

# Linear Algebra



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Abstract—This book provides solved examples on Linear Algebra.

#### 1 June 2019

1.1. Consider the vector space  $\mathbb{P}_n$  of real polynomials in x of degree  $\leq n$ . Define

$$T: \mathbb{P}_2 \to \mathbb{P}_3 \tag{1.1.1}$$

by

$$(Tf)(x) = \int_0^x f(t) dt + f'(x). \tag{1.1.2}$$

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Then find the matrix representation of T with respect to the bases

$$\{1, x, x^2\}$$
 and  $\{1, x, x^2, x^3\}$  (1.1.3)

1.2. Let  $P_A(x)$  denote the characteristic polynomial of a matrix A. Then for which of the following matrices is

$$P_A(x) - P_{A^{-1}}(x) \tag{1.2.1}$$

a constant?

a) 
$$\begin{pmatrix} 3 & 3 \\ 2 & 4 \end{pmatrix}$$
 c)  $\begin{pmatrix} 3 & 2 \\ 4 & 3 \end{pmatrix}$  b)  $\begin{pmatrix} 4 & 3 \\ 2 & 3 \end{pmatrix}$  d)  $\begin{pmatrix} 2 & 3 \\ 3 & 4 \end{pmatrix}$ 

1.3. Which of the following matrices is not diagonalizable over  $\mathbb{R}$ ?

a) 
$$\begin{pmatrix} 2 & 0 & 1 \\ 0 & 3 & 0 \\ 0 & 0 & 2 \end{pmatrix}$$
 c)  $\begin{pmatrix} 2 & 0 & 1 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{pmatrix}$  b)  $\begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$  d)  $\begin{pmatrix} 1 & -1 \\ 2 & 4 \end{pmatrix}$ 

1.4. What is the rank of the following matrix?

$$\begin{pmatrix}
1 & 1 & 1 & 1 & 1 \\
1 & 2 & 2 & 2 & 2 \\
1 & 2 & 3 & 3 & 3 \\
1 & 2 & 3 & 4 & 4 \\
1 & 2 & 3 & 4 & 5
\end{pmatrix}$$
(1.4.1)

- 1.5. Let V denote the vector space of real valued continuous functions on the close interval [0, 1]. Let W be the subspace of V spanned by  $\{\sin x, \cos x, \tan x\}$ . Find the dimension of W over  $\mathbb{R}$ .
- 1.6. Let V be the vector space of polynomials in the variable t of degree at most 2 over  $\mathbb{R}$ . An inner product on V is defined by

$$f^T g = \int_0^1 f(t)g(t) dt, \quad f, g \in V.$$
 (1.6.1)

Let

$$W = span \left\{ 1 - t^2, 1 + t^2 \right\}$$
 (1.6.2)

and  $W^{\perp}$  be the orthogonal complement of W in V. Which of the following conditions is satisfied for all  $h \in W^{\perp}$ ?

- a) h is an even function
- b) h is an odd function
- c) h(t) = 0 has a real solution
- d) h(0) = 0
- 1.7. Consider solving the following system by Jacobi iteration scheme

$$\begin{pmatrix} 1 & 2m & -2m \\ n & 1 & n \\ 2m & 2m & 1 \end{pmatrix} (x) = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$$
 (1.7.1)

where  $m, n \in \mathbb{Z}$ . With any initial vector, the 1.12. Let scheme converges provided m, n satisfy

- a) m + n = 3
- c) m < n
- b) m > n
- d) m = n
- 1.8. Consider a Markov Chain with state space  $\{0, 1, 2, 3, 4\}$  and transition matrix

$$P = \begin{array}{cccccc}
0 & 1 & 2 & 3 & 4 \\
0 & 1 & 0 & 0 & 0 & 0 \\
1 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 & 0 \\
0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 \\
0 & 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\
4 & 0 & 0 & 0 & 0 & 1
\end{array}$$
(1.8.1)

Then find

$$\lim_{n \to \infty} p_{23}^{(n)} \tag{1.8.2}$$

1.9. Let  $L(\mathbb{R})^n$  be the space of  $\mathbb{R}$ -linear maps from  $\mathbb{R}^n$  to  $\mathbb{R}^n$ . If Ker(T) denotes the kernel of T then which of the following are true?

- a) There exists  $T \in L(\mathbb{R}^5)$  {0} such that Range(T) = Ker(T)
- b) There does not exist  $T \in L(\mathbb{R}^5)$  {0} such that Range(T) = Ker(T)
- c) There exists  $T \in L(\mathbb{R}^6)$  {0} such that Range(T) = Ker(T)
- d) There does not exist  $T \in L(\mathbb{R}^6)$  {0} such that Range(T) = Ker(T)
- 1.10. Let V be a finite dimensional vector space over  $\mathbb{R}$  and  $T:V\to V$  be a linear map. Can you always write  $T = T_2 \circ T_1$  for some linear maps

$$T_1: V \to W, T: W \to V,$$
 (1.10.1)

