

ROB 501 HW #5

$$1(a) \quad A_3 = \left[ \begin{array}{ccc|c} 3 & 10 & 0 & \\ 0 & 2 & 4 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] \xrightarrow{\text{Row } 2 \rightarrow 2R_2 + R_1} \left[ \begin{array}{ccc|c} 3 & 10 & 0 & 0 \\ 0 & 2 & 4 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] \xrightarrow{\text{Row } 2 \rightarrow R_2 - 2R_3} \left[ \begin{array}{ccc|c} 3 & 10 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right]$$

$$A - \lambda I = \left[ \begin{array}{ccc} 3 - \lambda & 10 & 0 \\ 0 & 2 - \lambda & 4 \\ 0 & 0 & 1 - \lambda \end{array} \right]$$

$$\det(A - \lambda I) = 0$$

$$\Rightarrow (3-\lambda)[(2-\lambda)(1-\lambda) - 0] - 10[0-0] + 0(0)$$

$$\Rightarrow \det(A - \lambda I) = (3-\lambda)(2-\lambda)(1-\lambda) = 0.$$

$$\lambda = 3; \quad \lambda = 2; \quad \lambda = 1$$

$$Av = \lambda v$$

$$(A - \lambda)v = 0$$

$$\Rightarrow \left[ \begin{array}{ccc} 3-\lambda & 10 & 0 \\ 0 & 2-\lambda & 4 \\ 0 & 0 & 1-\lambda \end{array} \right] \left[ \begin{array}{c} x_1 \\ x_2 \\ x_3 \end{array} \right] = 0$$

$$0 = 3x_1 + 10x_2$$

$$0 = 2x_2 + 4x_3$$

$$0 = x_3$$

$$0 = 8x_2$$

Case - 1

$\lambda = 3$

$$\left[ \begin{array}{ccc|c} 0 & 10 & 0 & \lambda_1 \\ 0 & -1 & 4 & x_2 \\ 0 & 0 & -2 & x_3 \end{array} \right] \sim \left[ \begin{array}{ccc|c} 0 & 10 & 0 & 0 \\ 0 & -1 & 4 & 0 \\ 0 & 0 & -2 & 0 \end{array} \right] = 0$$

$$\begin{aligned} 10x_2 &= 0 & \text{--- } (1) \\ -x_2 + 4x_3 &= 0 & \text{--- } (2) \\ -2x_3 &= 0 & \text{--- } (3) \end{aligned}$$

$x_2 = x_3 = 0$

$x_1$  can be any arbitrary value. let  $x_1 = 1$

$(0)0 + (0-0)0 + (0-(\lambda-1)(\lambda-5))(\lambda-8) = 0$

$$\Rightarrow x_1^1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = (\lambda-1)(\lambda-5)(\lambda-8) \quad (3)$$

Case - 2

$\lambda = 2$

$$\left[ \begin{array}{ccc|c} 1 & 10 & 0 & 0 = \lambda(\lambda-1) \\ 0 & 0 & 4x & 0 \\ 0 & 0 & 1x & 0 \end{array} \right] \sim \left[ \begin{array}{ccc|c} x_1 & & & 0 \\ x_2 & 0 & & 0 \\ x_3 & 0 & & 0 \end{array} \right] = 0$$

$\Rightarrow x_1 + 10x_2 = 0$

$4x_3 = 0$

$x_3 = 0$

$x_2 = 0$

$$x_1 = -10x_2 \quad \text{and} \quad x_2 = 1 \quad ; \quad x_1 = -10.$$

$$\Rightarrow \begin{bmatrix} -10 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 4 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

Case - 3

$$\lambda = 1 \quad \text{from } \det(A - \lambda I) = 0$$

$$\begin{bmatrix} 1 & 10 & 0 \\ 0 & 1 & 4 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0$$

$$x_1 + 10x_2 = 0$$

$$x_2 + 4x_3 = 0$$

$$x_2 = -4x_3$$

$$x_1 + 10(-4x_3) = 0$$

$$\Rightarrow x_1 = 40x_3$$

$$x_1 = 20x_3 \quad (1)$$

$$\begin{bmatrix} 20 \\ -4 \\ 1 \end{bmatrix} = x_3 \quad (5)$$

$\therefore$  The sign vectors are

$$0: \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad \begin{bmatrix} -10 \\ 1 \\ 0 \end{bmatrix}, \quad \begin{bmatrix} 20 \\ -4 \\ 1 \end{bmatrix}$$

To find if these vectors are linearly independent let's consider the following vectors.

$$x^1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, x^2 = \begin{bmatrix} -10 \\ 1 \\ 0 \end{bmatrix}, x^3 = \begin{bmatrix} 20 \\ -4 \\ 1 \end{bmatrix}$$

$$a_1 x^1 + a_2 x^2 + a_3 x^3 = 0$$

$a_1 = a_2 = a_3 = 0$  (must satisfy this condition)

if  $x^1, x^2 + x^3$  have to be linearly independent:

$$a_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + a_2 \begin{bmatrix} -10 \\ 1 \\ 0 \end{bmatrix} + a_3 \begin{bmatrix} 20 \\ -4 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \cancel{a_1} + a_1 - 10a_2 + 20a_3 = 0 \quad \text{--- (1)}$$

$$0 + a_2 - 4a_3 = 0 \quad \text{--- (2)}$$

$$a_3 = 0 \quad \text{--- (3)}$$

from eqn (1), (2) & (3)

$$a_1 = a_2 = a_3 = 0$$

$\Rightarrow$  The e-vectors are linearly independent.

$$(b) A_4 = \begin{bmatrix} 5 & 1 & 1 \\ 0 & 5 & 2 \\ 0 & 0 & 2 \end{bmatrix}$$

$$|A_4 - \lambda I| = 0 \Leftrightarrow \begin{vmatrix} 5-\lambda & 1 & 1 \\ 0 & 5-\lambda & 2 \\ 0 & 0 & 2-\lambda \end{vmatrix} = 0$$

$$(5-\lambda)(5+\lambda)(2-\lambda) = 0$$

$$\therefore \lambda_1 = 5, \quad \lambda_2 = 2.$$

$\therefore$  the eigen values are repeated.

