

第四章 n 维向量空间

4.2 向量组的线性相关性

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一、向量组的线性组合

1. 向量组与矩阵

向量组 同维数的向量所组成的集合.

$A = (a_{ij})_{m \times n}$ 有 n 个 m 维列向量

$$A = \begin{pmatrix} \alpha_1 & \alpha_2 & \cdots & \alpha_j & \cdots & \alpha_n \\ \mathbf{a}_{11} & \mathbf{a}_{12} & \cdots & \mathbf{a}_{1j} & \cdots & \mathbf{a}_{1n} \\ \mathbf{a}_{21} & \mathbf{a}_{22} & \cdots & \mathbf{a}_{2j} & \cdots & \mathbf{a}_{2n} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \mathbf{a}_{m1} & \mathbf{a}_{m2} & \cdots & \mathbf{a}_{mj} & \cdots & \mathbf{a}_{mn} \end{pmatrix}$$

$\alpha_1, \alpha_2, \cdots, \alpha_n$ 称为矩阵 A 的 列向量组

对称地, 矩阵 $A = (a_{ij})_{m \times n}$ 有 m 个 n 维行向量

$$A = \begin{pmatrix} \boxed{a_{11} \quad a_{12} \quad \cdots \quad a_{1n}} & \beta_1 \\ \boxed{a_{21} \quad a_{22} \quad \cdots \quad a_{2n}} & \beta_2 \\ \vdots & \\ \boxed{a_{i1} \quad a_{i2} \quad \cdots \quad a_{in}} & \beta_i \\ \vdots & \\ \boxed{a_{m1} \quad a_{m2} \quad \cdots \quad a_{mn}} & \beta_m \end{pmatrix}$$

$\beta_1, \beta_2, \dots, \beta_m$ 称为矩阵 A 的行向量组.

反之, 给定行(列)向量组, 也可构造矩阵 A 使得:

A 的行(列)向量组恰为给定向量组.

2.线性组合、线性表出的概念

设给定向量 β , 向量组 $\alpha_1, \dots, \alpha_m$, 若存在数 k_1, \dots, k_m 使得

$$\beta = k_1\alpha_1 + k_2\alpha_2 + \dots + k_m\alpha_m,$$

则称向量 β 为向量组 $\alpha_1, \alpha_2, \dots, \alpha_m$ 的线性组合,

也称 β 可由 $\alpha_1, \alpha_2, \dots, \alpha_m$ 线性表出.

问题: (1) 如何判断向量 β 可否由某个向量组线性表出?

(2) 如何计算向量 β 被某向量组线性表出的关系式?

例1.

(1) 零向量是任一向量组的线性组合.

$$\mathbf{0} = \mathbf{0} \alpha_1 + \mathbf{0} \alpha_2 + \cdots + \mathbf{0} \alpha_m.$$

(2) 向量组 $\alpha_1, \alpha_2, \dots, \alpha_m$ 中任一向量
都可由该向量组自身线性表出.

$$\alpha_i = \mathbf{0} \alpha_1 + \cdots + \mathbf{0} \alpha_{i-1} + \mathbf{1} \alpha_i + \mathbf{0} \alpha_{i+1} + \cdots + \mathbf{0} \alpha_m.$$

(3) 3维几何空间中任一向量可以由 $\vec{i}, \vec{j}, \vec{k}$ 线性表出.

$$(a, b, c) = a \vec{i} + b \vec{j} + c \vec{k}$$

$L(\alpha_1, \dots, \alpha_m)$: $\alpha_1, \dots, \alpha_m$ 线性组合的全体.

$$L(\alpha_1, \dots, \alpha_m) = \{ \textcolor{red}{k}_1 \alpha_1 + \dots + \textcolor{red}{k}_m \alpha_m \mid k_1, \dots, k_m \in \mathbf{R} \}$$

(1) $\alpha_1, \dots, \alpha_m \in L(\alpha_1, \dots, \alpha_m)$;