where W is some finite dimensional vector space such that

- a) both  $T_1$  and  $T_2$  are onto
- b) both  $T_1$  and  $T_2$  are one to one
- c)  $T_1$  is onto,  $T_2$  is one to one
- d)  $T_1$  is one to one,  $T_2$  is onto
- 1.11. Let  $A = |a_{ij}|$  be a  $3 \times 3$  complex matrix. Identify the correct statements

a) 
$$det\left[\left(-1\right)^{i+j}a_{ij}\right] = det(A)$$

b) 
$$\det \left| (-1)^{i+j} a_{ij} \right| = -\det(A)$$

a) 
$$det \left[ (-1)^{i+j} a_{ij} \right] = det(A)$$
  
b)  $det \left[ (-1)^{i+j} a_{ij} \right] = -det(A)$   
c)  $det \left[ (\sqrt{-1})^{i+j} a_{ij} \right] = det(A)$ 

d) 
$$det\left[\left(\sqrt{-1}\right)^{i+j}a_{ij}\right] = -det(A)$$

$$p(x) = a_0 + a_1 x + \dots + a_n x^n$$
 (1.12.1)

be a non-constant polynomial of degree  $n \ge 1$ . Consider the polynomial

$$q(x) = \int_0^x p(t) dt, r(x) = \frac{d}{dx} p(x)$$
 (1.12.2)

Let V denote the real vector space of all polynomials in x. Then which of the following are true?

- a) q and r are linearly independent in V
- b) q and r are linearly dependent in V
- c)  $x^n$  belongs to the linear span of q and r
- d)  $x^{n+1}$  belongs to the linear span of q and r.
- 1.13. Let  $M_n(\mathbb{R})$  be the ring of  $n \times n$  matrices over  $\mathbb{R}$ . Which of the following are true for every  $n \geq 2$ ?
  - a) there exist matrices  $A, B \in M_n(\mathbb{R})$  such that  $AB - BA = I_n$ , where  $I_n$  denotes the identity matrix.

- b) If  $A, B \in M_n(\mathbb{R})$  and AB = BA, then A is diagonalisable over  $\mathbb{R}$  if and only if B is diagonalisable over  $\mathbb{R}$ .
- c) If  $A, B \in M_n(\mathbb{R})$ , then AB and BA have the same minimal polynomial.
- d) If  $A, B \in M_n(\mathbb{R})$ , then AB and BA have the same eigenvalues in  $\mathbb{R}$ .
- 1.14. Consider a matrix

$$A = [a_{ij}], 1 \le i, j \le 5$$
 (1.14.1)

such that

$$a_{ij} = \frac{1}{n_i + n_i + 1}, \quad n_i, n_j \in \mathbb{N}$$
 (1.14.2)

Then in which of the following cases A is a positive definite matrix?

- a)  $n_i = 1 \forall i = 1, 2, 3, 4, 5$ .
- b)  $n_1 < n_2 < \cdots < n_5$ .
- c)  $n_1 = n_2 = \cdots = n_5$ .
- d)  $n_1 > n_2 > \cdots > n_5$ .
- 1.15. For a nonzero  $w \in \mathbb{R}^n$ , define

$$T_w: \mathbb{R}^n \to \mathbb{R}^n \tag{1.15.1}$$

by

$$T_w = v - \frac{2v^T w}{w^T w} w, \quad v \in \mathbb{R}^n$$
 (1.15.2)

Which of the following are true?

- a)  $det(T_w) = 1$
- b)  $T_w(v_1)_w^T(v_2) = v_1^T v_2 \forall v_1, v_2 \in \mathbb{R}^n$
- $c) T_w = T_w^{-1}$
- $d) T_{2w} = 2T_w$
- 1.16. Consider the matrix

$$A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \tag{1.16.1}$$

over the field Q of rationals. Which of the following matrices are of the form  $P^{T}AP$  for suitable  $2 \times 2$  invertible matrix P over  $\mathbb{Q}$ ?

a) 
$$\begin{pmatrix} 2 & 0 \\ 0 & -2 \end{pmatrix}$$
 c)  $\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$   
b)  $\begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$  d)  $\begin{pmatrix} 3 & 4 \\ 4 & 5 \end{pmatrix}$ 

1.17. Consider a Markov Chain with state space

 $\{0, 1, 2\}$  and transition matrix

$$P = \begin{array}{ccc} 0 & 1 & 2 \\ 0 \begin{pmatrix} \frac{1}{4} & \frac{5}{8} & \frac{1}{8} \\ \frac{1}{4} & 0 & \frac{3}{4} \\ 2 \begin{pmatrix} \frac{1}{2} & \frac{3}{8} & \frac{1}{8} \end{pmatrix} \end{array}$$
(1.17.1)

Then which of the following are true?