Case - 1

$$0 = 0 + \lambda_1 = 5 \Rightarrow (0)$$

$$\begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 3 \\ 0 & 0 & 2-5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 3 \\ 0 & 0 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0$$

$$x_2 + x_3 = 0$$

$$3x_2 = 0$$

$$-3x_3 = 0$$

$x_3 = 0; x_2 = 0; \lambda_1 \rightarrow$  can be any arbitrary value

$$\text{let } x_1 = 1$$

$$\Rightarrow 0\text{-vector } \vec{x}^1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \quad \vec{x}^3 = \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix}$$

$0 = (A - 5I) \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$  satisfies the eqn

Case - 2

$$\lambda_2 = 2$$

$$0 = (A - 2I)^2$$

$$\begin{bmatrix} 5 & 1 & 1 \\ 0 & 3 & 3 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0^2 \Rightarrow$$

$$x^2 = \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}$$

$$3x_1 + x_2 + x_3 = 0$$

$$3x_2 + 3x_3 = 0 \Rightarrow x_2 = -x_3$$

Vectors

$\therefore$  The Eigen values are  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}$

Note these vectors are not linearly independent.

$$A = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & -1 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow 1(0) + (-1)(0) + 0 = 0$$

$$\text{Det}(A) = 0 \quad \{ \text{Thus linearly dependent} \}$$

$\therefore$  They cannot form a basis for  $\mathbb{R}^3$ .

(2) To show two square Matrix A & B are said to be similar if there exists an invertible matrix P.

S.T.  $B = P^{-1}AP$ , if A & B are similar then they have the same characteristic eq?

Basically the problem is asking us to show.

$$|A - \lambda I| = |B - \lambda I|$$

lets consider  $\det(\lambda I - B) = 0$

then lets assume  $B = P^{-1}AP$

$$\det(\lambda I - P^{-1}AP) = 0$$

$$\det(\lambda I; P P^{-1} - P A P^{-1}) = 0$$

$$\det(P^{-1}(\lambda I P - AP)) = 0$$

$$\det(P^{-1}P(\lambda I - A)) = 0$$

$$P^{-1}P = I$$

$$\Rightarrow \det(\lambda(\lambda I + A)) = 0$$

$$\Rightarrow \det(\lambda^2 I - \lambda A) = 0 \quad \{ (\lambda I)(I) = \lambda I \}$$

$$\Rightarrow \det(\lambda^2 I - A) = 0 \quad \left\{ \begin{array}{l} \text{since } \lambda^2 I = (\lambda I)^2 \\ I \text{ is the identity matrix} \end{array} \right.$$

$\lambda^2 I = A^2$  as  $I$  is the identity matrix

$\therefore$  we have proved that  $\det(\lambda I - B) = \det(\lambda I - A)$

$\Rightarrow$  the characteristic polynomial of  $B$  is the same as that of  $A$ , since eigen values are the roots of the characteristic "eqn". Similar matrices have the same eigen values

$$(3) \quad A_3 = \begin{bmatrix} 3 & 10 & 0 \\ 0 & 1 & 4 \\ 0 & 0 & 1 \end{bmatrix}$$

The eigen values of this matrix  $A_3$  is  $\lambda_1 = 3, \lambda_2 = 2, \lambda_3 = 1$

The eigen values can be represented as a diagonal matrix  $\Lambda = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

in order to show  $A_3$  is similar to  $\Lambda$

$$\text{we have to show } A_3 = P^{-1}\Lambda P \quad \text{or} \quad \Lambda = P^{-1}A_3 P$$

$$P = [x^1 \mid x^2 \mid x^3] = \begin{bmatrix} 1 & -10 & 20 \\ 0 & 1 & -4 \\ 0 & 0 & 1 \end{bmatrix}$$

$\Lambda = P^{-1}AP$  can be re-written as

$$PA = AP$$

constant terms

$$P_A = \begin{bmatrix} 1 & -10 & 20 \\ 0 & 1 & -4 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad P = 9^{-1}$$

$\therefore P_A^{-1} = \begin{bmatrix} 1 & 10 & -20 \\ 0 & 1 & 4 \\ 0 & 0 & 1 \end{bmatrix}$

$$PA^{-1}P = P^{-1}A^{-1}P = \begin{bmatrix} 3 & -20 & 20 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{--- ①}$$

$(A - 2I)P = (A - 2I)^{-1}$  will bring you to

Want to find what is left after removing 0 from 20 and 1 from 1

$$PA^{-1}P = \begin{bmatrix} 3 & 10 & 0 \\ 0 & 2 & 4 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & -10 & 20 \\ 0 & 1 & -4 \\ 0 & 0 & 1 \end{bmatrix}$$

$$A_{SP} = \begin{bmatrix} 3 & -30 + 10 & 60 - 40 \\ 0 & 2 & 0 \end{bmatrix}$$

$$PA^{-1}P = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{--- ②}$$

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I$$

$$\therefore ① = ② \Rightarrow PA^{-1}P = A_{SP}$$

$$\Rightarrow A = P^{-1}A_{SP}P \quad \text{as } A \text{ & } A_{SP} \text{ are}$$

similar matrices.

