^{k+1}) \supset \dots$

$$\therefore \text{im}(\varphi^k) = \text{im}(\varphi^{k+1}) = \dots$$

注: (1) 若  $\exists k \in \mathbb{Z}^+$ , s.t.  $\ker(\varphi^k) = \ker(\varphi^{k+1})$ , 则

$$\ker(\varphi^k) = \ker(\varphi^{k+1}) = \ker(\varphi^{k+2}) = \dots$$

$$\text{im}(\varphi^k) = \text{im}(\varphi^{k+1}) = \text{im}(\varphi^{k+2}) = \dots$$

证: (1) 先证  $\ker(\varphi^{k+1}) = \ker(\varphi^{k+2})$ .

已知  $\ker(\varphi^{k+1}) \subset \ker(\varphi^{k+2})$ . 下证  $\ker(\varphi^{k+2}) \subset \ker(\varphi^{k+1})$

$$\forall \vec{x} \in \ker(\varphi^{k+2}), \varphi^{k+2}(\vec{x}) = \varphi^{k+1}(\varphi(\vec{x})) = \vec{0} \Rightarrow \varphi(\vec{x}) \in \ker(\varphi^{k+1}) = \ker(\varphi^k)$$

$$\Rightarrow \varphi^k(\varphi(\vec{x})) = \varphi^{k+1}(\vec{x}) = \vec{0} \Rightarrow \vec{x} \in \ker(\varphi^{k+1})$$

$$\therefore \ker(\varphi^{k+1}) = \ker(\varphi^{k+2})$$

类似可证  $\ker(\varphi^{k+2}) = \ker(\varphi^{k+3}) = \dots$

因此  $\ker(\varphi^i)$  的关系如下:

$$\ker(\varphi) \subsetneq \ker(\varphi^2) \subsetneq \dots \subsetneq \ker(\varphi^k) = \ker(\varphi^{k+1}) = \dots$$

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[一旦出现  $\ker(\varphi^k) = \ker(\varphi^{k+1})$ , 则从  $\ker(\varphi^k)$  开始稳定 i.e.  $\ker(\varphi^k) = \ker(\varphi^{k+1}) = \dots$ ]

(2) 若  $\exists k \in \mathbb{Z}^+$  s.t.  $\text{im}(\varphi^k) = \text{im}(\varphi^{k+1})$   
 则  
 $\text{im}(\varphi^k) = \text{im}(\varphi^{k+1}) = \dots$   
 $\ker(\varphi^k) = \ker(\varphi^{k+1}) = \dots$   
 证:  $\text{im}(\varphi^k) = \varphi^k(\mathbb{R}^n)$   
 $\text{im}(\varphi^{k+1}) = \varphi^{k+1}(\mathbb{R}^n)$   
 $\varphi^k(\mathbb{R}^n) = \varphi^{k+1}(\mathbb{R}^n) \Rightarrow \varphi^{k+1}(\mathbb{R}^n) = \varphi^{k+2}(\mathbb{R}^n)$   
 i.e.  $\text{im}(\varphi^{k+1}) = \text{im}(\varphi^{k+2})$

因此  $\dim(\ker(\varphi)) < \dim(\ker(\varphi^2)) < \dots < \dim(\ker(\varphi^k)) = \dim(\ker(\varphi^{k+1})) = \dots$

由对偶定理可知

$$\dim(\operatorname{im}(\varphi)) > \dim(\operatorname{im}(\varphi^2)) > \dots > \dim(\operatorname{im}(\varphi^k)) = \dim(\operatorname{im}(\varphi^{k+1})) = \dots$$

$$\text{又} \because \operatorname{im}(\varphi) \supseteq \operatorname{im}(\varphi^2) \supseteq \dots$$

由维数的性质可知

$$\operatorname{im}(\varphi) \supsetneq \operatorname{im}(\varphi^2) \supsetneq \dots \supsetneq \operatorname{im}(\varphi^k) = \operatorname{im}(\varphi^{k+1}) = \dots$$

例:  $\varphi: \mathbb{R}^5 \xrightarrow{\varphi} \mathbb{R}^5 \xrightarrow{\varphi} \mathbb{R}^5 \xrightarrow{\varphi} \mathbb{R}^5 \xrightarrow{\varphi} \mathbb{R}^5$

$\vec{e}_1$	$\mapsto$	$\vec{e}_1$	$\mapsto$	$\vec{e}_1$	$\mapsto$	$\vec{e}_1$	$\mapsto$	$\vec{e}_1$
$\vec{e}_2$	$\mapsto$	$\vec{e}_2$	$\mapsto$	$\vec{e}_2$	$\mapsto$	$\vec{e}_2$	$\mapsto$	$\vec{e}_2$
$\vec{e}_3$	$\mapsto$	$\vec{0}$	$\mapsto$	$\vec{0}$	$\mapsto$	$\vec{0}$	$\mapsto$	$\vec{0}$
$\vec{e}_4$	$\mapsto$	$\vec{e}_3$	$\mapsto$	$\vec{0}$	$\mapsto$	$\vec{0}$	$\mapsto$	$\vec{0}$
$\vec{e}_5$	$\mapsto$	$\vec{e}_4$	$\mapsto$	$\vec{e}_3$	$\mapsto$	$\vec{0}$	$\mapsto$	$\vec{0}$
		$\vec{e}_5$	$\mapsto$	$\vec{e}_4$	$\mapsto$	$\vec{e}_3$	$\mapsto$	$\vec{0}$
				$\vec{e}_5$	$\mapsto$	$\vec{e}_4$	$\mapsto$	$\vec{e}_3$
						$\vec{e}_5$	$\mapsto$	$\vec{e}_4$

$$\ker \varphi = \langle \vec{e}_3 \rangle, \ker \varphi^2 = \langle \vec{e}_3, \vec{e}_4 \rangle, \ker \varphi^3 = \langle \vec{e}_3, \vec{e}_4, \vec{e}_5 \rangle, \ker \varphi^4 = \langle \vec{e}_3, \vec{e}_4, \vec{e}_5 \rangle, \dots$$

$$\operatorname{im} \varphi = \langle \vec{e}_1, \vec{e}_2, \vec{e}_3, \vec{e}_4 \rangle, \operatorname{im} \varphi^2 = \langle \vec{e}_1, \vec{e}_2, \vec{e}_3 \rangle, \operatorname{im} \varphi^3 = \langle \vec{e}_1, \vec{e}_2 \rangle, \operatorname{im} \varphi^4 = \langle \vec{e}_1, \vec{e}_2 \rangle, \dots$$

$$\Rightarrow \ker \varphi \subsetneq \ker \varphi^2 \subsetneq \ker \varphi^3 = \ker \varphi^4 = \dots$$

$$\operatorname{im} \varphi \supsetneq \operatorname{im} \varphi^2 \supsetneq \operatorname{im} \varphi^3 = \operatorname{im} \varphi^4 = \dots$$