- a)  $\lim_{n\to\infty} p_{12}^{(n)} = 0$ b)  $\lim_{n\to\infty} p_{12}^{(n)} = \lim_{n\to\infty} p_{21}^{(n)}$ c)  $\lim_{n\to\infty} p_{22}^{(n)} = \frac{1}{8}$ d)  $\lim_{n\to\infty} p_{21}^{(n)} = \frac{1}{3}$

# 2 December 2018

2.1. Consider the subspaces  $W_1$  and  $W_2$  of  $\mathbb{R}^3$  given

$$W_1 = \{ \mathbf{x} \in \mathbb{R}^3 : (1 \ 1 \ 1) \mathbf{x} = 0 \}$$
 (2.1.1)

$$W_2 = \{ \mathbf{x} \in \mathbb{R}^3 : (1 -1 1) \mathbf{x} = 0 \}.$$
 (2.1.2)

If  $W \subseteq \mathbb{R}^3$ , such that

- a)  $W \cap W_2 = \operatorname{span} \left\{ \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$
- b)  $\{W \cap W_1\} \perp \{W \cap W_2\}$ .

then

a) 
$$W = \operatorname{span} \left\{ \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$$

b) 
$$W = \operatorname{span} \left\{ \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} \right\}$$

c) 
$$W = \operatorname{span} \left\{ \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$$

d) 
$$W = \operatorname{span} \left\{ \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \right\}$$

2.2. Let

$$C = \left\{ \begin{pmatrix} 1\\2 \end{pmatrix}, \begin{pmatrix} 2\\1 \end{pmatrix} \right\} \tag{2.2.1}$$

be a basis of  $\mathbb{R}^2$  and

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x + y \\ x - 2y \end{pmatrix}. \tag{2.2.2}$$

If T [C] represents the matrix of T with respect to the basis C then which among the following is true?

a) 
$$T[C] = \begin{pmatrix} -3 & -2 \\ 3 & 1 \end{pmatrix}$$
  
b)  $T[C] = \begin{pmatrix} 3 & -2 \\ -3 & 1 \end{pmatrix}$   
c)  $T[C] = \begin{pmatrix} -3 & -1 \\ 3 & 2 \end{pmatrix}$   
d)  $T[C] = \begin{pmatrix} 3 & -1 \\ -3 & 2 \end{pmatrix}$ 

2.3. Let  $W_1 = \{ \mathbf{x} \in \mathbb{R}^4 : \}$ 

$$(1 1 1 0) \mathbf{x} = 0 (2.3.1)$$

$$(0 2 0 1) \mathbf{x} = 0 (2.3.2)$$

$$(2 \quad 0 \quad 2 \quad -1) \mathbf{x} = 0$$
 (2.3.3)

and  $W_2 = \{\mathbf{x} \in \mathbb{R}^4 : \}$ 

$$(1 1 0 1) \mathbf{x} = 0 (2.3.4)$$

$$(1 0 1 -2) \mathbf{x} = 0 (2.3.5)$$

$$(0 \quad 1 \quad 0 \quad -1)\mathbf{x} = 0. \tag{2.3.6}$$

Then which among the following is true?

- a)  $\dim(W_1) = 1$
- b)  $\dim(W_2) = 2$
- c) dim  $(W_1 \cap W_2) = 1$
- d)  $\dim(W_1 + W_2) = 3$
- 2.4. Let A be an  $n \times n$  complex matrix. Assume that A is self-adjoint and let B denote the inverse of A + II. Then all eigenvalues of (A - II)B are
  - a) purely imaginary
  - b) of modulus one
  - c) real
  - d) of modulus less than one
- 2.5. Let  $\{u_1, u_2, \dots, u_n\}$  be an orthonormal basis of  $\mathbb{C}^n$  as column vectors.Let

$$\mathbf{M} = \begin{pmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \dots & \mathbf{u}_k \end{pmatrix}, \tag{2.5.1}$$

$$\mathbf{N} = \begin{pmatrix} \mathbf{u}_{k+1} & \mathbf{u}_{k+2} & \dots & \mathbf{u}_n \end{pmatrix}$$
 (2.5.2)

and **P** be the diagonal  $k \times k$  matrix with diagonal entries  $\alpha_1, \alpha_2, \dots, \alpha_k \in \mathbb{R}$ . Then which of the following is true?

- a) rank(**MPM**\*) = k whenever  $\alpha_i \neq \alpha_i, 1 \leq$  $i, j \leq k$ .
- b)  $\operatorname{tr}(\mathbf{MPM}^*) = \sum_{i=1}^k \alpha_i$
- c)  $rank(\mathbf{M}^*\mathbf{N}) = min(k, n k)$
- d)  $\operatorname{rank}(\mathbf{MM}^* + \mathbf{NN}^*) < n$ .

2.6. Let  $B: \mathbb{R} \times \mathbb{R} \to \mathbb{R}$  be the function

$$B(a,b) = ab \tag{2.6.1}$$

Which of the following is true?

- a) B is a linear transformation
- b) B is a positive definite bilinear form
- c) B is symmetric but not positive definite
- d) B is neither linear nor bilinear
- 2.7. Let **A** be an invertible real  $n \times n$  matrix. Define a function

$$F: \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R} \tag{2.7.1}$$

by

$$F(\mathbf{x}, \mathbf{v}) = (F\mathbf{x})^T \mathbf{v} \tag{2.7.2}$$

Let  $DF(\mathbf{x}, \mathbf{y})$  denote the derivate of F at  $(\mathbf{x}, \mathbf{y})$ which is a linear transformation from

$$\mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R} \tag{2.7.3}$$

Then, if

- a)  $\mathbf{x} \neq 0, DF(\mathbf{x}, \mathbf{0}) \neq 0$
- b)  $y \neq 0, DF(0, y) \neq 0$
- c)  $(x, y) \neq (0, 0), DF(x, 0) \neq 0$
- d) x = 0 or y = 0, DF(x, y) = 0
- 2.8. Let

$$T: \mathbb{R}^n \to \mathbb{R}^n \tag{2.8.1}$$

be a linear map that satisfies

$$T^2 = T - I. (2.8.2)$$

Then which of the following is true?