4(b) The  $\hat{x}$ -hat values are:  $0.205$ ,  $0.284$ ,  $0.171$

$$-0.0370$$

$$2.2138$$

$$2.8901$$

$$-11.2271$$

$$7.6928$$

$$-1.5396$$

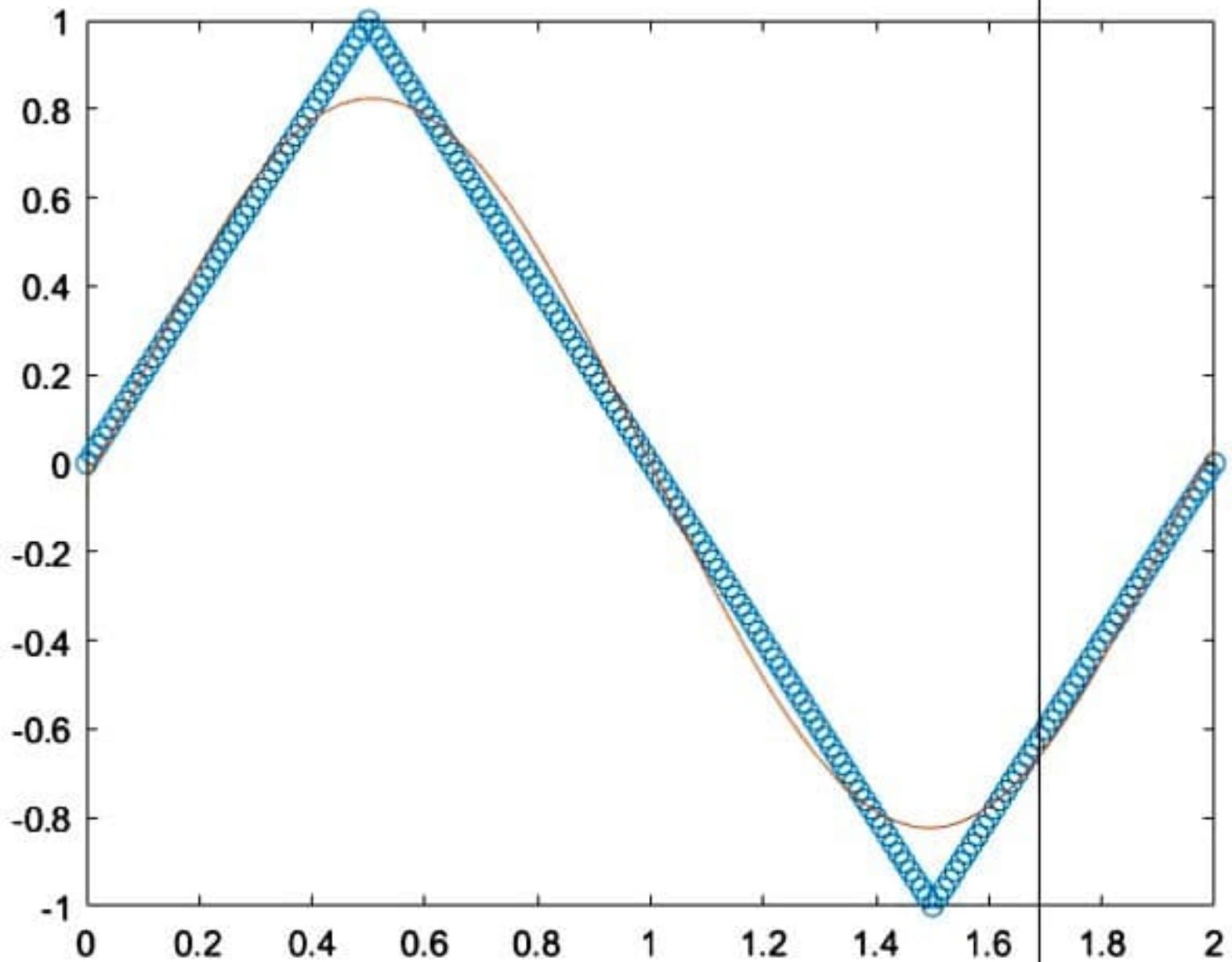
4(c) The  $\hat{x}$ -hat values are:  $0.8106$ ,  $0.0000$ ,  $-0.0301$ ,  $0.0000$ ,  $0.0325$

