

第六次作业.

已知 $\vec{v} \in V_1 \cap V_2$. 问?

$$V = \bigoplus_{i=1}^n V_i$$

1. (证) $\text{PSS. 1)} \forall \vec{v} \in V \exists! (\vec{v}_1, \vec{v}_2) \in V_1 \times V_2$ s.t. $\vec{v} = \vec{v}_1 + \vec{v}_2$

$$V_1 \cap \sum_{j \neq i} V_j = \{0\}$$

$\forall i \in \{1, \dots, n\}$ 表法唯一 \rightarrow $\vec{v}_i = \vec{v}$

2. $\forall \vec{v} \in V_1 \oplus V_2$ 则 $\exists \vec{v}_1 \in V_1, \vec{v}_2 \in V_2$ s.t. $\vec{v} = \vec{v}_1 + \vec{v}_2$

$$V = V_1 \oplus V_2$$

且假设 $\exists \vec{v}'_1 \in V_1, \vec{v}'_2 \in V_2$ s.t. $\vec{v} = \vec{v}'_1 + \vec{v}'_2$ 则 $\vec{v}_1 - \vec{v}'_1 = \vec{v}_2 - \vec{v}'_2 \in V_1 \cap V_2 = \{0\}$

$$\Rightarrow \vec{v}_1 = \vec{v}'_1, \vec{v}_2 = \vec{v}'_2 \text{ 即 } \exists! (\vec{v}_1, \vec{v}_2) \in V_1 \times V_2 \text{ s.t. } \vec{v} = \vec{v}_1 + \vec{v}_2 \Rightarrow \vec{v} \in V$$

$\forall \vec{v} \in V \because \exists \vec{v}_1 \in V_1, \vec{v}_2 \in V_2$ s.t. $\vec{v} = \vec{v}_1 + \vec{v}_2 \Rightarrow \vec{v} \in V_1 + V_2$

假设 $\exists \vec{u} \in V_1 \cap V_2$ 且 $\vec{u} \neq \vec{0}$ 则 $\vec{v} = (\vec{v}_1 + \vec{u}) + (\vec{v}_2 - \vec{u})$

其中 $\vec{v}_1 + \vec{u} \in V_1, \vec{v}_2 - \vec{u} \in V_2 \Rightarrow$ 唯一表示矛盾 $\therefore V_1 \cap V_2 = \{0\} \Rightarrow V_1 + V_2 = V_1 \oplus V_2$

综上 $V = V_1 \oplus V_2$.

2. 设 $V_1, V_2 \subseteq \mathbb{R}^n$ 为子空间. $V_1 + V_2$ 是直和 $\Leftrightarrow \dim(V_1 + V_2) = \dim V_1 + \dim V_2$.

证 如果 $V_1 + V_2$ 是直和 则 $V_1 \cap V_2 = \{0\}$ 由维数公式:

$$\dim(V_1 + V_2) = \dim V_1 + \dim V_2 - \dim(V_1 \cap V_2) \Rightarrow \dim(V_1 + V_2) = \dim V_1 + \dim V_2.$$

由反证法 如果 $\dim(V_1 \cap V_2) = 0 \Rightarrow V_1 \cap V_2 = \{0\} \Rightarrow V_1 + V_2$ 是直和.

3. (证) $\text{PSS. 4)} X_1 = \begin{pmatrix} 1 \\ 3 \end{pmatrix}, X_2 = \begin{pmatrix} 3 \\ 1 \end{pmatrix}$ 线性无关. $V = \langle X_1, X_2 \rangle$ 证明 $X = \begin{pmatrix} -5 \\ 9 \end{pmatrix} \in V$. 求解.

设 $\alpha_1, \alpha_2 \in \mathbb{R}$ s.t. $\alpha_1 X_1 + \alpha_2 X_2 = \vec{0} \Rightarrow \begin{cases} \alpha_1 + 3\alpha_2 = 0 \\ 2\alpha_1 + 2\alpha_2 = 0 \\ 3\alpha_1 + \alpha_2 = 0 \end{cases} \Rightarrow \begin{cases} \alpha_1 = 0 \\ \alpha_2 = 0 \end{cases} \quad (\begin{pmatrix} 1 & 3 \\ 2 & 2 \\ 3 & 1 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 3 & 0 \\ 0 & 4 & 0 \\ 0 & -8 & 0 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 3 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix})$

$\therefore X_1, X_2$ 线性无关. $V = \langle X_1, X_2 \rangle = \{ \alpha_1 X_1 + \alpha_2 X_2 \mid \alpha_1, \alpha_2 \in \mathbb{R} \}$.