- a) T is invertible.
- b) T I is not invertible.
- c) T has a real eigenvalue.
- d)  $T^3 = -I$ .
- 2.9. Let

$$\mathbf{M} = \begin{pmatrix} 2 & 0 & 3 & 2 & 0 & -2 \\ 0 & 1 & 0 & -1 & 3 & 4 \\ 0 & 0 & 1 & 0 & 4 & 4 \\ 1 & 1 & 1 & 0 & 1 & 1 \end{pmatrix}$$
 (2.9.1)

$$\mathbf{b}_1 = \begin{pmatrix} 5\\1\\1\\4 \end{pmatrix}, \mathbf{b}_2 = \begin{pmatrix} 5\\1\\3\\3 \end{pmatrix}. \tag{2.9.2}$$

Then which of the following are true?

a) both systems  $Mx = b_1$  and  $Mx = b_2$  are

inconsistent.

- b) both systems  $Mx = b_1$  and  $Mx = b_2$  are consistent.
- c) the system  $\mathbf{M}\mathbf{x} = \mathbf{b}_1 \mathbf{b}_2$  is consistent.
- d) the system  $\mathbf{M}\mathbf{x} = \mathbf{b}_1 \mathbf{b}_2$  is inconsistent.

## 2.10. Let

$$\mathbf{M} = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 1 & 4 \\ -2 & 1 & -4 \end{pmatrix}. \tag{2.10.1}$$

Given that 1 is an eigenvalue of M, then which among the following are correct?

- minimal polynomial a) The (x-1)(x+4)
- polynomial b) The minimal of M is  $(x-1)^2(x+4)$
- c) M is not diagonalizable.
- d)  $\mathbf{M}^{-1} = \frac{1}{4} (\mathbf{M} + 3\mathbf{I}).$
- 2.11. Let A be a real matrix with characteristic polynomial  $(x-1)^3$ . Pick the correct statements from below:
  - a) A is necessarily diagonalizable.
  - b) If the minimal polynomial of **A** is  $(x-1)^3$ , then A is diagonalizable.
  - c) The characteristic polynomial of  $A^2$  is  $(x-1)^3$
  - d) If A has exactly two Jordan blocks, then  $(\mathbf{A} - \mathbf{I})^2$  is diagonalizable.
- 2.12. Let  $P_3$  be the vector space of polynomials 2.16. Consider a Markov Chain with state space with real coefficients and of degree at most 3. Consider the linear map

$$T: P_3 \to P_3 \tag{2.12.1}$$

defined by

$$T(p(x)) = p(x-1) + p(x+1)$$
 (2.12.2)

Which of the following properties does the matrix of T with respect to the standard basis  $B = \{1, x, x^2, x^3\}$  of  $P_3$  satisfy?

- a) detT = 0.
- b)  $(T 2I)^4 = 0$  but  $(T 2I)^3 \neq 0$ .
- c)  $(T 2I)^3 = 0$  but  $(T 2I)^2 \neq 0$ .
- d) 2 is an eigenvalue with multiplicity 4.
- 2.13. Let **M** be an  $n \times n$  Hermitian matrix of rank  $k, k \neq n$ . If  $\lambda \neq = 0$  is an eigenvalue of M with corresponding unit column vector **u**, then which of the following are true?
  - a)  $\operatorname{rank}(\mathbf{M} \lambda \mathbf{u}\mathbf{u}^*) = k 1$ .

- b)  $\operatorname{rank}(\mathbf{M} \lambda \mathbf{u}\mathbf{u}^*) = k$ .
- c) rank( $\mathbf{M} \lambda \mathbf{u}\mathbf{u}^*$ ) = k + 1.
- d)  $(\mathbf{M} \lambda \mathbf{u}\mathbf{u}^*)^n = \mathbf{M}^n \lambda^n \mathbf{u}\mathbf{u}^*$ .
- 2.14. Define a real valued function B on  $\mathbb{R}^2 \times \mathbb{R}^2$  as

$$B(\mathbf{x}, \mathbf{y}) = x_1 y_1 - x_1 y_2 - x_2 y_1 + 4 x_2 y_2$$
 (2.14.1)

Let 
$$\mathbf{v}_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$
 and 
$$W = \left\{ \mathbf{v} \in \mathbb{R}^2 : B(\mathbf{v}_0, \mathbf{v}) = 0 \right\}$$
 (2.14.2)