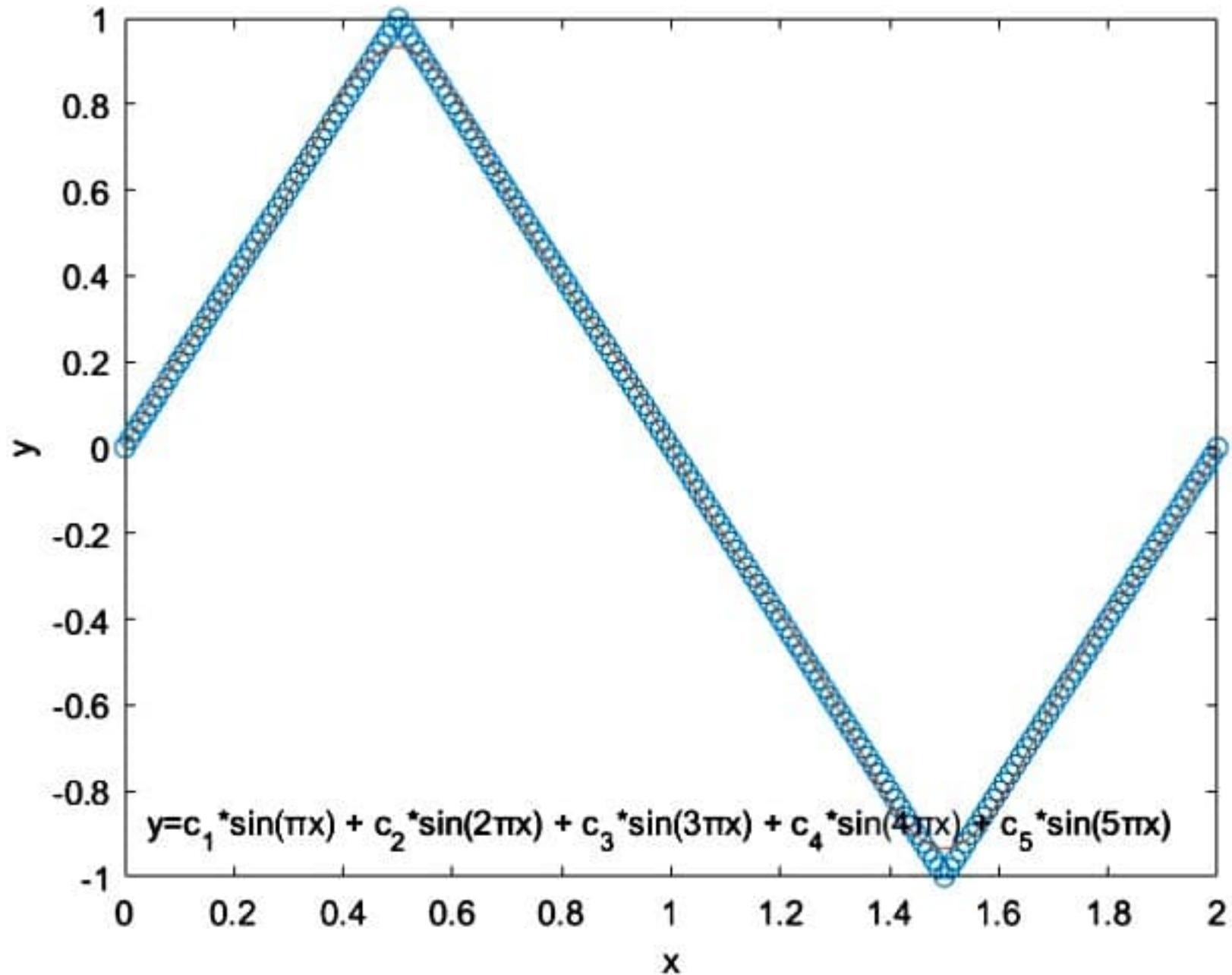
$$0.0000$$

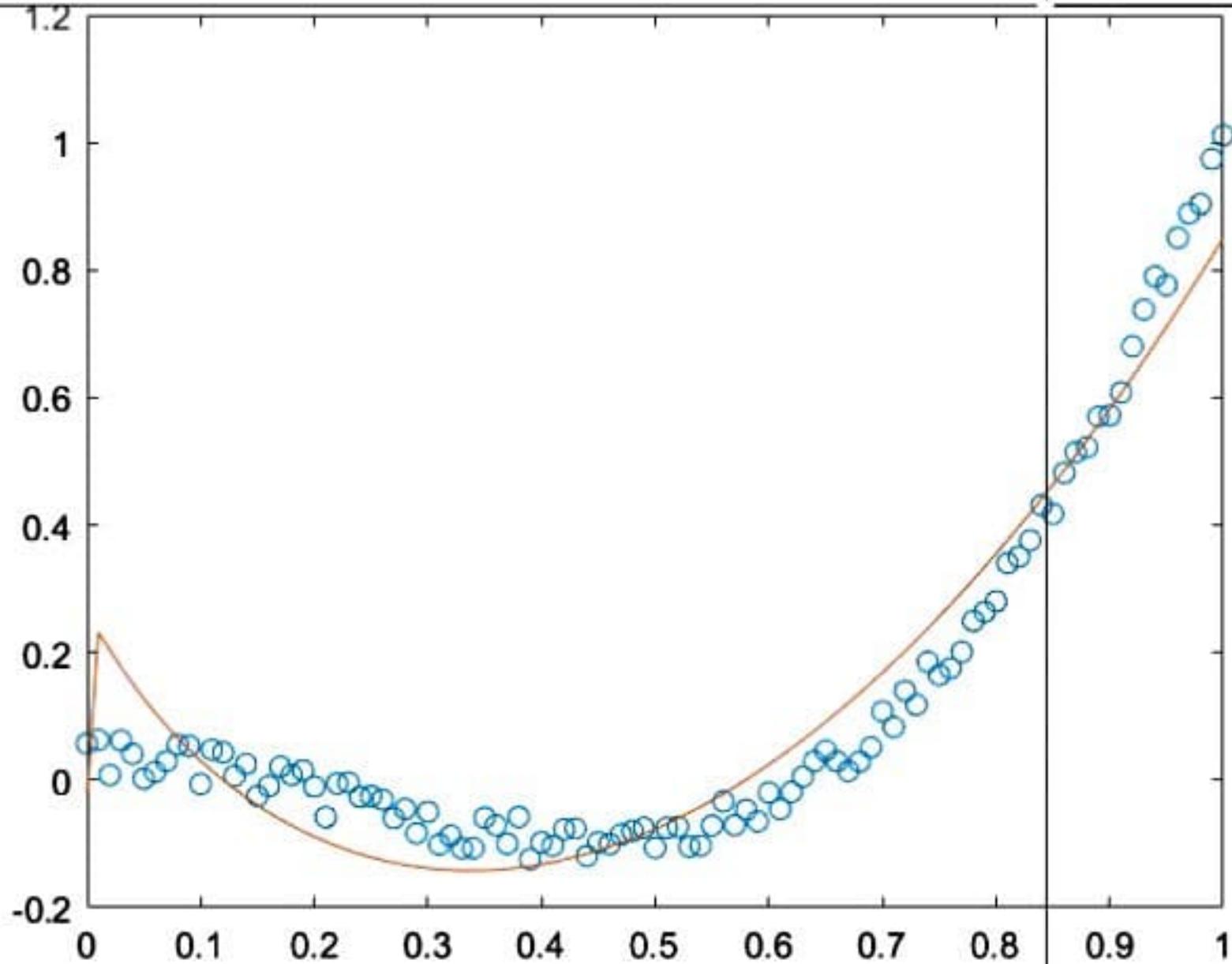
$$-0.0301$$

$$0.0000$$

$$0.0325$$







(6) There are two definitions of the inner product.

$$\text{Def - 1 } \langle x, y \rangle = \bar{x}^T y$$

$$\text{Def - 2 } \langle x, y \rangle = \bar{x}^T y.$$

To show both these definitions verify the following properties

(1) Conjugate symmetry

$$\langle x, y \rangle = \langle \bar{y}, x \rangle$$

$$(2) \langle x, (ay + bz) \rangle = a\langle x, y \rangle + b\langle x, z \rangle \quad (\text{linearity})$$

$$(3) \langle x, x \rangle \geq 0 \quad \& \quad \langle x, x \rangle = 0 \quad \text{if } x = 0 \quad (\text{definiteness})$$

$$\text{Def - 1 } \langle x, y \rangle = \bar{x}^T y \quad \dots \quad (1)$$

(1) Conjugate symmetry.