(2) $L(\alpha_1, \dots, \alpha_m)$ 是 \mathbf{R}^n 的子空间; $L(\alpha_1, \dots, \alpha_m) \neq \emptyset$

$$\forall \textcolor{blue}{k}_1 \alpha_1 + \dots + \textcolor{blue}{k}_m \alpha_m, \textcolor{red}{l}_1 \alpha_1 + \dots + \textcolor{red}{l}_m \alpha_m \in L(\alpha_1, \dots, \alpha_m), \textcolor{red}{c} \in \mathbf{R}$$

$$\Rightarrow (\textcolor{blue}{k}_1 + \textcolor{red}{l}_1) \alpha_1 + \dots + (\textcolor{blue}{k}_m + \textcolor{red}{l}_m) \alpha_m \in L(\alpha_1, \dots, \alpha_m),$$

$$\textcolor{red}{c} \textcolor{blue}{k}_1 \alpha_1 + \dots + \textcolor{red}{c} \textcolor{blue}{k}_m \alpha_m \in L(\alpha_1, \dots, \alpha_m).$$

$L(\alpha_1, \dots, \alpha_m)$ 称为 $\alpha_1, \dots, \alpha_m$ 生成的子空间.

(4) $\mathbf{R}^n = L(\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n)$, 其中

$$\varepsilon_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \varepsilon_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \dots, \varepsilon_n = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}.$$

即, 任一 n 维向量均可由 $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n$ 线性表出:

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = x_1 \varepsilon_1 + x_2 \varepsilon_2 + \dots + x_n \varepsilon_n.$$

3. 线性表出的充要条件和计算

定理1. 设 $A=(\alpha_1, \dots, \alpha_n)$, 则如下条件等价:

(1) $b \in L(\alpha_1, \dots, \alpha_n)$ (2) $AX=b$ 有解; (3) $R(A)=R(\overline{A})$.

证: (1) \Leftrightarrow (2) $b \in L(\alpha_1, \dots, \alpha_n)$



有数 x_1, \dots, x_n 使得 $x_1\alpha_1 + \dots + x_n\alpha_n = b$



有数 x_1, x_2, \dots, x_n 使得 $(\alpha_1, \dots, \alpha_n) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = b,$



$AX=b$ 有解

定理1. 设 $A=(a_1, \dots, a_n)$, 则如下条件等价:

(1) $b \in L(a_1, \dots, a_n)$ (2) $AX=b$ 有解; (3) $R(A)=R(\overline{A})$.

(2) \Leftrightarrow (3)

$$\begin{array}{l} \text{设 } R(A)=r, \\ \overline{A}=(A|b) \xrightarrow{\text{行初等变换}} \left(\begin{array}{cccc|c} c_{11} & \cdots & c_{1s} & \cdots & c_{1n} & d_1 \\ & \ddots & \vdots & & \vdots & \vdots \\ & & c_{rs} & \cdots & c_{rn} & d_r \\ & & & & \mathbf{0} & d_{r+1} \\ & & & & \mathbf{0} & \vdots \\ & & & & \mathbf{0} & d_{r+1} \end{array} \right) = (B|d). \end{array}$$

$AX=b$ 与 $BX=d$ 同解, 所以

$$\begin{aligned} AX=b \text{ 有解} &\Leftrightarrow d_{r+1}=\mathbf{0} \Leftrightarrow R(B|d)=R(B)=r \\ &\Leftrightarrow R(\overline{A})=R(A) \end{aligned}$$

例2. 将 $b = \begin{pmatrix} 1 \\ 0 \\ -4 \end{pmatrix}$ 用 $\alpha_1 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$, $\alpha_2 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$, $\alpha_3 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ 线性表出.

解:

$$(A|b) = (\alpha_1, \alpha_2, \alpha_3 | b) = \left(\begin{array}{ccc|c} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & -4 \end{array} \right) \rightarrow \left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & -4 \end{array} \right)$$
$$\rightarrow \left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & -1 & -4 \end{array} \right) \rightarrow \left(\begin{array}{ccc|c} 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & -2 & -5 \end{array} \right) \rightarrow \left(\begin{array}{ccc|c} 1 & 0 & 0 & -\frac{5}{2} \\ 0 & 1 & 0 & -\frac{3}{2} \\ 0 & 0 & 1 & \frac{5}{2} \end{array} \right)$$

$$\Rightarrow b = -\frac{5}{2} \alpha_1 - \frac{3}{2} \alpha_2 + \frac{5}{2} \alpha_3$$