$\alpha_1 X_1 + \alpha_2 X_2 = \begin{pmatrix} -5 \\ 9 \end{pmatrix} \Rightarrow \begin{cases} \alpha_1 + 3\alpha_2 = -5 \\ 2\alpha_1 + 2\alpha_2 = 2 \\ 3\alpha_1 + \alpha_2 = 9 \end{cases} \quad (\begin{pmatrix} 1 & 3 & -5 \\ 2 & 2 & 2 \\ 3 & 1 & 9 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 3 & -5 \\ 0 & -4 & 12 \\ 0 & -8 & 24 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 3 & -5 \\ 0 & 1 & -3 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 0 & -4 \\ 0 & 1 & -3 \\ 0 & 0 & 0 \end{pmatrix})$

$$\begin{cases} \alpha_1 = 4 \\ \alpha_2 = -3 \end{cases} \text{ 即 } 4 \begin{pmatrix} 1 \\ 3 \end{pmatrix} + (-3) \begin{pmatrix} 3 \\ 1 \end{pmatrix} = \begin{pmatrix} -5 \\ 9 \end{pmatrix}$$

$$U = \left\langle \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\rangle = \{ \alpha_1 \begin{pmatrix} 1 \\ 3 \\ 1 \end{pmatrix} + \alpha_2 \begin{pmatrix} 3 \\ 1 \\ 1 \end{pmatrix} \mid \alpha_1, \alpha_2 \in \mathbb{R} \} \Rightarrow \alpha_1 = \alpha_2 = \alpha_3 = 0$$

X_1, X_2, X_3 线性无关 $\therefore V \oplus U = \mathbb{R}^3$

直和. 即 $V \oplus U$ 的基张成 \mathbb{R}^3 . 不是 X_1, X_2 的基.

$$\begin{pmatrix} 1 & 2 & 3 \\ 3 & 2 & 1 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & -8 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & -2 \end{pmatrix} \xrightarrow{\sim} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 2 \end{pmatrix}$$

4. (TO Pg. 5) $x_1, x_2 \dots x_n \in \mathbb{R}^n$. $\langle x_1, \dots x_n \rangle = \mathbb{R}^n \Leftrightarrow x_1, \dots x_n$ 线性无关.

\Leftarrow (反证) 假设 $\langle x_1, \dots x_n \rangle = \mathbb{R}^n$. 且 $x_1, \dots x_n$ 线性相关. 则 $\exists \{x_{i_1}, \dots x_{i_d}\} \subseteq \{x_1, \dots x_n\}$ ($d < n$) 构成 \mathbb{R}^n 的一组基. $\therefore \dim(\mathbb{R}^n) = d < n$ 矛盾 $\therefore x_1, \dots x_n$ 线性无关.

\Leftarrow 设 $x_1, \dots x_n$ 线性无关. $\therefore \mathbb{R}^n$ 中任意 $n+1$ 个向量线性相关. $\therefore \forall X \in \mathbb{R}^n$.

$\exists \alpha, \alpha_1, \dots \alpha_n \in \mathbb{R}^n$ s.t. $\alpha X + \sum_{i=1}^n \alpha_i x_i = \vec{0}$ 且如果 $\alpha=0 \Rightarrow \alpha_i=0$ ($i=1, 2, \dots, n$) 为真.

$\because \alpha \neq 0 \therefore X = \sum_{i=1}^n (\frac{\alpha_i}{\alpha}) x_i \in \langle x_1, \dots x_n \rangle \Rightarrow \mathbb{R}^n \subseteq \langle x_1, \dots x_n \rangle$

5. 求矩阵的秩: $\boxed{\text{若 } x_i \in \mathbb{R}^n \Rightarrow \langle x_1, \dots, x_n \rangle \subseteq \mathbb{R}^n \text{ 且 } \dim(\langle x_1, \dots, x_n \rangle) = n = \dim(\mathbb{R}^n) \therefore \langle x_1, \dots, x_n \rangle = \mathbb{R}^n}$

$$(1) A_1 = \begin{pmatrix} 8 & 2 & 2 & -1 & 1 \\ 1 & 7 & 4 & -2 & 5 \\ -2 & 4 & 2 & -1 & 3 \end{pmatrix} \quad (2) A_2 = \begin{pmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 \end{pmatrix} \quad (3) A_3 = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 0 & 1 \end{pmatrix}$$