$$\langle y, x \rangle = \bar{y}^T \bar{x}$$

$$\langle \bar{y}, \bar{x} \rangle = \bar{\bar{y}}^T \bar{\bar{x}} \Rightarrow \bar{y}^T \bar{\bar{x}} = \bar{\bar{y}}^T \bar{x} \quad (\text{using the property of complex conjugates } \bar{A}^T = \bar{A}^T \neq \bar{\bar{z}} = z \text{ for any complex number } z)$$

$$\Rightarrow \langle \bar{y}, \bar{x} \rangle = \bar{\bar{y}}^T \bar{x}. \quad \dots \quad (2)$$

$$\text{Now to show } \bar{x}^T \bar{y} = \langle \bar{y}^T \bar{x} \rangle = \langle \bar{x}, \bar{y} \rangle$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$x^T = [x_1, x_2, \dots, x_n]$$

$$x^T \bar{y} = [x_1, x_2, x_3, \dots, x_n] \begin{bmatrix} \bar{y}_1 \\ \bar{y}_2 \\ \vdots \\ \bar{y}_n \end{bmatrix} \quad (1)$$

$$x^T \bar{y} = x_1 \bar{y}_1 + x_2 \bar{y}_2 + x_3 \bar{y}_3 + \dots + x_n \bar{y}_n \quad (1)$$

Now calculating

$$\bar{y}^T x = [\bar{y}_1, \bar{y}_2, \bar{y}_3, \dots, \bar{y}_n] \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} \quad (2)$$

$$\bar{y}^T x = [\bar{y}_1, \bar{y}_2, \bar{y}_3, \dots, \bar{y}_n] \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} \quad (2)$$

$$= \bar{y}_1 x_1 + \bar{y}_2 x_2 + \bar{y}_3 x_3 + \dots + \bar{y}_n x_n \quad (2)$$

W.K.T complex multiplication is commutative

$$\Rightarrow a, b \in \mathbb{C}$$

$$a \times b = b \times a$$

$$\Rightarrow (1) = (2) \cdot \{ \text{as } y_i x_i \bar{y}_i = y_i \bar{y}_i \quad [i=1, 2, \dots, n] \}$$

$$\therefore x^T \bar{y} = \bar{y}^T x \quad \Rightarrow \sum_{i=1}^n x_i \bar{y}_i = \sum_{i=1}^n \bar{y}_i x_i$$

$$\Rightarrow \langle x, y \rangle = \langle \bar{y}, x \rangle$$

(2) Linearity

$$\begin{aligned}\langle ax + bz, y \rangle &= (ax + bz)^T y = ax^T y + bz^T y \\ &= a \langle x, y \rangle + b \langle z, y \rangle.\end{aligned}$$

(3) Positive definiteness:-

$$\langle x, x \rangle = x^T x = \sum_{i=1}^n |x_i|^2 \geq 0$$

$$\& \langle x, x \rangle = 0 \Leftrightarrow x = 0.$$

||| the same has to be proved for Def - 2

$$\langle x, y \rangle = x^T y \quad \text{--- (1)}$$

$$\langle y, x \rangle = \bar{y}^T x \quad \langle \bar{y}, x \rangle = \bar{y}^T x = \bar{y}^T \bar{x} = \bar{y}^T \bar{x} = y^T \bar{x} \quad \text{--- (2)}$$

From ① & ② it is evident that  $\bar{x}^T y = \bar{y}^T \bar{x}$

To prove this:-

$$\bar{x} = \begin{bmatrix} \bar{x}_1 \\ \bar{x}_2 \\ \vdots \\ \bar{x}_n \end{bmatrix} \quad \bar{x}^T = \begin{bmatrix} \bar{x}_1 & \bar{x}_2 & \bar{x}_3 & \cdots & \bar{x}_n \end{bmatrix}$$
$$\Rightarrow \bar{x}^T y = \bar{x}_1 y_1 + \bar{x}_2 y_2 + \cdots + \bar{x}_n y_n \quad \text{--- } ③$$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \quad y^T = \begin{bmatrix} y_1 & y_2 & \cdots & y_n \end{bmatrix}$$

$$y^T = [y_1 \ y_2 \ \cdots \ y_n]$$

$$\bar{x} = \begin{bmatrix} \bar{x}_1 \\ \bar{x}_2 \\ \vdots \\ \bar{x}_n \end{bmatrix}$$

$$y^T \bar{x} = y_1 \bar{x}_1 + y_2 \bar{x}_2 + \cdots + y_n \bar{x}_n \quad \text{--- } ④$$

from ③ & ④ we have proved that

$$\bar{x}^T y = y^T \bar{x}$$

(2) Linearity  $\quad (a, b \in \mathbb{C})$   $\quad \langle ax + bz, y \rangle = \langle ax, y \rangle + \langle bz, y \rangle$

$$\begin{aligned}\langle ax + bz, y \rangle &= (\bar{ax} + \bar{bz})^T y = (a\bar{x} + b\bar{z})^T y \\ &= a\bar{x}^T y + b\bar{z}^T y = a\langle x, y \rangle + b\langle z, y \rangle.\end{aligned}$$

$\therefore$  def-2 satisfies linearity.

(3) +ve Definiteness

$$\langle x, x \rangle = \bar{x}^T x = \sum_{i=1}^n |x_i|^2 \geq 0$$

$$\& \langle x, x \rangle = 0 \Leftrightarrow x = 0.$$

$\therefore$  both definitions satisfy the properties of an inner product in the complex vector space  $\mathbb{C}^n$ .

$\Rightarrow$  both definitions are valid inner products when the field  $F = \mathbb{C}$ .

