

P2

(a) Inner product space ✓

①  $\langle f, g \rangle = \langle g, f \rangle$

② linearity:  $\langle af_1 + bf_2, g \rangle$

$$= \frac{1}{\pi} \int_{-\pi}^{\pi} (af_1(t) + bf_2(t))g(t) dt$$

$$= \frac{a}{\pi} \int_{-\pi}^{\pi} f_1(t)g(t) dt + \frac{b}{\pi} \int_{-\pi}^{\pi} f_2(t)g(t) dt$$

$$= a\langle f_1, g \rangle + b\langle f_2, g \rangle$$

③ positive define:  $\langle f, f \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f^2(t) dt$   
 $f^2(t) \geq 0 \Rightarrow \langle f, f \rangle \geq 0$

(b) 同 (a), inner product space ✓

(c) 不是 inner prod space: can diverge,  $\omega \notin \mathbb{R}$

P4

find two different inner products in  $\mathbb{R}^2$

① (dot product)  $\langle x, y \rangle = x_1y_1 + y_1y_2$

here  $e_1, e_2$  are orthonormal

② different weight:  $\langle x, y \rangle = 3x_1y_1 + 2x_2y_2$

P5

Every finite dimensional inner product space has an orthonormal basis.

Take:  $u_1 = \frac{v_1}{\sqrt{\langle v_1, v_1 \rangle}}$

$$u_2 = \frac{v_2 - \langle v_2, u_1 \rangle u_1}{\|v_2 - \langle v_2, u_1 \rangle u_1\|}$$

$$\therefore u_n = \frac{v_n - \sum_{j=1}^{n-1} \langle v_n, u_j \rangle u_j}{\|v_n - \sum_{j=1}^{n-1} \langle v_n, u_j \rangle u_j\|}$$

point: generalized Gram-Schmidt

P8

Pf: 任取 orthonormal basis  $u$

任取 inner product  $\langle \cdot, \cdot \rangle$

那么  $(V, \langle \cdot, \cdot \rangle)$  上,

$$\langle \vec{x}, \vec{y} \rangle = [\vec{x}]_u \cdot [\vec{y}]_u$$

Pf  $\vec{x} = \sum_{i=1}^n a_i \vec{u}_i$  for some  $a_1, \dots, a_n$

$\vec{y} = \sum_{j=1}^n b_j \vec{u}_j$  for some  $b_1, \dots, b_n$

$$\langle \vec{x}, \vec{y} \rangle = \langle \sum_{i=1}^n a_i \vec{u}_i, \sum_{j=1}^n b_j \vec{u}_j \rangle$$

$$= \sum_{i=1}^n \sum_{j=1}^n a_i b_j \langle \vec{u}_i, \vec{u}_j \rangle$$

$$\Rightarrow = \left( \sum_{i=1}^n a_i b_i \right) = [\vec{x}]_u \cdot [\vec{y}]_u$$

P9  $\forall$  inner product  $\langle \cdot, \cdot \rangle$  on  $\mathbb{R}^n$ ,

$\exists$   $n \times n$  (symmetric) matrix  $A$ , 使  $\forall \vec{x}, \vec{y} \in \mathbb{R}^n, \langle \vec{x}, \vec{y} \rangle = \vec{x}^T A \vec{y}$

Pf Claim ①: note that:

对任意  $n \times n$  matrix  $A$ ,  $A$  的  $(i, j)$  entry

是  $\vec{e}_i^T A \vec{e}_j$

$$A = \begin{bmatrix} | & | & | & | \\ A\vec{e}_1 & A\vec{e}_2 & A\vec{e}_j & \dots & A\vec{e}_n \\ | & | & | & | \end{bmatrix}$$

$$\vec{e}_i^T A \vec{e}_j = \begin{bmatrix} 0 & \dots & 1 & \dots & 0 \end{bmatrix} \begin{bmatrix} | \\ A\vec{e}_j \\ | \end{bmatrix}$$

= the  $i^{th}$  entry of  $A\vec{e}_j$

(= the  $(i, j)^{th}$  entry of  $A$ )

② 注意到  $\langle \vec{x}, \vec{y} \rangle = \left\langle \sum_{i=1}^n x_i \vec{e}_i, \sum_{j=1}^n y_j \vec{e}_j \right\rangle$

$$= \sum_{i=1}^n \sum_{j=1}^n x_i y_j \langle \vec{e}_i, \vec{e}_j \rangle$$

③ Consider:  $A$  的  $(i, j)$  entry 是  $\langle \vec{e}_i, \vec{e}_j \rangle$

$$= \sum_{i=1}^n x_i \sum_{j=1}^n y_j \langle \vec{e}_i, \vec{e}_j \rangle$$

$$\Rightarrow = \sum_{i=1}^n x_i \left( \sum_{j=1}^n y_j a_{ij} \right)$$

$$= \vec{x} \cdot \begin{bmatrix} y_1 a_{11} + y_2 a_{12} + \dots + y_n a_{1n} \\ \vdots \\ y_n a_{n1} + y_n a_{n2} + \dots + y_n a_{nn} \end{bmatrix}$$

$$= \vec{x} \cdot (A\vec{y})$$

$$= \vec{x}^T A \vec{y}$$

A 是 symmetric 是因为

$$\langle e_i, e_j \rangle = \langle e_j, e_i \rangle$$

因而可总结:

对于任意  $\mathbb{R}^n$  上的 inner product,

$$\langle \vec{x}, \vec{y} \rangle = \vec{x}^T \begin{bmatrix} \langle e_1, e_1 \rangle & \langle e_1, e_2 \rangle & \dots & \langle e_1, e_n \rangle \\ \langle e_1, e_2 \rangle & \dots & \dots & \dots \\ \vdots & \dots & \ddots & \vdots \\ \langle e_n, e_1 \rangle & \dots & \dots & \langle e_n, e_n \rangle \end{bmatrix} \vec{y}$$

$$(\quad = \vec{x} \cdot (A\vec{y}))$$

注意到: A 的 diagonal 上的 entry 都  $> 0$

(by positive-definite)

又: A 是 symmetric 的

显然: A 是可逆的 (full rank)

P10: All inner products on  $\mathbb{R}^2$  都是

$$B_A(\vec{x}, \vec{y}) = \vec{x}^T A \vec{y} \text{ for some } A = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

$$\textcircled{1} A = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

(by P9)

① symmetry matrix

$\downarrow$

$$B_A(\vec{x}, \vec{y})$$

$$= B_A(\vec{y}, \vec{x})$$

$$\textcircled{2} \text{且 } a > 0,$$

$$\textcircled{3} \det A = ac - b^2 > 0$$

$$\textcircled{2} \begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = ax^2 + 2bxy + cy^2 > 0$$

(这是一个初中数学: 充分条件为  $a > 0$  且  $ac - b^2 > 0$ )

③ linearity is guaranteed by matrix multiplication.

因而这三条为充要条件.

实际上可理解为: 把  $\vec{x}$  伸缩变换后再知  $\vec{x}$  dot product.

$\mathbb{R}^n$  的任何 inner product 都是: 把一个 vector 做一个 linear trans 后和另一个 dot prod.