

①  $\varphi_A$  的定義  
 $(\varphi(\vec{e}^{(1)}), \dots, \varphi(\vec{e}^{(n)}) = (A_1, \dots, A_n)$

若  $\varphi$  是  $\mathbb{R}^n \rightarrow \mathbb{R}^m$  中的映射  
 $A = (\varphi(\vec{e}^{(1)}), \dots, \varphi(\vec{e}^{(n)}))$

$$[\boxed{\varphi_A} \quad A(\varphi_A) = A]$$

$Ag$  定義為  $\mathbb{R}^n$  中的映射  
 $Ag = (\varphi(\vec{e}^{(1)}), \dots, \varphi(\vec{e}^{(n)}))$

$Ag$  定義為  $\mathbb{R}^n$  中的映射  
 $Ag = \varphi(\vec{e}^{(1)}) = \varphi(\vec{e}^{(2)}) = \dots = \varphi(\vec{e}^{(n)})$

由定理 3.2.  $\varphi$  是  $\mathbb{R}^n \rightarrow \mathbb{R}^m$  中的映射

$$[\boxed{\varphi_g} = \varphi]$$

定理 3.3.  $\forall g: \mathbb{R}^n \rightarrow \mathbb{R}^m$  若  $Ag$  定義為  $\mathbb{R}^n$  中的映射，則  $Ag$  定義為  $\mathbb{R}^m$  中的映射

$\exists g: \text{Hom}(\mathbb{R}^n, \mathbb{R}^m) \rightarrow \mathbb{R}^m$   
 $\forall g: Ag \in \text{Hom}(\mathbb{R}^n, \mathbb{R}^m)$

定理 3.4.  $\forall g: \mathbb{R}^n \rightarrow \mathbb{R}^m$  若  $A$  為  $n \times n$  的實矩陣  
 $\exists g: \mathbb{R}^n \rightarrow \mathbb{R}^m$  使得  $g(A_i) = \boxed{A_i}$ ,  $i = 1, \dots, n$   
 $\text{則 } A_{(1)}, \dots, A_{(n)} \in \mathbb{R}^m$   
 $\text{定義: } \forall A \in \mathbb{R}^{n \times n}$  定義  $\varphi_A$  為  $\mathbb{R}^n$  中的映射  
 $\varphi_A(\vec{x}) = x_1V_1 + \dots + x_nV_n$   
 $\varphi_A(\vec{v}_i) = V_i, i = 1, \dots, n$

定理 3.5.  $\forall \vec{x} \in \mathbb{R}^n$  有  $\varphi_A(\vec{x}) = \boxed{Ax}$   
 $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^m$   
 $\forall V_1, \dots, V_n \in \mathbb{R}^m$  有  $\boxed{EV = V}$   
 $\boxed{\text{四 4.2: 定理 3.2}}$

定理:

$$\text{Hom}(\mathbb{R}^n, \mathbb{R}^m) \hookrightarrow \text{Hom}(\mathbb{R}^n, \mathbb{R}^m)$$

$$A \xrightarrow{\varphi_A}$$

$$\Phi(A) = \varphi(A) = \varphi_{\varphi_A} = A \circ \varphi_A = A$$

$$\Phi = \text{id}_{\mathbb{R}^{m \times n}}$$

$$\Phi = \text{id}_{\text{Hom}(\mathbb{R}^n, \mathbb{R}^m)}$$



$$\text{例: } \varphi_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$\text{ker}(\varphi_A) = \left\{ \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n \mid x_1 A_{11} + \dots + x_n A_{1n} = 0_m \right\}$$

$$= V_A$$

$$\text{im}(\varphi_A) = \left\{ \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^m \mid x_i A_{i1} + \dots + x_n A_{in} = 0_m \quad | \quad x_i \in \mathbb{R} \right\}$$

$$= V_c(A)$$

由定理 3.1 可知  $\dim \text{ker}(\varphi_A) = \text{rank}(A)$  且  $\dim \text{im}(\varphi_A) = \text{rank}(A)$

$$\boxed{\text{定理 3.1}}: \dim \text{ker}(\varphi_A) + \dim \text{im}(\varphi_A) = n.$$

