

## Chapter 5 Inner Product Spaces

### 5-1 Inner Products and Norms

**Inner product  $\langle x, y \rangle$ :**  $V$  is a vector space over  $F$ , and  $\langle x, y \rangle$  has the following characteristics:

(a)  $\langle ax+bz, y \rangle = a\langle x, y \rangle + b\langle z, y \rangle$ , (b)  $\langle x, y \rangle = \langle y, x \rangle^*$ , (c)  $\langle x, x \rangle$  is positive if  $x \neq 0$ .

**Eg. Let  $u=(u_1, u_2, u_3)$  and  $v=(v_1, v_2, v_3)$ . Determine which of the following are inner products on  $R^3$ .**

(i)  $\langle u, v \rangle = u_1^2 v_1^2 + u_2^2 v_2^2 + u_3^2 v_3^2$ , (ii)  $\langle u, v \rangle = 2u_1 v_1 + u_2 v_2 + 3u_3 v_3$ , (iii)  $\langle u, v \rangle = u_1 v_1 - u_2 v_2 + u_3 v_3$ . [2003交大電信所]

(Sol.) Only (ii) is the inner product. (i) does not fulfill (a), (iii) does not fulfill (c) if  $u_1 v_1 + u_3 v_3 < u_2 v_2$ .

**Eg. The following calculations fulfill the definition of the inner product:**

(a)  $x=(a_1, a_2, \dots, a_n)$ ,  $y=(b_1, b_2, \dots, b_n)$ ,  $\langle x, y \rangle = \sum_{i=1}^n a_i \bar{b}_i$ , (b)  $A, B \in M_{n \times n}(F)$ ,  $\langle A, B \rangle = \text{tr}(B^* A)$ ,

(c)  $f, g \in V$ ,  $\langle f, g \rangle = \int_0^a f(x) \overline{g(x)} dx$ .

In **Matlab** language, we can use the following instructions to obtain the inner product and the outer product of two vectors:

```
>>A=[4,-1,3]; B=[-2,5,2]; C=dot(A,B); D=cross(A,B)
```

A =

```
4    -1    3
```

B =

```
-2    5    2
```

C =

```
-7
```

D =

```
-17   -14   18
```

**Norm  $\| \cdot \|$ :** A real-valued function on  $V$  and satisfies the following conditions for all  $x, y \in V$  and  $c \in F$ .

(a)  $\|cx\| = |c| \cdot \|x\|$ , (b)  $\|x\| \geq 0$ , and  $\|x\|=0$  if  $x=0$ , (c)  $\|x+y\| \leq \|x\| + \|y\|$ .

**Eg. Show that  $\|-x\| = \|x\|$ .** [1993 台大電研]

(Proof)  $\|-x\| = |-1| \cdot \|x\| = \|x\|$

**Eg. The following calculations fulfill the definition of the norm:**

(a)  $V=M_{m \times n}(F)$ ,  $\|A\| = \max_{i,j} |A_{ij}|$ , (b)  $V=C[0, a > 0]$ ,  $\|f\| = \max_{t \in [0, a]} |f(t)|$ , (c)  $V=F^n$ ,  $\|x\| = \left[ \sum_{i=1}^n |a_i|^2 \right]^{1/2}$

**Euclidean norm of a vector:**  $x \in V$ ,  $\|x\| = \sqrt{\langle x, x \rangle}$

**Theorem (a)**  $|\langle x, y \rangle| \leq \|x\| \cdot \|y\|$  (**Cauchy-Schwartz Inequality**), (b)  $\|x+y\|^2 + \|x-y\|^2 = 2\|x\|^2 + 2\|y\|^2$ .

(Remark) By definition of the Euclidean norm.

In **Matlab** language, we can use the following instructions to obtain the Euclidean norm of a vector:

```
>> x=[3 4 5]; c=norm(x)
```

c =

7.0711

**Orthogonal vectors  $x$  and  $y$ :**  $x \perp y \Leftrightarrow \langle x, y \rangle = 0$ .

**Orthonormal vectors  $x$  and  $y$ :**  $\|x\| = \|y\| = 1$  and  $\langle x, y \rangle = 0$  if  $x \neq y$ .

**Eg. Show that if  $\{v_1, v_2, v_3, \dots, v_n\}$  is an orthogonal set of non-zero vectors in an inner product space, then  $v_1, v_2, v_3, \dots, v_n$  are linearly independent. [2015 中央電研固態組、生醫電子組]**

(Proof) Let  $y = \sum_{i=1}^n a_i v_i$  be a linear combination of  $v_1, v_2, v_3, \dots, v_n$ .

Since  $\{v_1, v_2, v_3, \dots, v_n\}$  is an orthogonal set of non-zero vectors in an inner product space, we have  $\langle v_i, v_j \rangle = 0$  if  $j \neq i$  but  $\langle v_i, v_i \rangle \neq 0$  in case of  $j = i$ .

Consider an inner product  $\langle y, v_j \rangle = \langle \sum_{i=1}^n a_i v_i, v_j \rangle = \sum_{i=1}^n a_i \langle v_i, v_j \rangle = a_j \langle v_j, v_j \rangle$

Set  $y=0$  and then we have  $a_i=0$  because  $\langle v_i, v_i \rangle \neq 0$  for each  $i$ . It implies that  $v_1, v_2, v_3, \dots, v_n$  are linearly independent.