Then W

- a) is not a subspace of  $\mathbb{R}^2$ .
- b) equals **0**.
- c) is the y axis
- d) is the line passing through  $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .
- 2.15. Consider the Quadratic forms

$$Q_1(x, y) = xy (2.15.1)$$

$$Q_2(x, y) = x^2 + 2xy + y^2$$
 (2.15.2)

$$Q_3(x, y) = x^2 + 3xy + 2y^2 (2.15.3)$$

on  $\mathbb{R}^2$ . Choose the correct statements from below

- a)  $Q_1$  and  $Q_2$  are equivalent.
- b)  $Q_1$  and  $Q_3$  are equivalent.
- c)  $Q_2$  and  $Q_3$  are equivalent.
- d) all are equivalent.
- $\{0, 1, 2\}$  and transition matrix

$$P = \begin{array}{ccc} 0 & 1 & 2 \\ 0 \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & \frac{3}{4} \\ 2 \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{pmatrix} \end{array}$$
 (2.16.1)

For any two states i and j, let  $p_{ij}^{(n)}$  denote the n-step transition probability of going from i to j. Identify correct statements.

- a)  $\lim_{n\to\infty} p_{11}^{(n)} = \frac{2}{9}$ b)  $\lim_{n\to\infty} p_{21}^{(n)} = 0$ c)  $\lim_{n\to\infty} p_{32}^{(n)} = \frac{1}{3}$ d)  $\lim_{n\to\infty} p_{13}^{(n)} = \frac{1}{3}$

# 3 June 2018

- 3.1. Let **A** be a  $(m \times n)$  matrix and **B** be a  $(n \times m)$ matrix over real numbers with m < n. Then
  - a) **AB** is always nonsingular.

- b) **AB** is always singular.
- c) **BA** is always nonsingular.
- d) **BA** is always singular.
- 3.2. If **A** is a  $(2 \times 2)$  matrix over  $\mathbb{R}$  with  $det(\mathbf{A} + \mathbf{I}) = 1 + det(\mathbf{A})$ . Then we can conclude that
  - a) det(**A**) = 0.
  - b) **A**= 0.
  - c)  $tr(\mathbf{A}) = 0$ .
  - d) A is nonsingular.
- 3.3. The system of equations

$$x + 2x^2 + 3xy = 6 (3.3.1)$$

$$x + x^2 + 3xy + y = 5 (3.3.2)$$

$$x - x^2 + y = 7 ag{3.3.3}$$

- a) has solutions in rational numbers.
- b) has solutions in real numbers.
- c) has solutions in complex numbers.
- d) has no solutions.
- 3.4. The trace of the matrix

$$\begin{pmatrix} 2 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix}^{20} \tag{3.4.1}$$

is

- a)  $7^{20}$ .
- b)  $2^{20} + 3^{20}$ .
- c)  $2^{21} + 3^{20}$ .
- d)  $2^{20} + 3^{20} + 1$ .
- 3.5. Given that there are real constants a, b, c, dsuch that the identity

$$\lambda x^2 + 2xy + y^2 = (ax + by)^2 + (cx + dy)^2$$
,

 $\forall x, y \in \mathbb{R} \quad (3.5.1)$ 

This implies that

- a)  $\lambda = -5$
- b)  $\lambda \geq 1$
- c)  $0 < \lambda < 1$
- d) There is no such  $\lambda \in \mathbb{R}$
- 3.6. Let  $\mathbb{R}, n \geq 2$ , be equipped with the standard inner product. Let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  be n column vectors forming an orthonormal basis of  $\mathbb{R}^n$ . Let A be the  $n \times n$  matrix formed by the column vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ . Then

a) 
$$A = A^{-1}$$

- c)  $\mathbf{A}^{-1} = \mathbf{A}^{\mathsf{T}}$
- b)  $\mathbf{A} = \mathbf{A}^{\mathsf{T}}$
- d)  $det(\mathbf{A}) = 1$
- 3.7. Consider a Markov Chain with state space  $\{1, 2, 3, 4\}$  and transition matrix

$$P = \begin{array}{cccc} 1 & 2 & 3 & 4 \\ 1 \begin{pmatrix} \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{3} & 0 & \frac{1}{3} & \frac{1}{3} \\ 4 \begin{pmatrix} \frac{1}{2} & 0 & \frac{1}{2} & 0 \end{pmatrix} \end{array}$$
(3.7.1)

- a)  $\lim_{n\to\infty} p_{22}^{(n)} = 0$ ,  $\sum_{n=0}^{\infty} p_{22}^{(n)} = \infty$ b)  $\lim_{n\to\infty} p_{22}^{(n)} = 0$ ,  $\sum_{n=0}^{\infty} p_{22}^{(n)} < \infty$ c)  $\lim_{n\to\infty} p_{22}^{(n)} = 1$ ,  $\sum_{n=0}^{\infty} p_{22}^{(n)} = \infty$ d)  $\lim_{n\to\infty} p_{22}^{(n)} = 1$ ,  $\sum_{n=0}^{\infty} p_{22}^{(n)} < \infty$

- 3.8. Let V denote the vector space of all sequences  $\mathbf{a} = (a_1, a_2, \dots)$  of real numbers such that

$$\sum_{n} 2^n |a|_n \tag{3.8.1}$$

converges. Define

$$\|\cdot\|: V \to \mathbb{R} \tag{3.8.2}$$

by

$$\|\mathbf{a}\| = \sum_{n} 2^{n} |a|_{n}.$$
 (3.8.3)

Which of the following are true?