(7) First let's show these are orthogonal. Vectors are with

$$P_0(x) = 1, \quad P_1(x) = x, \quad P_2(x) = \frac{1}{2}(3x^2 - 1); \quad P_3(x) = \frac{1}{2}(5x^3 - 3x)$$

$$\langle P_0(x), P_3(x) \rangle = \int_{-1}^1 (1) \left[ \frac{1}{2}(5x^3 - 3x) \right] dx \quad \text{using } (1)$$

$$\begin{aligned}&\frac{1}{2} \int_{-1}^1 [5x^3 - 3x] dx = \frac{1}{2} \left[ \frac{5x^4}{4} \Big|_{-1}^1 - \frac{3x^2}{2} \Big|_{-1}^1 \right] \\ &\text{using } x = -x \Rightarrow -1 \quad \text{and } x = 1 \quad \text{and } x = 0\end{aligned}$$

$$\begin{aligned}&= \frac{1}{2} \left[ \frac{5[(1)^4 - (-1)^4]}{4} - \frac{3}{2} [(1)^2 - (-1)^2] \right] \\ &= \frac{1}{2} \left[ \frac{5(1^4 - 1^4)}{4} - \frac{3}{2} [(1)^2 - (-1)^2] \right]\end{aligned}$$

or  $\langle P_0(x), P_3(x) \rangle = 0 \quad \therefore$  These basis are orthogonal to each other.

$$\langle P_1(x), P_2(x) \rangle = \int_{-1}^1 P_1(x) P_2(x) dx$$

$$= \int_{-1}^1 (x) \left( \frac{1}{2} 3x^2 - 1 \right) dx$$

$$= \frac{1}{2} \int_{-1}^1 (3x^3 - x) dx$$

$$= \frac{1}{2} \left[ \frac{3x^4}{4} - \frac{x^2}{2} \right]_{-1}^1$$

$$= \frac{1}{2} \left[ \left( \frac{3}{4}(1)^4 - \frac{1}{2}(1)^2 \right) - \left( \frac{3}{4}(-1)^4 - \frac{1}{2}(-1)^2 \right) \right]$$

$$= 0$$

*and hence showing that below two vectors are orthogonal to each other.*

*i.e. The vectors  $P_1(x)$  &  $P_2(x)$  are orthogonal to each other.*

Now we have to show that these polynomials form the basis of  $P_3$ . [Note:- we have proved that these polynomials are orthogonal to each other]

(1) prove the polynomials are linearly independent  $\Rightarrow$  prove they are the basis

lets consider  $x_0 ; x_1 ; x_2 ; x_3$  the basis

The linear combination is given as.

$$x_0 P_0 + x_1 P_1 + x_2 P_2 + x_3 P_3 = 0$$

if  $x_0 = x_1 = x_2 = x_3 = 0$  Then these

$\Rightarrow$  (1)  $\Rightarrow$  polynomials are linearly independent

$$x_0(1) + x_1(x) + x_2 \left( \frac{1}{2} (3x^2 - 1) \right) + x_3 \left( \frac{1}{2} (5x^3 - 3x) \right) = 0$$

$$x_0(1) + x_1(x) + \frac{x_2}{2} (3x^2) - \frac{x_2}{2} (1) + x_3 \left( \frac{5}{2} x^3 \right) - x_3 (3x) = 0$$

$$\left(-3\alpha_4 + \alpha_1\right)x + \left(\frac{5\alpha_4}{4}\right)x^3 + \left(\frac{3}{2}\alpha_2\right)x^2 + \left(\alpha_0 - \frac{\alpha_2}{2}\right) = 0$$

$\therefore$  from ① we get  $-3\alpha_4 + \alpha_1 = 0 \quad \text{--- } ②$

$$5\alpha_4 = 0 \quad \text{--- } ③$$

$$\frac{3}{2}\alpha_2 = 0 \quad \text{--- } ④$$

$$\alpha_0 - \frac{\alpha_2}{2} = 0 \quad \text{--- } ⑤$$

$$\alpha_4 = \alpha_2 = 0 \quad \{ \text{from } ③ \text{ & } ④ \}$$

$$\text{if } \alpha_4 = 0 ; \alpha_1 = 0 \quad \{ \text{from } ② \}$$

$$\text{if } \alpha_2 = 0 ; \alpha_0 = 0 \quad \{ \text{from } ⑤ \}$$

$\therefore$  we have proved that  $\alpha_1 = \alpha_2 = \alpha_3 = \alpha_4 = 0$ .

hence these polynomials are linearly independent.

Each of these polynomials have to belong to  $P_3$

$$P_0 = 0x^3 + 0x^2 + 0x + 1 \Rightarrow \text{highest order} = 3 \Rightarrow \in P_3.$$

$$P_1 = 0x^3 + 0x^2 + x + 0 \Rightarrow \text{highest order} = 3 \Rightarrow \in P_3$$

$$P_2 = 0x^3 + \frac{1}{2}(3)x^2 + 0 - \frac{1}{2} \Rightarrow \text{highest order} = 3 \Rightarrow \in P_3$$

$$P_3 = \frac{5}{2}x^3 + 0x^2 + \frac{-3}{2}x + 0 \Rightarrow \text{highest order} = 3 \Rightarrow \in P_3.$$

(2)  $\therefore$  all of these polynomials belong to  $P_3$

(3) Span  $\{P_0, P_1, P_2, P_3\}$  contains all the elements of  $IP_3$ .

Let's consider the natural basis of  $IP_3$ .

$$\Rightarrow \{1, x, x^2, x^3\}$$

The highest order in (4) & are linearly independent.

$$\text{i.e. } \dim(IP_3) = 4$$

& the basis we have

$\left\{1, x, \frac{1}{2}(3x^2 - 1), \frac{1}{2}(5x^3 - 3x)\right\}$  are the higher order  
in (u) & are linearly independent  $\rightarrow$  As shown in (1)

$\Rightarrow \text{Span}\{P_0, P_1, P_2, P_3\}$  contains all the elements of  $IP_3$

$$8(b) \quad A = \left[ \begin{array}{ccccc|c} & 6 & 0 & 0 & 0 & 0 \\ 0.5 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0.5 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right] = [I_4 \quad 0.5I_2]$$

$$B = \left[ \begin{array}{ccccc} 3 & & & & \\ 0 & 0 & 0 & 0 & 0 \\ 2 & & & & \\ 0 & & & & \\ 1 & & & & \end{array} \right] \quad C = 0.25 \quad D = B^T$$

$$\Rightarrow D = [3 \ 0 \ 2 \ 0 \ 1]$$

$$BCD = \left[ \begin{array}{c} 3 \\ 0 \\ 0 \\ 2 \\ 1 \end{array} \right] \left[ \begin{array}{ccccc} 3 & 0 & 2 & 0 & 1 \end{array} \right]_{1 \times 5}$$