**Theorem**  $V$  is an inner product space and  $S = \{x_1, \dots, x_m\}$ . If  $y = \sum_{i=1}^m a_i x_i$ , and then (a)

$a_j = \langle y, x_j \rangle / \langle x_j, x_j \rangle$  for all  $j$  if  $S$  is orthogonal, and (b)  $a_j = \langle y, x_j \rangle$  for all  $j$  if  $S$  is orthonormal.

(Proof) (a)  $\langle y, x_j \rangle = \langle \sum_{i=1}^m a_i x_i, x_j \rangle = \sum_{i=1}^m a_i \langle x_i, x_j \rangle = a_j \langle x_j, x_j \rangle \Rightarrow a_j = \frac{\langle y, x_j \rangle}{\langle x_j, x_j \rangle}$

**Eg.  $(8, -7) = 8(1, 0) - 7(0, 1)$  where  $8 = \langle (8, -7), (1, 0) \rangle$  and  $-7 = \langle (8, -7), (0, 1) \rangle$**

**Theorem** A linear operator  $T: V \rightarrow W$ , let  $\beta = \{x_1, \dots, x_m\}$  and orthogonal set  $\beta' = \{y_1, \dots, y_m\}$  are the bases of  $V$  and  $W$ , respectively. Then the  $ij$ -entry  $T_{ij} = \frac{\langle T(x_j), y_i \rangle}{\langle y_i, y_i \rangle}$ . If  $\beta'$  is an orthonormal basis,

then  $T_{ij} = \langle T(x_j), y_i \rangle$ .

(Proof)  $\because T(x_j) = \sum_{k=1}^m T_{kj} y_k, \therefore \langle T(x_j), y_i \rangle = \langle \sum_{k=1}^m T_{kj} y_k, y_i \rangle = \sum_{k=1}^m T_{kj} \langle y_k, y_i \rangle = T_{ij} \langle y_i, y_i \rangle$   
 $\Rightarrow T_{ij} = \frac{\langle T(x_j), y_i \rangle}{\langle y_i, y_i \rangle}$

**Eg. For a linear operator  $T(a_1, a_2) = (a_1 - a_2, a_1 + 2a_2)$ , let  $\beta = \{(1, 2), (2, 3)\}$  be a basis for  $R^2$ , and  $\beta' = \{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$  be an orthonormal basis for  $R^3$ , then we have  $T(1, 2) = (-1, 1, 4) = -1(1, 0, 0) + 1(0, 1, 0) + 4(0, 0, 1)$  and  $T(2, 3) = (-1, 2, 7) = -1(1, 0, 0) + 2(0, 1, 0) + 7(0, 0, 1)$**

$\Rightarrow [T]_{\beta'}^{\beta} = \begin{bmatrix} -1 & -1 \\ 1 & 2 \\ 4 & 7 \end{bmatrix}, \quad \text{where} \quad T_{11} = -1 = \langle T(1, 2), (1, 0, 0) \rangle, \quad T_{21} = 1 = \langle T(1, 2), (0, 1, 0) \rangle,$

$T_{31} = 4 = \langle T(1, 2), (0, 0, 1) \rangle, T_{12} = -1 = \langle T(2, 3), (1, 0, 0) \rangle, T_{22} = 2 = \langle T(2, 3), (0, 1, 0) \rangle,$  and  $T_{32} = 7 = \langle T(2, 3), (0, 0, 1) \rangle$

## 5-2 Gram-Schmidt Orthogonalization Process

**Gram-Schmidt Orthogonalization Process:** Let  $S=\{y_1, y_2, y_3, \dots, y_n\}$  be a linearly independent subset of  $V$  and  $S'=\{x_1, x_2, x_3, \dots, x_n\}$  be an orthogonal subset of  $V$ . Set  $x_1=y_1$ ,  $x_2=y_2 - \frac{\langle y_2, x_1 \rangle}{\langle x_1, x_1 \rangle} \cdot x_1$ ,

$$x_3=y_3 - \frac{\langle y_3, x_1 \rangle}{\langle x_1, x_1 \rangle} \cdot x_1 - \frac{\langle y_3, x_2 \rangle}{\langle x_2, x_2 \rangle} \cdot x_2, \dots, \text{ and } x_k=y_k - \sum_{j=1}^{k-1} \frac{\langle y_k, x_j \rangle}{\langle x_j, x_j \rangle} \cdot x_j. \text{ Then } \text{Span}(S)=\text{Span}(S').$$

**Ex.** For  $V=\mathbb{R}^3$ ,  $\beta=\{(1,1,0), (2,0,1), (2,2,1)\}$ , find an orthogonal basis for  $V$  by the Gram-Schmidt orthogonalization process. [2007 台科大電研]

$$\text{(Sol.) Let } x_1 = (1,1,0), x_2 = (2,0,1) - \frac{\langle (2,0,1), (1,1,0) \rangle}{\langle (1,1,0), (1,1,0) \rangle} \cdot (1,1,0) = (2,0,1) - \frac{2}{2}(1,1,0) = (1,-1,1)$$

$$x_3 = (2,2,1) - \frac{\langle (2,2,1), (1,1,0) \rangle}{\langle (1,1,0), (1,1,0) \rangle} \cdot (1,1,0) - \frac{\langle (2,2,1), (1,-1,1) \rangle}{\langle (1,-1,1), (1,-1,1) \rangle} \cdot (1,-1,1) = (-\frac{1}{3}, \frac{1}{3}, \frac{2}{3}).$$

Then  $\beta'=\{x_1, x_2, x_3\}$  is an orthogonal basis.