②

求  $\ker(\phi)$  的一组基  $v_1, v_2, v_3$

$$\dim(\text{im}(\phi)) = 2 \Leftrightarrow W = \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\} \in V(A)$$

$$V_3 = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}, V_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\Leftrightarrow \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix} \leftarrow$$

$$\begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \leftarrow$$

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

$$\begin{pmatrix} 4 & -1 & -1 & 1 & 1 \\ 1 & 2 & 2 & 2 & 2 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} =$$

$$A_4 = \left( \varphi(e^{(1)}), \varphi(e^{(2)}), \varphi(e^{(3)}), \varphi(e^{(4)}) \right)$$

求  $\text{im}(\varphi)$  的一组基

$$\begin{pmatrix} 4x_1 + 2x_2 + 2x_3 + 2x_4 \\ x_1 - x_2 - x_3 - x_4 \\ x_1 + x_2 + x_3 + x_4 + x_5 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} \leftarrow$$

$$\text{设 } \varphi: \mathbb{R}^5 \rightarrow \mathbb{R}^3$$

求  $\text{im}(\varphi)$  的一组基

(解齐次线性方程组)

求  $\ker(\varphi)$  的一组基

(解齐次线性方程组)

求  $\varphi$  的核  $\ker \varphi$

求  $\ker(\varphi)$  的一组基.  $\text{im}(\varphi)$  的一组基

③

③

②

①

求

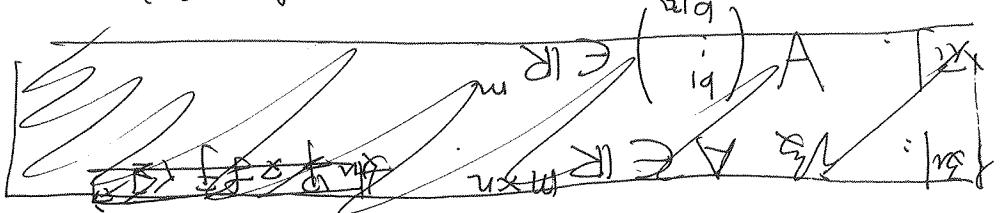
求

定理

$\exists g \in \text{Hom}(\mathbb{R}^n, \mathbb{R}^m)$ ,  $A \in g$  的矩阵;  $\text{rank}(g) = \text{rank}(A)$ .

3. 于  $\mathbb{R}^n$  空间中找  $\text{ker}(g)$

注: 设  $g \in \text{Hom}(\mathbb{R}^n, \mathbb{R}^m)$ ,  $A \in g$  的矩阵;  $\text{rank}(g) = \text{rank}(A)$ .



$\Leftrightarrow \text{rank}(A) = m \rightarrow$

$\dim V_c(A) = m \rightarrow$

$\dim V_c(A) = \mathbb{R}^m \rightarrow$

( $\text{rank } A = m$ )

( $\text{rank } A = n$ )

$\Leftrightarrow \text{rank}(A) = n \rightarrow$

$V_A = \{O_n\} \rightarrow$

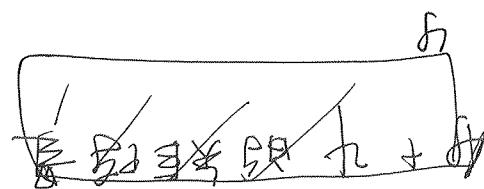
定理 (iii)  $g$  是满射  $\Leftrightarrow \text{ker}(g) = \{O_n\}$  ( $\text{rank } A = n$ )

(ii)  $g$  是满射  $\Leftrightarrow A$  为满秩

(ii)  $g$  是单射  $\Leftrightarrow A$  为单秩

设  $g: \mathbb{R}^n \rightarrow \mathbb{R}^m$  为满射,  $A \in g$  的矩阵

定理 3.4 (3 种判别单射和满射的充要条件)



$$g(\alpha x + \beta y) = \alpha(g(x)) + \beta(g(y))$$

$$= \alpha(g(x)) + \beta(g(y)) + \alpha x + \beta y$$

$$= \alpha g(x) + \beta g(y) + \alpha x + \beta y$$

$$(g+\alpha I)(\alpha x + \beta y) = g(\alpha x) + \beta y + (\alpha x + \beta y)$$

即:  $\forall \alpha, \beta \in \mathbb{R}, x, y \in \mathbb{R}^n$

也是线性映射

$$g(\alpha x + \beta y) = \alpha g(x) + \beta g(y)$$

定理 4.1.  $\forall g, h \in \text{Hom}(\mathbb{R}^n, \mathbb{R}^m)$

$$gA + hB = (g + h)A + hB$$

定理 4.1. 具有加法和数乘.