$$= [3 \cdot 3 + 0 \cdot 0 + 2 \cdot 0 + 0 \cdot 2 + 1 \cdot 1] = 10$$

$$BD = \begin{bmatrix} 9 & 0 & 6 & 0 & 3 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 6 & 0 & 4 & 0 & 2 \\ 0 & 0 & 0 & 10 & 0 \end{bmatrix}$$

$$0.25[BD] = \begin{bmatrix} 9/4 & 0 & 6/4 & 0 & 3/4 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 6/4 & 0 & 4/4 & 0 & 2/4 \\ 0 & 0 & 0 & 10 & 0 \end{bmatrix}$$

$$A + 0.25[BD]$$

$$= \begin{bmatrix} 0.5 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0.5 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} + \begin{bmatrix} 9/4 & 0 & 6/4 & 0 & 3/4 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 6/4 & 0 & 4/4 & 0 & 2/4 \\ 0 & 0 & 0 & 10 & 0 \end{bmatrix}$$

$$A + 0.25[BD] = \begin{bmatrix} 9/4 + 0.5 & 0 & 6/4 & 0 & 3/4 \\ 0 & 1 & 0 & 0 & 0 \\ 6/4 & 0 & 1+1 & 0 & 2/4 \\ 0 & 0 & 0 & 0.5 & 0 \\ 3/4 & 0 & 2/4 & 0 & 1/4+1 \end{bmatrix}$$

$$9(0.25 + 1) = 9 + 9 = 18$$

$$9(0.25 + 1) = 9 + 9 = 18$$

$$A + 0.25[BD] = \begin{bmatrix} 0.25[0.25 + 1]A & 0 & 0.25 \\ 0[0.25 + 1]A + 1 & 0 & 0 \\ 0.25[0.25 + 1]A & 0 & 0.25 \\ 0.25[0.25 + 1]A + 1 & 0 & 0.25 \\ 0.25[0.25 + 1]A & 0 & 0.25 \end{bmatrix}$$

$$0.25[0.25 + 1]A = 0.25[0.25 + 1]A$$

$$0.25[0.25 + 1]A = 0.25[0.25 + 1]A$$

$$\textcircled{2} \quad A + 0.25[BD] = A + 0.25[0.25 + 1]A = 1.25A$$

$$\textcircled{3} \quad [A + 0.25[BD]]^T = \begin{bmatrix} 1.25 & 0 & 0.444 & 0 & -0.2222 \end{bmatrix}$$

$$\textcircled{4} \quad -0.2222[0.25 + 1]A + 1.000A = 0 \quad 0 \quad 0$$

$$\textcircled{5} \quad -0.2222[0.25 + 1]A + 1.000A = 0.8519 \quad 0 \quad -0.0741$$

$$\textcircled{6} \quad -0.2222[0.25 + 1]A + 1.000A = 2.000 \quad 0$$

$$\textcircled{7} \quad -0.2222[0.25 + 1]A + 1.000A = -0.0741 \quad 0 \quad 0.9630$$

(a) Identity - I

$$(I+P)^{-1} = (I+P)^{-1}(I+P-P)$$

$$= 1 - (I+P)^{-1}P \quad \dots \quad (1)$$

Identity 2

$$P + PQP = P(I + QP) = (I + PQ)P$$

$$(I + PQ)^{-1}P = P(I + QP)^{-1} \quad \dots \quad (2)$$

$$(A + BCD)^{-1} = [A(I + A^{-1}BCD)]^{-1}$$

$$= [I + A^{-1}BCD]^{-1}A^{-1}$$

$$= [I - [I - A^{-1}BCD]^{-1}A^{-1}BCD]A^{-1} \quad \text{using (1)}$$

$$= A^{-1} - (I + A^{-1}BCD)^{-1}A^{-1}BCDA^{-1}$$

Repeatedly using (2) in sequence now produces.

$$(A + BCD)^{-1} = A^{-1} - (I + A^{-1}BCD)^{-1}A^{-1}BCDA^{-1} \quad \dots \quad (3)$$

$$= A^{-1} - A^{-1}B(I + BCD A^{-1}B)^{-1}CDA^{-1} \quad \dots \quad (4)$$

$$= A^{-1} - A^{-1}BC(I + DA^{-1}BC)^{-1}DA^{-1} \quad \dots \quad (5)$$

$$= A^{-1} - A^{-1}BCD(I + A^{-1}BCD)^{-1}A^{-1} \quad \dots \quad (6)$$

$$= A^{-1} - A^{-1}BCD(I + A^{-1}BCD)^{-1}A^{-1} \quad \dots \quad (7)$$

$$= A^{-1} - A^{-1}BCD(I + BCPA^{-1})^{-1} \quad \dots \quad (8)$$

if (c) is also invertible from (5)

$$\begin{aligned} (A + BCD)^{-1} &= A^{-1} - A^{-1}B \left[ I + (CDA^{-1}B) \right]^{-1} CDA^{-1} \\ &= A^{-1} - A^{-1}B \left[ C^{-1} + DA^{-1}B \right]^{-1} DA^{-1} \quad \text{--- (9)} \end{aligned}$$

(9) (a)  $f(x)$ ,  $\|\cdot\|$  is a norm if

- (1)  $\|x\| > 0 \Leftrightarrow x \in X \setminus \{0\}$  ( $\|x\| = 0 \Leftrightarrow x = 0$ )
- (2)  $\|x+y\| \leq \|x\| + \|y\| \quad \forall x, y \in X$
- (3)  $\|ax\| = |a| \|x\| \quad \forall x \in X, a \in F$ .