- a) V contains only the sequence  $(0,0,\ldots)$
- b) V is finite dimensional
- c) V has a countable linear basis
- d) V is a complete normed space
- 3.9. Let *V* be a vector space over  $\mathbb{C}$  with dimension n. Let  $T: V \to V$  be a linear transformation with only1 as eigenvalue. Then which of the following must be true?
  - a) T I = 0
  - b)  $(T-I)^{n-1}=0$
  - c)  $(T-I)^n=0$
  - d)  $(T-I)^{2n} = 0$
- 3.10. If **A** is a  $5 \times 5$  matrix and the dimension of the solution space of Ax = 0 is at least two, then
  - a) rank $(\mathbf{A}^2) \leq 3$
  - b) rank  $(\mathbf{A}^2) \ge 3$
  - c) rank $(\mathbf{A}^2) = 3$
  - d)  $\det(\mathbf{A}^2) = 0$

- - a) minimal polynomial of A can only be of degree 2
  - b) minimal polynomial of A can only be of degree 3
  - c) either A = I or A = -I
  - d) there can be uncountably many A satisfying the above.
- 3.12. Let **A** be an  $n \times n$ , n > 1 matrix satisfying

$$\mathbf{A}^2 - 7\mathbf{A} + 12\mathbf{I} = \mathbf{0}$$
 (3.12.1) 3.15. Let

Then which of the following statements is true?

- a) A is invertible
- b)  $t^2 7t + 12n = 0$  where t = tr(A)
- c)  $d^2 7d + 12 = 0$  where  $d = det(\mathbf{A})$
- d)  $\lambda^2 7\lambda + 12 = 0$  where  $\lambda$  is an eigenvalue of
- 3.13. Let **A** be a  $6 \times 6$  matrix over  $\mathbb{R}$  with characteristic polynomial

$$(x-3)^2 (x-2)^4$$
 (3.13.1)

and minimal polynomial

$$(x-3)(x-2)^2$$
 (3.13.2)

Then the Jordan canonical form of A can be

a) 
$$\begin{pmatrix} 3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 & 0 \\ 0 & 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 2 \end{pmatrix}$$
b) 
$$\begin{pmatrix} 3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 2 \end{pmatrix}$$
c) 
$$\begin{pmatrix} 3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 2 \end{pmatrix}$$
d) 
$$\begin{pmatrix} 3 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 2 \end{pmatrix}$$

- 3.11. Let  $A \in M_3(\mathbb{R})$  be such that  $A^3 = I_{3\times 3}$ . Then 3.14. Let V be an inner product space and S be a subset of V. Let  $\bar{S}$  denote the closure of S in V with respect to the topology induced by the metric given by the inner product. Which of the following statements is true?
  - a)  $S = (S^{\perp})^{\perp}$
  - b)  $\bar{S} = (S^{\perp})^{\perp}$
  - c)  $\overline{span(S)} = (S^{\perp})^{\perp}$
  - d)  $S^{\perp} = \left( \left( S^{\perp} \right)^{\perp} \right)^{\perp}$

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & -2 \\ 0 & 0 & 1 \end{pmatrix} \tag{3.15.1}$$

and

$$Q(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x} \tag{3.15.2}$$

Which of the following statements is true?

- a) The matrix of second order partial derivatives of the quadratic form Q is 2A
- b) The rank of the quadratic form Q is 2
- c) The signature of the quadratic form Q is ++0
- d) The quadratic form Q take the value 0 for some non-zero vector x
- 3.16. Assume that a non-singular matrix

$$\mathbf{A} = \mathbf{L} + \mathbf{D} + \mathbf{U} \tag{3.16.1}$$

where L and U are lower and upper triangular matrices respectively with all diagonal entries are zero, and **D** si a diagonal matrix. Let  $\mathbf{x}^*$  be the solution of Ax = b. Then the Gauss-Seidel iteration method

$$\mathbf{x}_{k+1} = \mathbf{H}\mathbf{x}_k + \mathbf{c}, k = 0, 1, 2, \dots$$
 (3.16.2)

with  $\|\mathbf{H}\| < 1$  converges to  $\mathbf{x}^*$  provided  $\mathbf{H}$  is equal to

- a)  $-\mathbf{D}^{-1}(\mathbf{L} + \mathbf{U})$
- b)  $-(\mathbf{D} + \mathbf{L})^{-1} \mathbf{U}$
- c)  $-\mathbf{D}(\mathbf{L} + \mathbf{U})^{-1}$
- $(\mathbf{L} \mathbf{D})^{-1} \mathbf{U}$
- 3.17. Consider a Markov Chain with state space S =

 $\{1, 2, 3\}$  and transition matrix

$$P = \begin{array}{ccc} 1 & 2 & 3 \\ 1 \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix}$$
 (3.17.1)

Let  $\pi$  be a stationary distribution of the Markov chain and d(1) denote the period of state 1. Which of the following statements are correct?