$\xrightarrow{\text{列交换}} \quad \xrightarrow{\text{列交换}} \quad \xrightarrow{\text{列交换}}$

$$\therefore (1) A_1 \rightarrow \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -39 & -3 & -6 & 3 & 5 \\ -26 & -2 & -4 & 2 & 3 \end{pmatrix} \rightarrow \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 3 & 5 \\ 0 & 0 & 0 & 2 & 3 \end{pmatrix} \Rightarrow \text{rank } A_1 = 2.$$

$$(2) A_2 \rightarrow \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & -1 & 0 & 1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix} \Rightarrow \text{rank } A_2 = 3.$$

$$(3) A_3 \rightarrow \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & -1 & 0 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \Rightarrow \text{rank } A_3 = 5.$$

6. 设 $a_0, \dots, a_{n-1}, \lambda, \mu \in \mathbb{R}$ 求下列矩阵的秩.

$$A_1 = \begin{pmatrix} 0 & 0 & \cdots & 0 & a_0 \\ 1 & 0 & \cdots & 0 & a_1 \\ 0 & 1 & \cdots & 0 & a_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & 1 & \cdots & 1 & a_{n-1} \\ 0 & 0 & \cdots & 1 & a_{n-1} \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & \cdots & 0 & a_1 \\ 0 & 1 & \cdots & 0 & a_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & a_{n-1} \\ 0 & 0 & \cdots & 0 & a_0 \end{pmatrix}$$

如果 $a_0 = 0$ 则 $\text{rank } A_1 = n-1$
 $a_0 \neq 0$ 则 $\text{rank } A_1 = n$.

看阶梯形的非零行求矩阵的秩.

列或行变换.

$$A_2 = \begin{pmatrix} \lambda & \mu & \cdots & \mu & 1 \\ \mu & \lambda & \cdots & \mu & 1 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ \mu & \mu & \cdots & \lambda & 1 \\ \mu & \mu & \cdots & \mu & 1 \end{pmatrix} \rightarrow \begin{pmatrix} \lambda-\mu & 0 & \cdots & 0 & 1 \\ 0 & \lambda-\mu & \cdots & 0 & 1 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & \lambda-\mu & 1 \\ 0 & 0 & \cdots & 0 & 1 \end{pmatrix}$$

如果 $\lambda=\mu$ 则 $\text{rank } A_2 = 1$
 $\lambda \neq \mu$ 则 $\text{rank } A_2 = n$.

列交换.

(3). $A = (a_{ij})_{n \times n}$, $a_{ij} = \min(i, j)$

$$A = \begin{pmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & 2 & 2 & \cdots & 2 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 1 & 2 & 3 & \cdots & (n-1) \\ 1 & 2 & 3 & \cdots & n \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 1 & \cdots & 1 \\ 0 & 1 & 1 & \cdots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ 0 & 0 & 0 & \cdots & 0 \end{pmatrix} \Rightarrow \text{rank } A = n.$$

$$\begin{pmatrix} 1 & 1 & 1 & \cdots & 1 \\ 0 & 1 & 1 & \cdots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 0 \end{pmatrix} \xrightarrow{-\text{②} + \text{④}, i=1, 2, \dots, l}$$

从下往上消.

线性映射的像与核

$$\begin{cases} \varphi(\vec{u} + \vec{v}) = \varphi(\vec{u}) + \varphi(\vec{v}) \\ \varphi(\alpha \vec{u}) = \alpha \varphi(\vec{u}) \end{cases}$$

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

Def1 设映射 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 满足对 $\forall \alpha, \beta \in \mathbb{R}$, $\vec{u}, \vec{v} \in \mathbb{R}^n$. $\varphi(\alpha \vec{u} + \beta \vec{v}) = \alpha \varphi(\vec{u}) + \beta \varphi(\vec{v})$

则称 φ 为线性映射 (即 可和 线性运算 (加法, 数乘) 交换 ~ 映射)

Def2 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 为线性映射

(i) $\text{Im } \varphi := \{\vec{w} \in \mathbb{R}^m \mid \exists \vec{v} \in \mathbb{R}^n \text{ s.t. } \vec{w} = \varphi(\vec{v})\} \subseteq \mathbb{R}^m$. 为像空间, 记为 $\text{Im } \varphi$. (image)

(ii) $\text{Ker } \varphi := \{\vec{v} \in \mathbb{R}^n \mid \varphi(\vec{v}) = \vec{0} \in \mathbb{R}^m\} \subseteq \mathbb{R}^n$. 为核空间, 记为 $\text{ker } \varphi$ (kernel).