(1) Since  $A$  is positive definite

$$\forall x \in \mathbb{R}^n, x \neq 0 \Rightarrow x^T A x > 0 \rightarrow \text{by def of } A$$

$\Downarrow$

$$\begin{aligned} \forall x \in \mathbb{R}^n, x \neq 0 \Rightarrow (x^T A_0)^{1/2} > 0 \\ \text{If } x=0, \text{ then } (0^T A_0)^{1/2} = 0 \Rightarrow g(x) = 0 \Leftrightarrow x=0 \end{aligned}$$

(2) P:  $g(x, y) = x^T A y$  is an inner product

$$A \in \mathbb{R}^{n \times n}, x \in \mathbb{R}^n \Rightarrow x^T A y \in \mathbb{R}$$

$$x^T A y = (x^T A y)^T = y^T (A^T x) = y^T A^T x$$

given  $A$  is symmetric  $\Rightarrow A^T = A$

$\Downarrow$

$$x^T A y = y^T A^T x = y^T A x$$

$$g(x, y) = x^T A y = g(y, x) = y^T A x.$$

$\Downarrow$

$$g(x, y) = g(y, x)$$

$$= ag(x, z) + bg(y, z)$$

$$= g(ax+by, z) = ag(x, z) + bg(y, z)$$

(3)  $\left. \begin{array}{l} g(x, x) > 0 \quad \forall x \in X, x \neq 0 \\ g(x, x) = 0 \iff x = 0 \end{array} \right\} \rightarrow \text{shown in (1)}$

All in All we have shown

$$1. g(x, y) = g(y, x) \rightarrow \text{for real } x-y$$

$$2. g(ax+by, z) = ag(x, z) + bg(y, z)$$

$$3. g(x, x) > 0 \quad \forall x \in X, x \neq 0$$

$$\times 3 \quad g(x, x) = 0 \iff x = 0$$

4.

$g(x, y) = (x^T A y)^{1/2}$  is an Norm by definition.

if we replace A with  $2A$  the new function becomes:-

$$g(x) = (\sqrt{(2A)x})^{1/2} = \sqrt{2} (x^T A x)^{1/2} = \sqrt{2} f(x).$$

So we are scaling the original function by  $\sqrt{2}$

i.e. only the magnitude is changed

$\Rightarrow$  All the three properties satisfy

$$\|x\| > 0 \quad \forall x \in X / \{0\}$$

$$\|x+y\| \leq \|x\| + \|y\| \quad \forall x, y \in X$$

$$\|ax\| = |a| \|x\| \quad \forall a \in \mathbb{R}, x \in X$$

$\therefore$  The function is still a Norm.

$$x^T A^T B = (x^T B)^T A = B^T A^T = (B^T A)^T$$

$$(x^T B)^T = (Bx)^T$$

$$g(x) \quad f(x+hy) = (x+hy)^T A (x+hy)^T = (x^T A x + y^T A x + x^T A y + y^T A y)^T$$

continued

continued

Since  $y^T A x$  is a scalar  $y^T A x = (y^T A x)^T = (A x)^T y = x^T A y$

$$f(x,y) = (x^T A x + x^T A y + x^T A y + y^T A y)^{r_2}$$

$$f(\text{new}) = \left( \mathbf{v}^T \mathbf{A} \mathbf{x} + 2\mathbf{v}^T \mathbf{A} \mathbf{y} + \mathbf{y}^T \mathbf{A} \mathbf{y} \right) v_L$$

The lemma from above tells us that  $x^*y$  is an inner product

$\Rightarrow$  using Cauchy-Schwarz Inequality for inner products

$$(g(x,y)) \leq g^{v_2}(y,x) \cdot g^{v_2}(y,y)$$

$$(x^T A y) \leq (x^T A x)^{1/2} \cdot (y^T A y)^{1/2}$$

$$(x^T A x + 2x^T A y + y^T A y)^{1/2} \leq (x^T A x + 2(b^T A x))^{1/2} (y^T A y)^{1/2} +$$

$y^T A y)^{1/2}$

F(x,y)

g(b) Yes, it is a norm, called induced norm. Here we only prove that it satisfies triangular inequality since non-negativity and scalability is trivial.  $\forall A, B \in \mathcal{X}$ .

$$\text{defn } f_V(A+B) = \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \frac{\| (A+B)x \|_V}{\| x \|_V} = \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \frac{\| Ax + Bx \|_V}{\| x \|_V}$$

(continued from previous page)  $\leq \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \| Ax \|_V + \| Bx \|_V$  by triangular inequality in normed space

$$= (\mathbb{R}^n, \mathbb{R}, \| \cdot \|_V)$$

$$\leq \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \| Ax \|_V + \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \| Bx \|_V = f_V(A) + f_V(B)$$

for an induced norm  $\| Ax \|_V = \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \| Ax \|_V$

$$(a+bi) = (a+i) + bi$$

$\| Ax \|_V = \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \frac{\| Ax + 0x \|_V}{\| x \|_V}$  if we let  $y = 0x$ , then

$$\| y \|_V = 1 \Rightarrow \| Ax \|_V = \sup_{\substack{x \in \mathbb{R}^n \\ x \neq 0}} \| Ax \|_V$$

$$f_2(A) =$$

$$f_{\infty}(A) =$$

$\Rightarrow$  Using the function  $f_V(A) = \sup_{\|x\|=1} (Ax)$

1. for  $\|A\|_1, V=1$ , so we use the 1-vector norm  
for both  $\|Ax\|_1 \geq \|x\|_1$ ,

$$\|A\|_1 = \sup_{\|x\|=1} \sum |a_{ij}| \cdot (\text{sum of absolute values of elements in } x)$$

$$\|A\|_1 = \sup_{\|x\|=1} \sum |a_{ij}| v_j \cdot (\text{sum of absolute values of elements in } A^v)$$

The sup occurs when  $x$  is chosen to have the largest column sum of  $A$ .

$$\|A\|_1 = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}|$$

This means  $\|A\|_1$  is the max. absolute column sum of  $A$ .

The  $\infty$ -norm of Matrix  $A$  is defined as the max. absolute row-sum of  $A$ .

1. for  $\|A\|_1^\infty, V=\infty$ , so we use the  $\infty$ -vector norm for both  $\|Ax\|_\infty \geq \|x\|_\infty$

$$2. \|x\|_\infty^\infty = \max_i \|x\|_1 \cdot (\text{max. absolute value of elements in } x)$$

$$3. \|Ax\|_\infty^\infty = \max_i \left\| \sum a_{ij} v_j \right\|_1 \cdot (\text{max. absolute value of elements in } x)$$

$\Rightarrow \|A\|_\infty^\infty$  is the max. absolute row sum of  $A$ .