**Ex.** Let the vector space  $P_2$  have the inner product  $\langle p, q \rangle = \int_{-1}^1 p(x)q(x)dx$ . Apply the Gram-Schmidt process to transform the standard basis  $S=\{1, x, x^2\}$  into an orthonormal basis. [2005 北科大電腦通訊所]

$$\text{(Sol.) } S=\{1, x, x^2\}=\{y_1, y_2, y_3\}. \text{ Let } x_1=y_1=1, x_2=y_2 - \frac{\langle y_2, x_1 \rangle}{\langle x_1, x_1 \rangle} \cdot x_1 = x - \frac{\int_{-1}^1 x \cdot 1 dx}{\int_{-1}^1 1 \cdot 1 dx} \cdot 1 = x - 0 = x,$$

$$x_3=y_3 - \frac{\langle y_3, x_1 \rangle}{\langle x_1, x_1 \rangle} \cdot x_1 - \frac{\langle y_3, x_2 \rangle}{\langle x_2, x_2 \rangle} \cdot x_2 = x^2 - \frac{\int_{-1}^1 x^2 \cdot 1 dx}{\int_{-1}^1 1 \cdot 1 dx} \cdot 1 - \frac{\int_{-1}^1 x^2 \cdot x dx}{\int_{-1}^1 x \cdot x dx} \cdot x = x^2 - \frac{1}{3} \cdot 1 - 0 = x^2 - \frac{1}{3}.$$

$$\therefore \text{Orthogonal basis: } \{1, x, x^2 - \frac{1}{3}\} \Rightarrow \text{Orthonormal basis: } \left\{ \sqrt{\frac{1}{2}}, \sqrt{\frac{3}{2}}x, \sqrt{\frac{45}{8}}\left(x^2 - \frac{1}{3}\right) \right\}$$

**S<sub>perp</sub>:**  $S^\perp=\{x \in V: \langle x, y \rangle = 0 \text{ for all } y \in S\}$  is an orthogonal set of  $S$ .

**Ex.** For  $V=\mathbb{C}^3$ ,  $S=\text{Span}\{(1,0,i), (1,2,1)\}$ , compute  $S^\perp$ .

(Sol.) Suppose  $S^\perp=\text{Span}\{(a,b,c)\}$ ,

$$\begin{cases} \langle (a,b,c), (1,0,i) \rangle = 0 \Rightarrow a = -ci \\ \langle (a,b,c), (1,2,1) \rangle = 0 \Rightarrow b = -\frac{1-i}{2}c \end{cases}, S^\perp = \text{Span}\left\{(-i, -\frac{1-i}{2}, 1)\right\}$$

**Ex.** Show that (a)  $\{x_1, x_2, \dots, x_k\}$  is an ordered basis for  $W$  (subspace of  $V$ ) and  $\{x_1, x_2, \dots, x_k, x_{k+1}, \dots, x_n\}$  is an ordered basis  $V$ , then  $\{x_{k+1}, \dots, x_n\}$  is an order basis of for  $W^\perp$ .

(b)  $W$  is a subspace of  $V$ , then  $\dim(V)=\dim(W)+\dim(W^\perp)$ . [台大電研]

$$\text{(Proof) (a) } \forall x \in V \Leftrightarrow x = \sum_{i=1}^n \langle x, x_i \rangle x_i$$

If  $x \in W^\perp$ , then  $\langle x, x_i \rangle = 0$  for  $1 \leq i \leq k$ . Therefore,  $x = \sum_{i=k+1}^n \langle x, x_i \rangle x_i \in \text{Span}(\{x_{k+1}, \dots, x_n\})$

(b) According to (a),  $\dim(V)=n=k+(n-k)=\dim(W)+\dim(W^\perp)$

**Eg. Show that (a)  $(W_1 + W_2)^\perp = W_1^\perp \cap W_2^\perp$  and (b)  $(W_1 \cap W_2)^\perp = W_1^\perp + W_2^\perp$ . [1990, 1999 台大電研]**

$$\text{(Proof)} \quad (W_1 + W_2) \oplus (W_1 + W_2)^\perp = W_1 \oplus W_1^\perp = W_2 \oplus W_2^\perp = V$$

$$W_1 \text{ and } W_2 \subset W_1 + W_2, \therefore (W_1 + W_2)^\perp \subset W_1^\perp \text{ and } W_2^\perp \Rightarrow (W_1 + W_2)^\perp \subseteq W_1^\perp \cap W_2^\perp$$

$$\forall x \in W_1^\perp \cap W_2^\perp, \text{ then } \forall y \in W_1 \text{ and } z \in W_2 \Rightarrow \langle y, x \rangle = \langle z, x \rangle = 0$$

$$\therefore cy + dz \in W_1 + W_2, \langle cy + dz, x \rangle = c \langle y, x \rangle + d \langle z, x \rangle = 0$$

$$\Rightarrow x \in (W_1 + W_2)^\perp \Rightarrow W_1^\perp \cap W_2^\perp \subseteq (W_1 + W_2)^\perp, \therefore (W_1 + W_2)^\perp = W_1^\perp \cap W_2^\perp$$

$$\text{On the other hand, } (U_1 + U_2)^\perp = U_1^\perp \cap U_2^\perp \Rightarrow U_1 + U_2 = (U_1^\perp \cap U_2^\perp)^\perp$$

$$\text{Set } U_1 = W_1^\perp, U_2 = W_2^\perp \Rightarrow W_1^\perp + W_2^\perp = (W_1 \cap W_2)^\perp$$