$\therefore \text{Im } \varphi \subseteq \mathbb{R}^m$, $\text{ker } \varphi \subseteq \mathbb{R}^n$ 均为线性子空间.

2) φ 是单射 $\Leftrightarrow \text{ker } \varphi = \{\vec{0}\} \Rightarrow n \leq m \quad \cancel{n \geq m, \therefore m=n} \rightarrow$ 单射 \rightarrow 满射.

φ 是满射 $\Leftrightarrow \text{Im } \varphi = \mathbb{R}^m \Rightarrow n \geq m$.

Thm $\dim(\text{ker } \varphi) + \dim(\text{Im } \varphi) = n$ (也可从齐次线性方程组角度理解).

Def3 设 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 为线性映射. $\vec{e}^{(1)}, \vec{e}^{(2)}, \dots, \vec{e}^{(n)}$ 为 \mathbb{R}^n 的标准基 ($\vec{e}^{(j)} = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \end{pmatrix}_{n \times 1}^{j \text{ 位}}$)

设 $\varphi(\vec{e}^{(j)}) = \begin{pmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{mj} \end{pmatrix} \quad (j=1, 2, \dots, n)$ 令 $A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} = (\varphi(\vec{e}^{(1)}), \dots, \varphi(\vec{e}^{(n)}))$

为 φ 在标准基下的矩阵表示. 一般简记 A 为 φ 的矩阵.

Thm $\text{Hom}(\mathbb{R}^n, \mathbb{R}^m) \rightarrow \mathbb{R}^{n \times n}$ 是双射. 注: $\begin{cases} \text{Im } \varphi = V_c(A) \quad (\text{列空间}) \\ = \{\varphi(\vec{x}) \mid \vec{x} \in \mathbb{R}^n\}, \vec{x} = \sum x_i \vec{e}_i, \varphi(\vec{x}) = \sum x_i \varphi(\vec{e}_i) \end{cases}$

联系线性 $\Rightarrow \text{ker } \varphi$ 为 $\vec{A}^{(1)} x_1 + \cdots + \vec{A}^{(n)} x_n = \vec{0}$

方程组理解 \rightarrow 解空间. 理解秩公式:

$\{\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m \mid \varphi \text{ 为线性映射}\}$

i.e. $\varphi(\vec{x}) = A\vec{x}$

则 $\{\alpha_1, \alpha_2, \alpha_3\}$ 构成 $\ker \varphi$ 的一组基.

取 $\vec{\beta}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \vec{\beta}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$, 显然 $\vec{\beta}_1, \vec{\beta}_2 \notin \alpha_1, \alpha_2, \alpha_3$ 线性无关.

则 $\varphi(\vec{\beta}_1) = \begin{pmatrix} 1 \\ 3 \\ 5 \end{pmatrix}, \varphi(\vec{\beta}_2) = \begin{pmatrix} 1 \\ 2 \\ 4 \end{pmatrix}$ 构成 $\text{Im } \varphi$ 的一组基.

注: $\dim(\ker \varphi) + \dim(\text{Im } \varphi) = \dim(\text{V}_H) + \text{rank } A = \text{rank } A$ 由 A 为系数矩阵 \sim 齐次线性方程组解空间.

$\beta_1, \beta_2 \in \mathbb{R}^n$ 无关 $\Rightarrow \varphi(\beta_1), \varphi(\beta_2)$ 无关. 但这里 β_1, β_2 与 $\ker(\varphi)$ 中一组基线性无关.

可保证 $\varphi(\beta_1), \varphi(\beta_2)$ 仍无关.

§1 线性方程组

$$\left\{ \begin{array}{l} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{array} \right.$$

x_1, \dots, x_n 称未知元. $A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$ 为系数矩阵
 b_1, \dots, b_n 称为常数项.