- a) d(1) = 1
- b) d(1) = 2
- c)  $\pi_1 = \frac{1}{2}$ d)  $\pi_1 = \frac{1}{3}$

# 4 December 2017

- 4.1. Let A be a real symmetric matrix and B = I + iA, where  $i^2 = -1$ . Then choose the correct option.
  - a) **B** is invertible if and only if **A** is invertible.
  - b) All Eigenvalues of **B** are necessarily real.
  - c)  $\mathbf{B} \mathbf{I}$  is necessarily invertible.
  - d) **B** is necessarily invertible.

**Solution:** See Table 4.1.1.

Statement 1.	<b>B</b> is invertible if and only if <b>A</b> is invertible.	
False statement	Matrix <b>B</b> is invertible even if <b>A</b> is non invertible.	
Example:	Consider a matrix	
	$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$	(4.1.1)
	a real non invertible, symmetric matrix.	
	$\implies \mathbf{B} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + i \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 1+i & 0 \\ 0 & 1 \end{pmatrix}$	(4.1.2)
	is invertible even if <b>A</b> is non invertible.	
Statement 2.	All Eigenvalues of <b>B</b> are necessarily real.	
False statement	Matrix <b>B</b> can have complex Eigenvalues.	
Proof:	Eigen values of $\mathbf{B}$ = Eigen values of $(\mathbf{I})$ + i (Eigen values of $\mathbf{A}$ ). Clearly from (4.1.2) above Eigen values of $\mathbf{B}$ are 1 and 1 + i respectively. Hence $\mathbf{B}$ can also have complex Eigen value.	
Statement 3.	$\mathbf{B} - \mathbf{I}$ is necessarily invertible.	
False statement	$\mathbf{B} - \mathbf{I} = i\mathbf{A}$ will be invertible if $\mathbf{A}$ , is invertible.	
Proof:	We have $\mathbf{B} - \mathbf{I} = i\mathbf{A}$	
	$\implies \mathbf{B} - \mathbf{I} = i\mathbf{A} = \begin{pmatrix} i & 0 \\ 0 & 0 \end{pmatrix}, \text{from (4.1.1)}$	
	Hence $\mathbf{B} - \mathbf{I}$ is not invertible, unless $\mathbf{A}$ is invertible.	
Statement 4.	<b>B</b> is necessarily invertible.	
Correct Statement:	Matrix <b>B</b> has non zero Eigen values corresponding to Eigenv	ector X.
Proof:	Let $X$ be an Eigen vector of $\mathbf{A}$ corresponding to Eigen value $\lambda$	
	also, $\lambda\epsilon\mathbb{R}$	
	$\implies \mathbf{A}X = \lambda X$	
	$\therefore \mathbf{B}X = (\mathbf{I} + i\mathbf{A})X = \mathbf{I}X + i\mathbf{A}X = X + i\lambda X$	
	$\Longrightarrow \mathbf{B}X = (1+i\lambda)X$	
	Therefore, $1 + i\lambda$ is an Eigen value of <b>B</b> ,	
	corresponding to Eigen vector <i>X</i> , which are non zero. Hence, <b>B</b> is necessarily invertible.	

TABLE 4.1.1: Solution summary

4.2. Let  $\mathbf{A} = \begin{pmatrix} 0 & 1 \\ -1 & 1 \end{pmatrix}$ . Then the smallest positive integer n such that  $\mathbf{A}^n = \mathbf{I}$  is

**Solution:** *Property of eigen values of A:* Let **A** be an arbitary  $n \times n$  matrix of complex numbers with eigen values  $\lambda_1, \lambda_2, \ldots, \lambda_n$ . Then the eigen values of  $\mathbf{k}^{\text{th}}$  power of **A**, that is the eigen values of  $\mathbf{A}^k$ , for any positive integer **k** are  $\lambda_1^k, \lambda_2^k, \ldots, \lambda_n^k$ . Let us calculate the eigen values of **A**.

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ -1 & 1 \end{pmatrix} \tag{4.2.1}$$

$$\det(\mathbf{A} - \lambda \mathbf{I}) = 0 \tag{4.2.2}$$

$$\begin{vmatrix} -\lambda & 1 \\ -1 & 1 - \lambda \end{vmatrix} = 0 \tag{4.2.3}$$

$$-\lambda(1 - \lambda) + 1 = 0 \tag{4.2.4}$$

$$\lambda^2 - \lambda + 1 = 0 \tag{4.2.5}$$

$$\implies \lambda = \frac{-1 \pm \sqrt{3}i}{2} \tag{4.2.6}$$

From the above property, the eigen values of  $A^n$  are  $\lambda^n$ . Also as it is given that  $A^n = I$ ,

$$\implies \lambda^n = 1$$
 (4.2.7)

$$\Longrightarrow \left(\frac{-1 \pm \sqrt{3}i}{2}\right)^n = 1 \tag{4.2.8}$$

Clearly  $n \neq 1$ . For n = 2,

$$\left(\frac{-1 \pm \sqrt{3}i}{2}\right)^2 = \frac{-1 \mp \sqrt{3}i}{2} \tag{4.2.9}$$

For n = 4,

$$\left(\frac{-1 \pm \sqrt{3}i}{2}\right)^4 = \frac{-1 \pm \sqrt{3}i}{2} \tag{4.2.10}$$

For n = 6,

$$\left(\frac{-1 \pm \sqrt{3}i}{2}\right)^6 = 1\tag{4.2.11}$$

Hence n = 6 is the smallest positive integer.