### 5-3 Various Matrices

**Adjoint matrix:**  $A^*$  is the complex conjugate transpose of  $A$ . **Note:**  $A^* \neq \text{adj}(A)$  and  $(AB)^* = B^*A^*$ .

**Eg.** For  $A = \begin{bmatrix} 1-i & 2 \\ -3i & 4 \end{bmatrix}$ , then  $A^* = \begin{bmatrix} 1+i & 3i \\ 2 & 4 \end{bmatrix}$ .

**Eg.** For  $T: C^2 \rightarrow C^2$  by  $T(a_1, a_2) = (2ia_1 + 3a_2, a_1 - a_2)$ , if  $\beta$  is the standard ordered basis, find  $T^*(a_1, a_2)$ .

[1998 台大電研]

(Sol.)  $T(1,0) = (2i, 1) = 2i \cdot (1,0) + 1 \cdot (0,1)$  and  $T(0,1) = (3, -1) = 3 \cdot (1,0) + (-1) \cdot (0,1)$

$$\Rightarrow [T]_{\beta} = \begin{bmatrix} 2i & 3 \\ 1 & -1 \end{bmatrix} \Rightarrow [T^*]_{\beta} = \begin{bmatrix} -2i & 1 \\ 3 & -1 \end{bmatrix}, \therefore T^*(a_1, a_2) = (-2ia_1 + a_2, 3a_1 - a_2).$$

**Theorem For**  $A \in M_{m \times n}(F)$ ,  $x \in F^n$ ,  $y \in F^m$ , then  $\langle Ax, y \rangle = \langle x, A^*y \rangle$ .

(Note:  $\langle T(x), y \rangle = \langle x, T^*(y) \rangle$  for  $\forall x, y \in V, T: V \rightarrow V$ )

**Eg.**  $\left\langle \begin{bmatrix} 1 & 1 \\ -2i & 3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}, \begin{bmatrix} u \\ v \end{bmatrix} \right\rangle = \left\langle \begin{bmatrix} x+y \\ -2ix+3y \end{bmatrix}, \begin{bmatrix} u \\ v \end{bmatrix} \right\rangle = x\bar{u} + y\bar{u} - 2ix\bar{v} + 3y\bar{v} = \left\langle \begin{bmatrix} x \\ y \end{bmatrix}, \begin{bmatrix} 1 & 2i \\ 1 & 3 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} \right\rangle = \left\langle \begin{bmatrix} x \\ y \end{bmatrix}, \begin{bmatrix} u+2iv \\ u+3v \end{bmatrix} \right\rangle$

**Theorem (a)**  $A \in M_{m \times n}(F)$ ,  $\text{Rank}(A^*A) = \text{Rank}(A)$ . **(b)**  $A \in M_{m \times n}(F)$ , if  $\text{Rank}(A) = n$ , then  $A^*A$  is invertible.

(Proof) (a)  $A^*Ax = 0 \Leftrightarrow Ax = 0$

1. " $\Leftarrow$ ":  $Ax = 0$  implies  $A^*Ax = 0$

2. " $\Rightarrow$ ":  $0 = \langle A^*Ax, x \rangle_n = \langle Ax, A^*x \rangle_m = \langle Ax, Ax \rangle_m$ ,  $\therefore Ax = 0$

(b)  $A \in M_{m \times n}(F) \Rightarrow A^*A \in M_{n \times n}(F)$

$\text{Rank}(A^*A) = \text{Rank}(A) = n$ ,  $\therefore A^*A$  is invertible

**Orthogonal matrix A:**  $AA^t = A^tA = I$ . **Unitary matrix A:**  $AA^* = A^*A = I$ .

**Eg.**  $A = \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$ ,  $A^t = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$ ,  $AA^t = A^tA = I$ ,  $\therefore A$  is orthogonal.

**Eg.**  $A = \begin{bmatrix} i/\sqrt{2} & -i/\sqrt{2} \\ i/\sqrt{2} & i/\sqrt{2} \end{bmatrix}$ ,  $A^* = \begin{bmatrix} -i/\sqrt{2} & -i/\sqrt{2} \\ i/\sqrt{2} & -i/\sqrt{2} \end{bmatrix}$ ,  $AA^* = A^*A = I$ ,  $\therefore A$  is unitary.

**Normal Matrix A:**  $AA^* = A^*A$ .

**Eg.**  $A = \begin{bmatrix} i & -i \\ i & i \end{bmatrix}$ ,  $A^* = \begin{bmatrix} -i & -i \\ i & -i \end{bmatrix}$ ,  $AA^* = A^*A$ ,  $\therefore A$  is normal.