如果 $b_1 = \dots = b_n = 0$ 称为齐次线性方程组.
否则称非齐次.

初等变换: I型(交换两行), II型(把某行乘一个倍数加到另一行), III型(某行乘倍数).
I型: 1) 初等变换不改变方程的解.

2) 任何方程组均可通过初等变换转化为阶梯型方程组.

3) 方程组有无穷多解(相容且不确定) \Leftrightarrow 未知元个数 $>$ 非零阶梯形方程个数
(不考虑不相容情形).

2 集合与映射

$f: X \rightarrow Y$ 是一个映射(如果 $\forall x \in X \exists ! y \in Y$ st. $f(x)=y$). 反义.

像集: $m_f = f(X) = \{y \in Y \mid \exists x \in X \text{ st. } y=f(x)\} = \{f(x) \mid \forall x \in X\}$

原像(逆像)集: 设 $V \subseteq Y$ 为子集. $f^{-1}(V) := \{x \in X \mid f(x) \in V\}$

如果 $V = \{y\}$ 则 $f^{-1}(\{y\}) = \{x \in X \mid f(x)=y\}$.

集合非空元素: f^{-1} 无意义.

f 是单射 $\Leftrightarrow (a \neq b \Rightarrow f(a) \neq f(b)) \Leftrightarrow (f(a)=f(b) \Rightarrow a=b)$ 双射.

f 是满射 $\Leftrightarrow f(X) = Y \Leftrightarrow \forall y \in Y, \exists x \in X$ st. $f(x)=y$.

§3 等价关系(与函数)

Def. $\sim \subseteq S \times S$ 为等价关系(相当于 $S \times S \sim \sim$, 其中 S 为集合) 满足

- 1) 反身性. ($\forall x \in S, x \sim x$).
- 2) 对称性 ($x \sim y \Rightarrow y \sim x$)
- 3) 传递性 ($x \sim y, y \sim z \Rightarrow x \sim z$)

§4 置换.

Def. 有限集 $X = \{1, 2, \dots, n\}$. $\sigma: X \rightarrow X$ 为双射, 则称 σ 为一个置换.

$$\text{记为 } \begin{pmatrix} 1 & 2 & 3 & \cdots & n \\ i_1 & i_2 & i_3 & \cdots & i_n \end{pmatrix}$$

循环分解: $\sigma = \pi_1 \pi_2 \cdots \pi_s$ 不相交循环之积.

设 $l_i = \text{len}(\pi_i)$ 为循环长度, 则.

σ 为 PP (ie $\sigma^k = \text{id}_X$ 最小正整数) 为 $\text{lcm}(l_1, l_2, \dots, l_s)$

σ 的符号为 $(-1)^{\frac{s(s-1)}{2}}$ (即可分解为奇/偶数个时积之积?)

§5 整数 - 算术.

1. 素数: $p \in \mathbb{Z}_{\geq 2}$; 是因为只有 1 和 $\pm p$. (有无穷多的素数)

2. 算术基本定理: $\forall n \in \mathbb{Z}_{\geq 2}$ 都可分解为若干素数之积, 且不计序唯一.

算术基本定理: $\forall n \in \mathbb{Z}_{\geq 2}$ 都可分解为若干素数之积, 且不计序唯一.

2. $\gcd \leq 1cm$.

扩展欧几里得算法: 对 $a, b \in \mathbb{Z}$ 存 $u, v \in \mathbb{Z}$ s.t. $ua + vb = \gcd(a, b)$

利用带余除法辗转相除求 \gcd . $r_0 = 1 \cdot a + 0 \cdot b = u_0a + v_0b$

$$\begin{cases} u_0 = 1 \\ v_0 = 0 \end{cases}$$

$$\begin{cases} u_1 = 0 \\ v_1 = 1 \end{cases}$$

$$\begin{cases} u_i = u_{i-2} - q_{i-1}u_{i-1} \\ v_i = v_{i-2} - q_{i-1}v_{i-1} \end{cases} \quad (i=2, \dots, k)$$

$$r_1 = 0 \cdot a + 1 \cdot b = u_1a + v_1b$$

$$r_2 = r_0 - q_1 r_1, r_1 = u_2a + v_2b$$

:

$$\boxed{\gcd(a, b) = r_k = r_{k-2} - q_{k-1}r_{k-1} = u_k a + v_k b}$$

$$0 = r_{k-1} - q_k r_k$$

$$\text{lcm}(a, b) = \frac{a \cdot b}{\gcd(a, b)}$$

§6 向量空间

1. 线性相关性 设 $\vec{v}_1, \dots, \vec{v}_k \in \mathbb{R}^n$

线性相关 $\Leftrightarrow \exists$ 不全为 0 的 $\alpha_1, \dots, \alpha_k \in \mathbb{R}$ st. $\alpha_1 \vec{v}_1 + \dots + \alpha_k \vec{v}_k = \vec{0}$