④

4. 矩阵运算

$\varphi_A + \varphi_B$  的定义域是

$$(\varphi_A + \varphi_B)(e^{(1)}), \dots, (\varphi_A + \varphi_B)(e^{(n)})$$

$$= (\varphi_A(e^{(1)}) + \varphi_B(e^{(1)}), \dots, \varphi_A(e^{(n)}) + \varphi_B(e^{(n)}))$$

$$= (\overbrace{A^{(1)} + B^{(1)}, \dots, A^{(n)} + B^{(n)}}^{\text{等效和函数}})$$

$$A + B = (a_{i1} + b_{i1}) \underset{m \times n}{\ldots} (a_{in} + b_{in})$$

$\forall \alpha: \forall A, B \in \mathbb{R}^{m \times n} \quad \forall \alpha$

$$= (a_{ij} + b_{ij}) \underset{m \times n}{\ldots}$$

$\forall \alpha: \forall A, B \in \mathbb{R}^{m \times n} \quad \forall \alpha$

$$\begin{aligned} & A + B = (\overbrace{A^{(1)} + B^{(1)}, \dots, A^{(n)} + B^{(n)}}^{\text{等效和函数}}) \\ & = (\overbrace{A_1 + B_1, \dots, A_m + B_m}^{\text{等效和函数}}) \end{aligned}$$

$\forall \alpha: \forall g \in \text{Hom}(\mathbb{R}^n, \mathbb{R}^m), \forall \alpha \in \mathbb{R}$

$$\begin{aligned} & g(A) = (g(A^{(1)}), \dots, g(A^{(n)})) \\ & g(A) = (g(a_{ij})) \underset{m \times n}{\ldots} \\ & \exists \alpha: \forall \alpha \in \mathbb{R}^{m \times n}, \forall \alpha \in \mathbb{R} \\ & g(A) = (\alpha g_A(e^{(1)}), \dots, \alpha g_A(e^{(n)})) \\ & = (\alpha \varphi_A(e^{(1)}), \dots, \alpha \varphi_A(e^{(n)})) \end{aligned}$$

$\forall \alpha: \forall \alpha, \beta \in \mathbb{R}, \forall \alpha, \beta \in \mathbb{R}^n$

$$\begin{aligned} & (\alpha \varphi_A)(e^{(1)}) + \beta (\varphi_A)(e^{(1)}) = \alpha \varphi_A(e^{(1)}) + \beta \varphi_A(e^{(1)}) \\ & (\alpha \varphi_A)(\alpha \varphi_A(e^{(1)}) + \beta \varphi_A(e^{(1)})) = \alpha \varphi_A(\alpha e^{(1)} + \beta e^{(1)}) \\ & \varphi_A(\alpha e^{(1)} + \beta e^{(1)}) = \varphi_A(\alpha e^{(1)}) + \varphi_A(\beta e^{(1)}) \end{aligned}$$

$$\text{rank}(A-B) = \dim V_c(A) + \dim V_c(B) - \dim(V_c(A) \cap V_c(B)) \leq \dim V_c(A) + \dim V_c(B) = \dim(V_c(A) + V_c(B))$$

$$2A - B = 2 \begin{pmatrix} 3 & 4 \\ 1 & 2 \end{pmatrix} + (-1) \begin{pmatrix} -3 & 0 \\ 1 & 2 \end{pmatrix} = \begin{pmatrix} 0 & 6 \\ 5 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 10 & 12 \\ 6 & 8 \end{pmatrix} + \begin{pmatrix} -5 & 0 \\ 9 & 7 \end{pmatrix} = \begin{pmatrix} 5 & 12 \\ 3 & 2 \end{pmatrix} =$$

$$\text{rank}(A+B) \leq \text{rank}(A) + \text{rank}(B)$$

$$\text{def: } A+B = \langle \underline{A^{(1)}+B^{(1)}}, \dots, \underline{A^{(n)}+B^{(n)}} \rangle$$

$$V_c(A+B) = \langle \underline{A^{(1)}+B^{(1)}}, \dots, \underline{A^{(n)}+B^{(n)}} \rangle$$

$$= \langle \underline{A^{(1)}}, \dots, \underline{A^{(n)}} \rangle + \langle \underline{B^{(1)}}, \dots, \underline{B^{(n)}} \rangle$$