**Eg.  $T$  is a normal operator on  $V$ . Show that (a)  $\|T(x)\| = \|T^*(x)\|$ , (b)  $T(x) = \lambda x \Rightarrow T^*(x) = \bar{\lambda}x$ , (c)  $\lambda_1$  and  $\lambda_2$  are distinct eigenvalues of  $T$  with corresponding eigenvector  $x_1$  and  $x_2$ . Then  $x_1$  and  $x_2$  are orthogonal. [2012 台大電研]**

(Proof) (a)  $\|T(x)\|^2 = \langle T(x), T(x) \rangle = \langle T^*T(x), x \rangle = \langle TT^*(x), x \rangle = \langle T^*(x), T^*(x) \rangle = \|T^*(x)\|^2$

(b) Suppose  $T(x) = \lambda x$  for some  $x \in V$ . Let  $U = T - \lambda I$ ,

$$0 = \|U(x)\| = \|U^*(x)\| = \|(T^* - \bar{\lambda}I)(x)\|, \therefore T^*(x) = \bar{\lambda}x.$$

(c) Let  $T(x_1) = \lambda_1 x_1, T(x_2) = \lambda_2 x_2$ ,

$$\lambda_1 \langle x_1, x_2 \rangle = \langle \lambda_1 x_1, x_2 \rangle = \langle T(x_1), x_2 \rangle = \langle x_1, T^*(x_2) \rangle = \langle x_1, \bar{\lambda}_2 x_2 \rangle = \bar{\lambda}_2 \langle x_1, x_2 \rangle \quad \because \lambda_1 \neq \bar{\lambda}_2, \therefore \langle x_1, x_2 \rangle = 0.$$

**Self-adjoint (Hermitian) matrix  $A: A = A^*$ .**

**Eg.  $\begin{bmatrix} 1 & 2-5i \\ 2+5i & -3 \end{bmatrix}$  is a self-adjoint matrix.**

**(Note: A self-adjoint matrix is also a normal matrix, but a normal matrix may not be self-adjoint)**

**Theorem If  $A$  is a self-adjoint matrix, all eigenvalues of  $A$  are real. [2000 台大電研]**

(Proof)  $\lambda x = Ax = A^*x = \bar{\lambda}x$  for some  $x \neq 0, \lambda = \bar{\lambda} \Rightarrow \lambda$  is real.

**Gramian matrix  $A: \exists B \in M_{m \times n}(F)$  such that  $A = BB^t$  then  $A$  is called the Gramian matrix.**

**Theorem  $A$  is a Gramian matrix  $\Leftrightarrow \begin{cases} A \text{ is symmetric.} \\ \text{all eigenvalues of } A \text{ are nonnegative.} \end{cases}$**

(Proof) (1)  $A^t = (BB^t)^t = BB^t = A, \therefore A$  is symmetric. (2) For some  $x \neq 0$ , and  $Ax = \lambda x$ ,

$$\text{then } \langle Ax, x \rangle = \langle \lambda x, x \rangle = \lambda \langle x, x \rangle = \langle BB^t x, x \rangle = \langle B^t x, B^t x \rangle \geq 0, \because \langle x, x \rangle \geq 0, \therefore \lambda \geq 0$$

**Eg. Which of the following matrices is Hermitian? Which is normal?**

$$A = \begin{bmatrix} 1 & 2-i \\ 2+i & -1 \end{bmatrix} \quad B = \begin{bmatrix} \frac{i}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{i}{\sqrt{2}} \end{bmatrix} \quad C = \begin{bmatrix} 0 & i & 1 \\ i & 0 & -2+i \\ -1 & 2+i & 0 \end{bmatrix} \quad D = \begin{bmatrix} 3 & 1+i & i \\ 1-i & 1 & 3 \\ -i & 3 & 1 \end{bmatrix} \quad \text{[交大電子所]}$$

(Sol.)  $A = A^*$  and  $D = D^*, \therefore A$  and  $D$  are both Hermitian matrices,

$$CC^* = C^*C, BB^* = \begin{bmatrix} 1 & i \\ -i & 1 \end{bmatrix}, B^*B = \begin{bmatrix} 1 & -i \\ i & 1 \end{bmatrix}, \therefore C \text{ is normal but } B \text{ is not normal.}$$

## 5-4 Special Characteristics of Matrices

**Positive definite:** If  $\langle Ax, x \rangle > 0$  for all  $x \neq 0$ , then  $A$  is positive definite.

**Eg. If  $A$  is a positive-definite matrix, then all the eigenvalues of  $A$  are positive.**

(Proof) For some  $x \neq 0$  and  $Ax = \lambda x$ , positive definite:  $\langle Ax, x \rangle = \langle \lambda x, x \rangle = \lambda \langle x, x \rangle > 0 \Rightarrow \lambda > 0$ .

**Eg. Let  $A$  be a complex normal (or real symmetric)  $n \times n$  matrix with eigenvalues:  $\lambda_1, \lambda_2, \dots, \lambda_n$ ,**

**and then show that (a)  $tr(A) = \sum_{i=1}^n \lambda_i$  and  $tr(A^*A) = \sum_{i=1}^n |\lambda_i|^2$ , (b)  $det(A) = \prod_{i=1}^n \lambda_i$ . [1998 台大電研]**

(Proof) Let  $A = QDQ^{-1}$  and  $D$  be diagonal,  $tr(A) = tr(QDQ^{-1}) = tr(DQQ^{-1}) = tr(D) = \sum_{i=1}^n \lambda_i$

$det(A) = det(QDQ^{-1}) = det(Q)det(D)det(Q^{-1}) = det(Q)det(D)[det(Q)]^{-1} = det(D) = \prod_{i=1}^n \lambda_i$ .