线性无关 \Leftrightarrow 如果 $\alpha_1, \dots, \alpha_k \in \mathbb{R}$ 满足 $\alpha_1 \vec{v}_1 + \dots + \alpha_k \vec{v}_k = \vec{0}$ 则 $\alpha_1 = \dots = \alpha_k = 0$

Thm 如果 $\vec{v}_1, \dots, \vec{v}_k$ 可由 $\vec{u}_1, \dots, \vec{u}_l$ 线性表示 则且 $k > l$ 则 $\vec{v}_1, \dots, \vec{v}_k$ 线性相关

2. 极大线性无关组

设 $S \subseteq \mathbb{R}^n$ T ⊆ S 极性无关且 $\forall v \in S$ 可由 T 中元素线性表示 保证(唯一)

3. 线性子空间

$V \subseteq \mathbb{R}^n$ 非空, $\forall \alpha, \beta \in \mathbb{R}, \vec{u}, \vec{v} \in V \quad \alpha\vec{u} + \beta\vec{v} \in V$ (即对加法, 数乘封闭)

$\langle \vec{v}_1, \dots, \vec{v}_k \rangle := \{ \sum_{i=1}^k \alpha_i \vec{v}_i \mid \alpha_i \in \mathbb{R} \}$ 是包含 $\{\vec{v}_1, \dots, \vec{v}_k\}$ 的最小子空间

4. 基与维数.

线性子空间 $V \subseteq \mathbb{R}^n$ 且 $\vec{v}_1, \dots, \vec{v}_k \in \mathbb{R}^n$ \Leftrightarrow 线性无关 st. $V = \langle \vec{v}_1, \dots, \vec{v}_k \rangle$

则 $\dim V = k$ 称为 V 的维数. (基不唯一)

维数公式: $V_1, V_2 \subseteq \mathbb{R}^n$ 为线性子空间 则:

$$\dim(V_1 + V_2) = \dim V_1 + \dim V_2 - \dim(V_1 \cap V_2)$$

矩阵的秩.

$A \in \mathbb{R}^{m \times n}$ $\text{rank } A = \dim V_r(A) = \dim V_c(A)$ (高斯消去得列阶梯形)

$\text{rank } A + \dim V_n(A) = n$ 变量个数.

§8 线性映射.

$\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$ 满足 $\forall \alpha, \beta \in \mathbb{R}, \vec{u}, \vec{v} \in \mathbb{R}^n \quad \varphi(\alpha\vec{u} + \beta\vec{v}) = \alpha\varphi(\vec{u}) + \beta\varphi(\vec{v})$ $\langle \varphi(\vec{e}_1), \dots, \varphi(\vec{e}_n) \rangle$

$\exists! A \in \mathbb{R}^{m \times n}$ st. $\varphi(\vec{v}) = A\vec{v}$ (矩阵)

具体如何 st. $\ker \varphi$ 和 $\text{im } \varphi$

$\dim \ker \varphi + \dim \text{im } \varphi = n$ $\begin{cases} \ker \varphi = V_H(A) \\ \text{im } \varphi = V_C(A) \end{cases} \quad \dim \ker \varphi = \dim \text{im } \varphi = \text{rank } A$

$(\vec{e}_1, \dots, \vec{e}_n) = \mathbb{R}^n$ A $\xrightarrow{\text{阶梯形}}$ 求其次线性方程组

$$\forall \vec{x} \in \mathbb{R}^n \quad \vec{x} = \sum_{i=1}^n x_i \vec{e}_i$$

解空间

$$\varphi(\vec{x}) = \sum x_i \varphi(\vec{e}_i) = A\vec{x}$$

\mathbb{R}^n 中基的像确定 则 φ 确定 \mathbb{R}^m 中研究之方法.