$$= V_c(A) + V_c(B)$$

$$\text{rank}(A+B) = \dim V_c(A+B) \leq \dim(V_c(A) + V_c(B))$$

~~定理~~

$$\alpha \in \mathbb{R}$$

$\alpha$  が  $\mathbb{R}$  上の  $A$  の  $\mathbb{R}$  倍である

$\Rightarrow A + \alpha B$  は  $A$  と  $B$  の和である

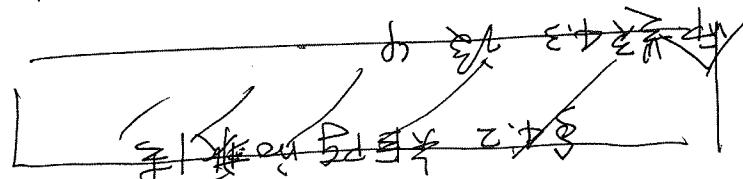
つまり  $\alpha$  は  $\mathbb{R}$  上の  $A, B$  の和である

$$\alpha \varphi_A = \varphi_{\alpha A}$$

$$\varphi_A + \varphi_B = \varphi_{A+B}$$

$\forall \alpha \in \mathbb{R}$  は  $A, B$  の和である

⑥

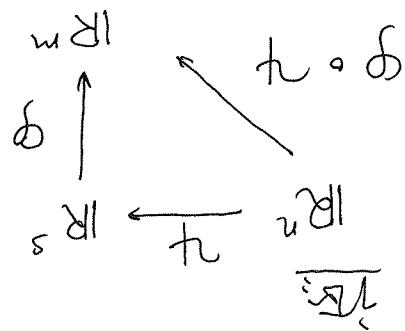


图

$$\leq \dim V_c(A) + \dim V_c(B) = \text{rank}(A) + \text{rank}(B)$$

④

$$\forall x, y \in \mathbb{R}^n \quad \varphi \circ f(x + y) = f(x) + f(y)$$



⑤

$$A + B = B + A, \quad (A + B) + C = A + (B + C)$$

$$A, B \in \mathbb{R}^{m \times n}$$

从下而上地推导之

$$A + O^{m \times n} = A$$

$$\alpha(A + B) = \alpha A + \alpha B, \quad (\alpha + \beta)A = \alpha A + \beta A$$

$$A \in \mathbb{R}^{m \times n}$$

矩阵的乘法

$$\alpha(\beta A) = (\alpha\beta)A$$

~~定理 3.1. 矩阵乘法的结合律~~

$$\alpha(A + B) = \alpha A + \alpha B, \quad (\alpha + \beta)A = \alpha A + \beta A$$

例 4.2 矩阵乘法的结合律

例 4.3.  $\forall A \in \mathbb{R}^n, \forall B \in \mathbb{R}^{n \times m}, \forall C \in \mathbb{R}^{m \times s}$

$(A \cdot B) \cdot C = A \cdot (B \cdot C)$

$$A = (a_{ik})_{i=1 \dots n, k=1 \dots m}, \quad B = (b_{ik})_{k=1 \dots m}$$

$\forall x \in \mathbb{R}^n \quad \varphi \circ f(x) = f(x)$

$$\begin{aligned} &= \varphi(f(x)) \\ &= \varphi(f(x) + f(y)) \\ &= \varphi(f(x) + f(y)) \\ &= \varphi(f(x + y)) \end{aligned}$$

$$\alpha, \beta \in \mathbb{R}$$

⑥

$$A + O^{m \times n} = A$$

$A, B \in \mathbb{R}^{m \times n}$

从下而上地推导之

$$= \sum_{k=1}^r a_{ik} b_{kj}$$

$$= a_{i1} b_{1j} + \dots + a_{is} b_{sj}$$

$$c_{ij} = b_{ij} A_{i1} + \dots + b_{is} A_{is}$$

且

$$a_{i1} b_{1j} + \dots + a_{is} b_{sj} = \begin{pmatrix} s \\ 1 & \dots & s \end{pmatrix} \cdot (a_{i1}, \dots, a_{is})$$

$i = 1, \dots, m, j = 1, \dots, n$

$$c_{ij} = a_{i1} b_{1j} + \dots + a_{is} b_{sj}$$

即矩阵  $C = (c_{ij})_{m \times n}$ .