**Eg. Let  $\lambda_i$  be the eigenvalues of  $A = \begin{bmatrix} 2 & 1 & -1 & 0 \\ 1 & 3 & 4 & 2 \\ -1 & 4 & 1 & 2 \\ 0 & 2 & 2 & 1 \end{bmatrix}$ ,  $1 \leq i \leq 4$ , (a)  $\sum_{i=1}^4 \lambda_i = ?$  (b)  $\prod_{i=1}^4 \lambda_i = ?$**

**(c) Is  $A$  positive definite? [台大機研]**

(Sol.)  $\because A$  is a real symmetric matrix,  $\therefore$  (a)  $\sum_{i=1}^4 \lambda_i = tr(A) = 2+3+1+1=7$ , (b)  $\prod_{i=1}^4 \lambda_i = det(A) = 10$ ,

(c) Method 1:  $\exists \lambda < 0 \Rightarrow A$  is not positive definite.

Method 2:  $det([2])=2 > 0$  and  $det\left(\begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}\right)=5 > 0$ , but  $det\left(\begin{bmatrix} 2 & 1 & -1 \\ 1 & 3 & 4 \\ -1 & 4 & 1 \end{bmatrix}\right)=-38 < 0 \Rightarrow$  Not positive definite.

**$A$  is unitarily equivalent to  $B$ :**  $\exists P$  fulfills  $B = P^*AP$  and  $P^{-1} = P^*$ .

**Theorem (a)  $A$  is complex normal  $\Leftrightarrow A$  is unitary equivalent to a diagonal matrix. (b)  $A$  is real symmetric  $\Leftrightarrow A$  is orthogonally equivalent to a real diagonal matrix.**

(Proof)  $AA^* = (PDP^*)(PDP^*)^* = PDP^*PD^*P^* = PDD^*P^*$

$$A = PDP^*, A^*A = (PDP^*)^*(PDP^*) = PD^*DP^*. \because DD^* = D^*D, \therefore A^*A = AA^*$$

**Projection:**  $V = W_1 \oplus W_2$ ,  $x_1 \in W_1$ , and  $x_2 \in W_2$ ,  $x = x_1 + x_2$ . If  $T(x) = x_1$ , then  $T$  is projection on  $W_1$ . That is,  $R(T) = W_1 = \{x: T(x) = x\}$  and  $N(T) = W_2$ . (Note:  $T^2 = T$  if  $T$  is a projection.)

**Orthogonal Projection:** If  $R(T)^\perp = N(T)$  and  $N(T)^\perp = R(T)$  for  $T: V \rightarrow V$  be a projection.

**Theorem  $T$  is a linear operator on  $V$ . Then  $T$  is an orthogonal projection.  $\Leftrightarrow T^2 = T = T^*$ .**

**Eg. Let  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be a linear operator. For any  $z \in \mathbb{R}^2$ ,  $T(z) = p$ , where  $p$  is the projection of  $z$  on the line  $x=y$ . Find  $[T]$ . [交大電信所]**

(Sol.) The orthonormal basis of line  $x=y$  is  $\left\{ \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\}$  and the standard basis of  $\mathbb{R}^2$  is  $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$ .

Inner product:  $\left\langle \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\rangle = \frac{1}{\sqrt{2}}$ , so the orthogonal projection of  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  onto the line  $x=y$  is

$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Inner product:  $\left\langle \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\rangle = \frac{1}{\sqrt{2}}$ , so the orthogonal projection of  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$  onto the line  $x=y$  is

$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1/2 \\ 1/2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad \therefore [T] = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}, \text{ and we have } [T]^2 = [T] = [T]^*$$

**Eg. Find the orthogonal projection of the vector  $v = \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}$  onto the subspace  $S = \text{Span}\left(\begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}\right)$ .**

[交大電子所]

(Sol.) Transform the orthogonal basis  $\left\{ \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$  into the orthonormal one:  $\left\{ \begin{bmatrix} 0 \\ -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}, \begin{bmatrix} 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\}$ .

$\left\langle \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}, \begin{bmatrix} 0 \\ -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\rangle = -\sqrt{2}$ , so the orthogonal projection of the vector  $v = \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}$  onto  $\begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}$  is

$$-\sqrt{2} \begin{bmatrix} 0 \\ -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}.$$



$\left\langle \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\rangle = -\sqrt{2}$ , so the orthogonal projection of the vector  $v = \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}$  onto  $\begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$  is

$$-\sqrt{2} \begin{bmatrix} 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \\ -1 \end{bmatrix}.$$

$$\therefore \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix} - (-\sqrt{2}) \begin{bmatrix} 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}, \quad \therefore v = \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} + \begin{bmatrix} 0 \\ -1 \\ -1 \end{bmatrix}$$

and  $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\} \perp S = \text{Span} \left( \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right)$

**Rotation:**  $\exists$  orthogonal basis  $\beta = \{x_1, x_2\}$  for  $W$ , and  $\exists \theta$  is real and fulfills  $T(x_1) = x_1 \cos \theta + x_2 \sin \theta$ ,  $T(x_2) = -x_1 \sin \theta + x_2 \cos \theta$ , and  $T(y) = y, \forall y \in W^\perp$ , where  $W$  is a 2-dimensional subspace. And then  $T$  is a rotation of  $W$  about  $W^\perp$ , where  $W^\perp$  is the axis of rotation.

