

0514 ①

$$\text{Im}(A) := \text{Im}(\Phi A)$$

$$\text{Ker}(A) := \text{Ker}(\Phi A)$$

Thm 2.24 (Rank-Nullity)
 $\begin{matrix} \text{Rank} & \text{Nullity} \\ \parallel & \parallel \\ \dim \text{Im}(A) & \dim \text{Ker} A \end{matrix}$

Let V, W be finite-dim vector spaces

Let $\Phi: V \rightarrow W$ be a linear transformation

$$\dim V = \dim(\text{Ker } \Phi) + \dim(\text{Im } \Phi)$$

Rmk

① If $\dim(\text{Im}(\Phi)) < \dim V$
 $\Rightarrow \text{Ker } \Phi \neq \{0\} \Rightarrow \Phi$ is not 1-1

② 0

③ If $\dim V = \dim W$, then:

- Φ is 1-1
 - Φ is onto
 - Φ is bijective
-) 3個會同時對/錯

2.8. Affine Subspaces

2.8.1. Affine Subspaces

Def'n (AS)

Let V be a vector space

$$x_0 \in V \text{ \& } U \leq V$$

$$\Rightarrow L = x_0 + U \text{ (平移)} = \{x_0 + u \mid u \in U\} \subseteq V$$

L is called an affine subspaces / linear manifold

* Note that if $x_0 \notin U \Rightarrow 0 \notin x_0 + U \Rightarrow$ 他不是向量空間
 \Rightarrow AS 可能不是向量空間

Rmk

Let $L = x_0 + U$ be an AS, $\{b_1, \dots, b_k\}$ be a basis of U

$$\dim L := \dim U$$

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1. $\frac{n-1}{2}$ 維 AS 叫 line
2. $\frac{n-2}{2}$ 維 AS 叫 plane
3. hyperplane: \mathbb{R}^n 裡的 AS "L", $\dim L = n-1$

2.8.2. Affine Mapping $\phi: V \rightarrow W$ Def'n 在 W 裡存在一向量 a , 及一個線性轉換 ψ

$$\text{s.t. } \phi(x) = a + \psi(x)$$

Rmk ① 每一個 AM 都是線性轉換再做平移

② 2 個 AM 合完 still AM

CH3 Analytic Geometry 解析几何

3.1 Norms

Def'n 3.1 (Norm)

現有一 sp V , A norm on V is a fcn

$$\|\cdot\|: V \rightarrow \mathbb{R} \quad (\|x\| = \text{向量 } x \text{ 的長度})$$

s.t. $\forall x \in \mathbb{R}, x, y \in V$:

① Absolutely homogeneous

$$\|\lambda \cdot x\| = |\lambda| \cdot \|x\|$$

② Triangle inequality

$$\|x+y\| \leq \|x\| + \|y\|$$

③ Positive definite 正定

$$\text{① } \|x\| \geq 0 \quad \text{② } \|x\| = 0 \Leftrightarrow x = 0$$

Manhattan Norm (One Norm)

$$\|x\|_1 := \sum_{i=1}^n |x_i|, \quad x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

Euclidean Norm (Two Norm)

$$\|x\|_2 := \sqrt{\sum_{i=1}^n x_i^2}$$

3.2. Inner Product 內積

3.2.1 Dot Product

(我們平常熟悉的 $x \cdot y$) = $\langle x, y \rangle$

$$= x^T y$$

3.2.2 General Inner Product

bilinear map: 一個固定 x 的時候是線性

$$\Omega(x, y), \quad \forall x, y, z \in V, \lambda, \psi \in \mathbb{R}$$

$$\Omega(\lambda x + \psi y, z) = \lambda \Omega(x, z) + \psi \Omega(y, z)$$

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Def'n 3.2 Let $\Omega: V \times V$ be a bilinear map

$$\text{① if } \Omega(x, y) = \Omega(y, x)$$

 $\Rightarrow \Omega$ is symmetric② 所有不是 0 的向量, 自己和自己內積都 > 0 $\Rightarrow \Omega$ is 正定

Def'n 3.3

① Ω is 正定 & symmetric on V $\Rightarrow \Omega$ is inner product通常內積符號: $\langle x, y \rangle$ ② 內積空間: $(V, \langle \cdot, \cdot \rangle)$

$$\langle x, y \rangle = \underbrace{[x_1, x_2]}_{x^T} \underbrace{\begin{bmatrix} a & c \\ d & b \end{bmatrix}}_A \underbrace{\begin{bmatrix} y_1 \\ y_2 \end{bmatrix}}_y$$

$$\bullet A^T = A \quad (\text{symmetric})$$

 $\bullet A$ is 正定

$$\forall x \in \mathbb{R}^2, x^T A x \geq 0, \quad x^T A x = 0 \Leftrightarrow x = 0$$

$$\text{Let } \langle x, y \rangle = [x]^T_B A [y]_B$$

$$a_{ij} = \langle b_i, b_j \rangle$$