AB ( $A, B$  的积) 等于矩阵

$$\boxed{A \neq B}$$

$B \in \mathbb{R}^{m \times n}$

$A \in \mathbb{R}^{n \times k}$

$B \in \mathbb{R}^{k \times m}$

$A \in \mathbb{R}^{n \times k}$

$B \in \mathbb{R}^{m \times n}$

$A \in \mathbb{R}^{n \times k}$

$B \in \mathbb{R}^{k \times m}$

$A \in \mathbb{R}^{n \times k}$

$B \in \mathbb{R}^{m \times n}$

$\{ \dots, r \} = \{$

$$c_{ij} = b_{ij} \underbrace{A_{i1} + \dots + A_{is}}_{A(s)}$$

$$C = (c_{11}, \dots, c_{nn})$$

$$C = \boxed{(b_{ij})}$$

$$g(\underbrace{B_{(1)}}_{b_{11}} + \dots + \underbrace{B_{(s)}}_{b_{ss}}) = \underbrace{A_{(1)}}_{a_{11}} + \dots + \underbrace{A_{(s)}}_{a_{ss}}$$

$$\underbrace{\underbrace{B_{(1)}}_{b_{11}}, \dots, \underbrace{B_{(s)}}_{b_{ss}}} = \begin{pmatrix} b_{11} \\ \vdots \\ b_{s1} \end{pmatrix} =$$

$$= (g(B_{(1)}), \dots, g(B_{(s)}))$$

( $B$  的各列)

$$(g(A_{(1)}), \dots, g(A_{(s)})) =$$

$$((\underbrace{e_{(1)}}_{(1)}, \dots, \underbrace{e_{(s)}}_{(s)}), \dots, (\underbrace{e_{(1)}}_{(1)}, \dots, \underbrace{e_{(s)}}_{(s)})) =$$

$\forall C \in \mathbb{R}^{m \times n} \Rightarrow g \circ f \circ g^{-1}$  为恒等映射.

4.2 矩阵的乘法

定理 BA 没有零元

$$\begin{pmatrix} 4 & 9 & 6 \\ 1 & 3 & 3 \end{pmatrix} =$$

$$\begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \\ 4 & 5 & 6 \\ 1 & 2 & 3 \end{pmatrix}$$

不可逆

$$\text{AB} \in \mathbb{R}^{2 \times 3}, \quad A \in \mathbb{R}^{2 \times 3}, \quad B \in \mathbb{R}^{3 \times 3}$$

① AB 为非零矩阵

$$⑨ \quad \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad A = \begin{pmatrix} 1 & 2 & 3 \end{pmatrix} \quad B = \begin{pmatrix} 4 & 5 & 6 \end{pmatrix}$$

$$C_{ij} = \overbrace{A_i}^{\left( \begin{array}{c} b_1 \\ \vdots \\ b_m \end{array} \right)} \overbrace{B_j^T}^{\left( \begin{array}{c} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{array} \right)} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{im}b_{mj}$$

$$C = \underbrace{\left( \begin{array}{c} \vdots \\ - & | & \dots \\ \vdots \end{array} \right)}_B \underbrace{\left( \begin{array}{c} \vdots \\ - & | & \dots \\ \vdots \end{array} \right)}_A$$

$$C_{ij} = \sum_{k=1}^s a_{ik}b_{kj}$$

$$C = (C_{ij})_{i=1 \dots m, j=1 \dots n}$$

$$A = (a_{ik})_{i=1 \dots m, k=1 \dots s}$$

$$B = (b_{kj})_{k=1 \dots s, j=1 \dots n}$$

$$C = AB \in \mathbb{R}^{m \times n}$$

$$\text{④ 2: } \forall A \in \mathbb{R}^{m \times s}, B \in \mathbb{R}^{s \times n}$$

(10)

例 1 等号右侧

$$\begin{aligned} \text{设 } A = (a_{ij})_{m \times n}, \quad \underline{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad \text{则 } Ax = \begin{pmatrix} a_{11}x_1 + \dots + a_{1n}x_n \\ a_{21}x_1 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n \end{pmatrix} = \end{aligned}$$

例 2 AB 不等于 BA 及其乘法不满足结合律

例 1 AB 不等于 BA 且不满足结合律

$$\begin{aligned} A &= \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \\ AB &= \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \\ BA &= \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \end{aligned}$$