**Reflection:**  $T(x) = -x, \forall x \in W$  and  $T(y) = y, \forall y \in W^\perp$ . And then  $T$  is a reflection of  $W$  about  $W^\perp$ , where  $W$  is a 1-dimensional subspace.

**Eg.  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  by  $T(a, b) = (-a, b)$ . Let  $W = \text{Span}(\{e_1\})$  and  $W^\perp = \text{Span}(\{e_2\})$ . It is reflection of  $\mathbb{R}^2$  about  $W^\perp$  (the  $y$ -axis).**

**Theorem  $T$  is an orthogonal operator on a 2-dimensional real inner product space. If  $\det(T) = 1$ , then  $T$  is a rotation. If  $\det(T) = -1$ , then  $T$  is a reflection.**

(Proof) By definition,  $[T]_\beta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$  is a rotation,  $\therefore \det(T) = 1$ .

Choose  $\gamma = \{z_1, z_2, \dots\}$  be an orthogonal basis for  $V$  and  $z_1 \in W$ ,

$$[T]_\gamma = \begin{bmatrix} -1 & 0 & \dots & \dots & 0 \\ 0 & 1 & \ddots & & \vdots \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ \vdots & & \ddots & \ddots & 0 \\ 0 & \dots & \dots & 0 & 1 \end{bmatrix} \Rightarrow \det(T) = -1.$$

**Theorem** The composition of a reflection and a rotation is a reflection.

(Proof)  $\det(T_1 T_2) = \det(T_1) \det(T_2) = -1$

**Spectral Theorem** If  $T$  is  $\begin{cases} \text{normal on } V \text{ over } F = \mathbb{C} \\ \text{self-adjoint on } V \text{ over } F = \mathbb{R} \end{cases}$ , and  $\lambda_1, \dots, \lambda_k$  are distinct eigenvalues of

$T$ . Let  $W_i$  be the eigenspace of  $T$  corresponding to  $\lambda_i$ , and  $T_i$  be an orthogonal projection on  $W_i$ .

( $T_i(x) = x_i \in W_i$ ). Then (a)  $V = W_1 \oplus W_2 \oplus W_3 \cdots \oplus W_k$ , (b)  $W_i^\perp = \sum_{j \neq i} W_j$  (direct sum), (c)  $T_i T_j = \delta_{ij} T_i$ ,

(d)  $I = T_1 + T_2 + \cdots + T_k$ , (e)  $T = \lambda_1 T_1 + \lambda_2 T_2 + \cdots + \lambda_k T_k$ .

(Proof) (c) Let  $x \neq 0$  and  $T_j(x) = x_j$ , then  $T_i T_j(x) = T_i(x_j) = \begin{cases} 0, i \neq j \\ x_i, i = j \end{cases} = \delta_{ij} T_i(x) \Rightarrow T_i T_j = \delta_{ij} T_i$

(d)  $x = Ix = x_1 + \dots + x_k = T_1(x) + T_2(x) + \dots + T_k(x) = (T_1 + T_2 + \dots + T_k)(x)$ ,  $\therefore I = T_1 + T_2 + \dots + T_k$ .

(e) For  $x \in V$ , then  $x = x_1 + \dots + x_k$ , where  $x_i \in W_i$ ,

$T(x) = T(x_1) + \dots + T(x_k) = \lambda_1 x_1 + \dots + \lambda_k x_k = \lambda_1 T_1(x) + \dots + \lambda_k T_k(x) = (\lambda_1 T_1 + \dots + \lambda_k T_k)(x)$ ,

$\therefore T = \lambda_1 T_1 + \dots + \lambda_k T_k$

**Theorem** (a)  $F = \mathbb{C}$ ,  $T$  is a unitary operator  $\Leftrightarrow T^*$  is normal and  $|\lambda| = 1$  for all eigenvalues. (b)  $T = -T^* \Leftrightarrow$  each  $\lambda$  is pure imaginary. (c)  $T$  is a projection  $\Leftrightarrow$  each  $\lambda$  is either 0 or 1.

## 5-5 Bilinear Forms

**Bilinear form,  $H$ :** If (a)  $H(ax_1 + bx_2, y) = aH(x_1, y) + bH(x_2, y)$  and (b)  $H(x, ay_1 + by_2) = aH(x, y_1) + bH(x, y_2)$  for  $x_1, y_1, x_2, y_2, x, y \in V, a, b \in F$ .

**Theorem** For each  $H \in B(F^n)$ ,  $\exists ! A \in M_{n \times n}(F)$  fulfills  $H(x, y) = x^t A y$  for all  $x, y \in F^n$ .