例 2,  $A \neq O_{2 \times 2}, B \neq O_{2 \times 2}$  且

例 3. 请举出事例

$$\begin{aligned} AB &= CB \Rightarrow A=C \\ AB &= AC \Rightarrow B=C \end{aligned}$$

$$AB = CB \Rightarrow A=C$$

$$\begin{aligned} \text{设 } A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}, \quad \underline{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad \text{则 } Ax = \begin{pmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \end{pmatrix} = \end{aligned} \quad (11)$$

$$\varphi_{AB} = \varphi_A \circ \varphi_B$$

$\varphi_A$

$\varphi_B$

$\varphi_{AB}$

$$\varphi_A \circ \varphi_B = \varphi_{AB}$$

AB

3.4.4 函數乘積律: 若  $\varphi_A, \varphi_B$  為  $\mathbb{R}^n$  到  $\mathbb{R}^m$  之映射

$$\text{則 } \varphi_A: \mathbb{R}^s \rightarrow \mathbb{R}^m, \varphi_B: \mathbb{R}^n \rightarrow \mathbb{R}^s$$

則  $A \in \mathbb{R}^{m \times s}, B \in \mathbb{R}^{s \times n}$

3.4.3 共函數乘積律的直觀觀念

$$d(x) = \begin{pmatrix} a_{11}x_1 + \dots + a_{1n}x_n \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n \end{pmatrix} = Ax$$

$$A = d(x), x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

設  $d: \mathbb{R}^n \rightarrow \mathbb{R}^m$  為 3.4.4 的映射

~~3.4.3~~

$$\text{則 } (AB)C = A(BC)$$

$$\varphi_{BC} = \varphi_B \circ \varphi_C$$

$\varphi_B$

$\varphi_C$

$\varphi_{BC}$

$\varphi_A$

$\varphi_B$

$\varphi_A$

$\varphi_{AB}$

3.4.3 合併律: 設  $A \in \mathbb{R}^{m \times s}, B \in \mathbb{R}^{s \times k}, C \in \mathbb{R}^{k \times n}$

$$\text{則 } E_m A = A E_n = A$$

單位元的性質:  $E_n$  當  $n \neq k$  單位元的性質

$$\text{則: } \text{設 } A \in \mathbb{R}^{m \times n}, E_m \text{ 當 } m \neq n$$

$$\text{由定理3.2 } A(BC) = (AB)C$$

$$\varphi_{A(BC)} = \varphi_A \circ \varphi_{BC}$$

$$\varphi_A \circ (\varphi_B \circ \varphi_C) = (\varphi_A \circ \varphi_B) \circ \varphi_C$$

$$\varphi_A \circ (\varphi_B \circ \varphi_C) = (\varphi_A \circ \varphi_B) \circ \varphi_C$$

由映射的合併律的定義:

$$\varphi_{BC} = \varphi_B \circ \varphi_C$$

$\varphi_B$

$\varphi_C$

$\varphi_{BC}$

$\varphi_A$

$\varphi_B$

$\varphi_A$

$\varphi_{AB}$

①

✓

$$C(\alpha A) = \alpha(CA)$$

✓

$$C(A+B) = CA+CB$$

✓  $C \in \mathbb{R}^{k \times m}$

✓  $\forall A, B \in \mathbb{R}^{m \times n}$

□

$$= (a_{11}, a_{12}, \dots, a_{1n}) \begin{pmatrix} & \\ & \ddots \\ & & i \\ & & & \end{pmatrix} = a_{1i}$$

$$\text{A } b_{ij} = \overbrace{\text{A}_i \cdot E_j^T}^1 = \overbrace{\text{A}_i^T \cdot E_j}^2$$

✓  $B = (b_{ij})^{m \times n} = AE^n$

$$AE^n = A$$

$$\Rightarrow A = E^mA$$

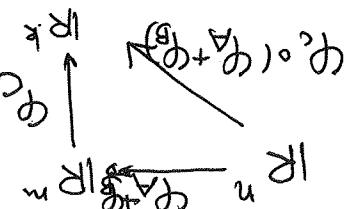
$$= \varphi_{E^mA}$$

$$\downarrow id_{E^m} \quad id_E \circ \varphi_A$$

$$id_E \circ \varphi_A : \mathbb{R}^n \rightarrow \mathbb{R}^n$$

映射  $\varphi_A : id_E^m \circ \varphi_A \circ id_E^m$

✓  $id_E^m : \mathbb{R}^m \rightarrow \mathbb{R}^m$



$$\text{② } \varphi_C \circ (\varphi_A + \varphi_B) = \varphi_C \circ \varphi_A + \varphi_C \circ \varphi_B$$