**Eg. Define**  $H: \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$  by  $H\left(\begin{bmatrix} a_1 \\ a_2 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}\right) = 2a_1 b_1 + 3a_1 b_2 + 4a_2 b_1 - a_2 b_2$  for all  $\begin{bmatrix} a_1 \\ a_2 \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \in \mathbb{R}^2$ ,

then let  $x = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}, y = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}, A = \begin{bmatrix} 2 & 3 \\ 4 & -1 \end{bmatrix} \Rightarrow H(x, y) = x^t A y = [a_1 \ a_2] \begin{bmatrix} 2 & 3 \\ 4 & -1 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$ .

**Symmetric bilinear form:**  $H(x, y) = H(y, x)$  for all  $x, y \in V$ .

**Quadratic form:**  $K: V \rightarrow F$  if  $\exists H \in B(V)$  fulfills  $K(x) = H(x, x) = x^t A x$ .

**Eg. Let**  $f(t_1, t_2, t_3) = 2t_1^2 - t_2^2 + 6t_1 t_2 - 4t_2 t_3 = K\left(\begin{bmatrix} t_1 \\ t_2 \\ t_3 \end{bmatrix}\right)$ , and then  $A = \begin{bmatrix} 2 & 3 & 0 \\ 3 & -1 & -2 \\ 0 & -2 & 0 \end{bmatrix}$

$\Rightarrow f(t_1, t_2, t_3) = [t_1 \ t_2 \ t_3] \begin{bmatrix} 2 & 3 & 0 \\ 3 & -1 & -2 \\ 0 & -2 & 0 \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \\ t_3 \end{bmatrix}$  is a quadratic form.

**Eg. There are two bases  $(x_1, y_1)$  and  $(x_2, y_2)$  in  $R^2$ . If the equations of the same ellipse represented by two distinct bases are described as follows, respectively:  $2x_1^2 - 4x_1y_1 + 5y_1^2 - 36 = 0$  and  $x_2^2 + 6y_2^2 - 36 = 0$ . Please find the transformation matrix between these two bases. [2004 台大電研]**

(Sol.) Let  $K \begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = 2x_1^2 + 5y_1^2 - 4x_1y_1$ ,  $A = \begin{bmatrix} 2 & -2 \\ -2 & 5 \end{bmatrix}$ ,  $\det \begin{pmatrix} 2-\lambda & -2 \\ -2 & 5-\lambda \end{pmatrix} = (\lambda-1)(\lambda-6)$ ,  $\lambda = 1, 6$

$$\lambda_1 = 1 \Rightarrow [A - \lambda_1 I] \cdot \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 1 & -2 \\ -2 & 4 \end{bmatrix} \cdot \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = 0 \Rightarrow \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = a \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \quad a \neq 0,$$

$$\lambda_2 = 6 \Rightarrow [A - \lambda_2 I] \cdot \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} -4 & -2 \\ -2 & -1 \end{bmatrix} \cdot \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = 0 \Rightarrow \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = b \begin{bmatrix} 1 \\ -2 \end{bmatrix}, \quad b \neq 0$$

$$Q = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \end{bmatrix} \text{ fulfills } Q^{-1}AQ = Q^T A Q = \begin{bmatrix} 1 & 0 \\ 0 & 6 \end{bmatrix} \text{ for } \beta = \left\{ \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ -2 \end{bmatrix} \right\}.$$

$$\therefore \begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \end{bmatrix} \cdot \begin{bmatrix} x_2 \\ y_2 \end{bmatrix} \Rightarrow K \begin{bmatrix} x_2 \\ y_2 \end{bmatrix} = x_2^2 + 6y_2^2 \Rightarrow 2x_1^2 - 4x_1y_1 + 5y_1^2 - 36 = 0 \Rightarrow x_2^2 + 6y_2^2 - 36 = 0$$

**Eg.  $[t_1, t_2, t_3]^T \in R^3$ ,  $3t_1^2 + 3t_2^2 + 3t_3^2 - 2t_1t_3 + 2\sqrt{2}(t_1 + t_3) + 1 = 0$ . Find a basis  $\beta$  for  $R^3$  such that the above equation is simplified.**

(Sol.) Let  $K \begin{bmatrix} t_1 \\ t_2 \\ t_3 \end{bmatrix} = 3t_1^2 + 3t_2^2 + 3t_3^2 - 2t_1t_3$ , then  $A = \begin{bmatrix} 3 & 0 & -1 \\ 0 & 3 & 0 \\ -1 & 0 & 3 \end{bmatrix}$ . It is found that an orthogonal

$$\text{matrix } Q = \begin{bmatrix} 0 & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ 1 & 0 & 0 \\ 0 & \frac{-1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \text{ fulfills } Q^T A Q = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 2 \end{bmatrix} \text{ for } \beta = \left\{ \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 0 \\ -1 \end{bmatrix} \right\}.$$

$$\therefore \begin{bmatrix} t_1 \\ t_2 \\ t_3 \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ 1 & 0 & 0 \\ 0 & \frac{-1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \cdot \begin{bmatrix} s_1 \\ s_2 \\ s_3 \end{bmatrix} \Rightarrow K(x) = 3s_1^2 + 4s_2^2 + 2s_3^2$$

$$\Rightarrow 3t_1^2 + 3t_2^2 + 3t_3^2 - 2t_1t_3 + 2\sqrt{2}(t_1 + t_3) + 1 = 0 \Rightarrow 3s_1^2 + 4s_2^2 + 2s_3^2 - 4s_3 + 1 = 0.$$