$$\varphi_C(\varphi_A(x)) = \varphi_C(\varphi_C \circ \varphi_A)(x)$$

$$\varphi_C \circ (\alpha \varphi_A)(x) = \varphi_C(\alpha \varphi_A(x))$$

$$\Leftrightarrow C(A+B) = CA+CB$$

$$\text{左邊等式 } \varphi_C \circ (\varphi_A + \varphi_B) = \varphi_C \circ \varphi_A + \varphi_C \circ \varphi_B$$

$$= \varphi_C \circ \varphi_A(x) + \varphi_C \circ \varphi_B(x)$$

$$= \varphi_C(\varphi_A(x)) + \varphi_C(\varphi_B(x))$$

$$= \varphi_C(\varphi_A(x) + \varphi_B(x))$$

$$A \in \mathbb{R}^n \quad \varphi_A \in \mathbb{R}^n$$

$$= \varphi_C(A+B)$$

$$= \varphi_C \circ (\varphi_A + \varphi_B) = \varphi_C \circ \varphi_A + \varphi_C \circ \varphi_B$$

注意

$$C \circ \varphi_A = \varphi_{AC} \circ \varphi_A \Leftrightarrow C(\alpha A) = \alpha(CA)$$

$$B^t = (b_{j,k}^t)_{\substack{j=1, \dots, n \\ k=1, \dots, s}} \quad b_{j,k}^t = b_{k,j}$$

$$A^t = (a_{i,k}^t)_{\substack{i=1, \dots, s \\ k=1, \dots, n}} \quad a_{i,k}^t = a_{k,i}$$

$$\text{則: } \forall A = (a_{i,k})_{\substack{i=1, \dots, m \\ k=1, \dots, n}}, B = (b_{i,k})_{\substack{i=1, \dots, m \\ k=1, \dots, n}}$$

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

$$\text{總量: } \forall A \in \mathbb{R}_{max}^m; B \in \mathbb{R}_{sym}^n$$

$$A(\alpha C) = \alpha AC$$

$$(A+B)C = AC + BC$$

$$\text{則: } \forall C \in \mathbb{R}_{n \times k}^n$$

$$C \circ \varphi_A = \varphi_{AC} \circ \varphi_A \Leftrightarrow$$

$$C = AB = (c_{i,j})_{\substack{i=1, \dots, m \\ j=1, \dots, n}}$$

$$C^t = (c_{j,i}^t)_{\substack{j=1, \dots, m \\ i=1, \dots, n}} \quad c_{j,i}^t = c_{i,j}$$

$$D^t = B^t A^t = (d_{j,i}^t)_{\substack{j=1, \dots, m \\ i=1, \dots, n}} \quad d_{j,i}^t = \sum_{k=1}^n b_{j,k}^t a_{k,i}^t$$

$$C^t = \sum_{j=1}^m c_{j,i}^t = \sum_{j=1}^m \sum_{k=1}^n a_{k,i}^t b_{j,k}^t = \sum_{k=1}^n a_{k,i}^t \sum_{j=1}^m b_{j,k}^t = c_{i,j} = c_{j,i}$$

$$AB = AC \Leftrightarrow \begin{cases} A = O_{max} \\ B, C \in \mathbb{R}_{sym}, \\ A \neq O_{max} \end{cases} \quad \text{只有} \begin{cases} A = O_{max} \\ B, C \in \mathbb{R}_{sym}, \\ A \neq O_{max} \end{cases} \text{時} \quad AB = AC$$

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这等于  $\text{diag}(x_1 \dots x_n)$

$$\begin{pmatrix} x_1 & & \\ & \ddots & \\ & & x_n \end{pmatrix}$$

例： $\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$

对角矩阵

$$A = \begin{pmatrix} a_{11} & & & \\ & a_{22} & \dots & \\ & & \ddots & \\ & & & a_{nn} \end{pmatrix}$$

$$x_i = a_{ii}$$

$$\begin{aligned} A &= (a_{ij})_{m \times n} \\ B &= \text{diag}(x_1, \dots, x_m) \\ b_{ij} &= B_i A_j = (0 \dots 0 x_i 0 \dots 0) \end{aligned}$$

$$\text{证： } ① \quad \forall A = (a_{ij})_{m \times n}$$

$$② A \text{diag}(x_1, \dots, x_n) = (x_1 A_1, \dots, x_n A_n)$$

$$\begin{pmatrix} x_1 A_1 & & & \\ & \ddots & & \\ & & x_n A_n & \end{pmatrix} = A \text{diag}(x_1, \dots, x_n)$$

$\text{diag}_n(x) \neq x \in \mathbb{R}^n$  不对称矩阵

$$\text{diag}_n(O) = \begin{pmatrix} O & & \\ & \ddots & \\ & & O \end{pmatrix}$$

(14)

$$AB = (a_{ij} b_{ij})_{m \times n}$$

例： Hadamard's product

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

AA 不 AA

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} e & f \\ g & h \end{pmatrix} = \begin{pmatrix} ae+bg & af+bh \\ ce+dg & cf+dh \end{pmatrix}$$

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \odot \begin{pmatrix} p & q \\ r & s \end{pmatrix} = \begin{pmatrix} ap & bq \\ cr & ds \end{pmatrix}$$

这等于  $\text{diag}(x_1 \dots x_n)$

$$\left( \begin{matrix} \beta_1 & \beta_2 & \dots & \beta_n \\ \alpha_1 & \alpha_2 & \dots & \alpha_n \end{matrix} \right)^t = \left( \begin{matrix} \beta_1 \\ \vdots \\ \beta_n \\ \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{matrix} \right)$$

类似地：

$$A = (A^{(1)}, \dots, A^{(n)}) , \quad A^t =$$

$$A = \left( \begin{matrix} A_1 & & & \\ & \ddots & & \\ & & A_m & \end{matrix} \right)_{n \times m}, \quad A^t = (A_1^t, \dots, A_m^t)$$

$$\left( \begin{matrix} \beta_1 & \beta_2 & \dots & \beta_m \\ \alpha_1 & \alpha_2 & \dots & \alpha_m \end{matrix} \right)^t = (\beta_1, \dots, \beta_m)$$

$$\left( \begin{matrix} \alpha_1 & \alpha_2 & \dots & \alpha_n \\ \beta_1 & \beta_2 & \dots & \beta_n \end{matrix} \right)^t = (\alpha_1, \dots, \alpha_n)$$

首先介紹轉置的某些特性和規則

但我們利用轉置來做：

② 由上可知：  $(A^t)^t = A$

$$\begin{aligned} & \left( \begin{matrix} x_1 & x_2 & \dots & x_n \\ x_1^t & x_2^t & \dots & x_n^t \end{matrix} \right)^t = \left( \begin{matrix} x_1^t & x_2^t & \dots & x_n^t \\ x_1 & x_2 & \dots & x_n \end{matrix} \right) \\ & = \text{diag}(x_1, \dots, x_n) \cdot A^t \\ & = \text{diag}(x_1, \dots, x_n)^t \cdot A^t \\ & B^t = [A \text{ diag}(x_1, \dots, x_n)]^t \end{aligned}$$

$$\therefore B = A \text{ diag}(x_1, \dots, x_n)$$

$$+ \left( \begin{matrix} 0 & 0 & \dots & 0 & x_n \\ 0 & - & & & \\ x_1 & 0 & 0 & \dots & 0 \\ \vdots & & & & \\ 0 & 0 & \dots & 0 & x_n \end{matrix} \right)$$

這

③

由上可知： $(A^t)^t = A$

$$\Rightarrow DA = AD$$

$$AD = (A(E_n) - \alpha(AE_n)) = \alpha A$$

$$DA = (\alpha E_n) A = \alpha(EMA) = \alpha A$$

$$\text{由此得 } DA = AD$$

$$\text{例 } \forall D = \alpha E_n, A \in \mathbb{R}^{n \times n}$$

$$DA \neq AD$$

$$\text{例 } DA = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 10 & 18 \\ 14 & 9 \end{pmatrix} \quad AD = \begin{pmatrix} 7 & 16 & 27 \\ 21 & 24 & 27 \\ 8 & 10 & 12 \end{pmatrix}$$

$$\therefore DA \neq AD$$

$$\text{例: } \forall A = \begin{pmatrix} 4 & 2 & 3 \\ 7 & 5 & 6 \\ 8 & 9 & 3 \end{pmatrix}, D = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$