4.3. Let 
$$\mathbf{A} = \begin{pmatrix} 1 & -1 & 1 \\ 1 & 1 & 1 \\ 2 & 3 & \alpha \end{pmatrix}$$
 and  $\mathbf{b} = \begin{pmatrix} 1 \\ 3 \\ \beta \end{pmatrix}$ . Then the system  $\mathbf{A}\mathbf{X} = \mathbf{b}$  over the real numbers has

a) No solution when  $\beta \neq 7$ 

- b) Infinite number of solutions when  $\alpha \neq 2$
- c) Infinite number of solutions when  $\alpha = 2$  and  $\beta \neq$

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d) A unique solution if  $\alpha \neq 2$ 

**Solution:** First we derive the Row Reduced Echelon Form (RREF) of the augmented matrix of the system AX = b as follows,

$$\begin{pmatrix} 1 & -1 & 1 & 1 \\ 1 & 1 & 1 & 3 \\ 2 & 3 & \alpha & \beta \end{pmatrix} \xrightarrow{R_2 = R_2 - R_1} \begin{pmatrix} 1 & -1 & 1 & 1 \\ 0 & 2 & 0 & 2 \\ 0 & 5 & \alpha - 2 & \beta - 2 \end{pmatrix}$$

$$(4.3.1)$$

$$\stackrel{R_2 = \frac{1}{2}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & -1 & 1 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 5 & \alpha - 2 & \beta - 2 \end{pmatrix} \tag{4.3.2}$$

$$\stackrel{R_1 = R_1 + R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 1 & 2 \\
0 & 1 & 0 & 1 \\
0 & 5 & \alpha - 2 & \beta - 2
\end{pmatrix}$$
(4.3.3)

$$\stackrel{R_3=R_3-5R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 1 & 2 \\
0 & 1 & 0 & 1 \\
0 & 0 & \alpha-2 & \beta-7
\end{pmatrix}$$
(4.3.4)

From the RREF of the augmented matrix of the system  $\mathbf{AX} = \mathbf{b}$  in (4.3.4) we make the following observations for different values of  $\alpha$  and  $\beta$  in Table 4.3.1.

Values	Observations	
	Then the existence of solution and	
$\beta \neq 7$	the number of solutions will entirely	
	depend on value of $\alpha$	
	Then RREF in (4.3.4) will contain	
$\alpha = 2$	Zero Row in $R_3$ . Moreover solvability	
$\beta \neq 7$	condition will not satisfy.	
	⇒ system will have Zero solutions	
	RREF in (4.3.4) will have all pivots	
$\alpha \neq 2$	$\implies$ RREF in (4.3.4) will be fullrank	
	$\implies$ <b>AX</b> = <b>b</b> have unique solution.	

**TABLE 4.3.1** 

Hence, if  $\alpha \neq 2$  then the system  $\mathbf{AX} = \mathbf{b}$  has unique solution.

4.4. Consider a Markov chain  $\{X_n | n \ge 0\}$  with state space  $\{1, 2, 3\}$  and transition matrix

$$\mathbf{P} = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix}$$

Then,  $P(X_3 = 1 | X_0 = 1)$  equals

**Solution:** The three step transitional probabilities are given as,

$$P(X_3 = j | X_0 = i) = P(X_{n+3} = j | X_n = i) =$$

$$(\mathbf{P}^3)_{ij} \text{ for any } n$$
(4.4.1)

$$\mathbf{P}^{3} = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix}^{3} = \begin{pmatrix} \frac{1}{4} & \frac{3}{8} & \frac{3}{8} \\ \frac{3}{8} & \frac{1}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{3}{8} & \frac{1}{4} \end{pmatrix}$$
(4.4.2)

From (4.4.2),

$$P(X_3 = 1 \mid X_0 = 1) = (\mathbf{P}^3)_{11} = \frac{1}{4}$$
 (4.4.3)

- 4.5. Let **A** be an  $m \times n$  matrix with rank r. If the linear system AX = b has a solution for each  $\mathbf{b} \in \mathbf{R}^m$ , then
  - a) m = r
  - b) the column space of A is a proper subspace of
  - c) the null space of A is a non-trivial subspace of  $\mathbf{R}^n$  whenever m = n
  - d)  $m \ge n$  implies m = n

**Solution:** *Theorem* 

**Theorem 4.1.** Consider the  $m \times n$  system Ax =b, with either  $b \neq 0$  or b = 0. We distinguish the following cases:

- a) Unique Solution: If  $rank[A,b] = rank(A) = n \le$ m, then and only then the system has a unique solution. In this case, indeed as many as m - nequations are redundant. And the solution X = $A^{-1}b$ . This is called as **Exactly Determined**.
- b) No Solution: If rank[A,b] > rank(A) which necessarily implies  $\mathbf{b} \neq 0$  and m > rank(A), then and only then the system has no solution. This is called as **Overdetermined**.

See Table 4.5.1 If the columns of an  $m \times n$ matrix A span  $\mathbf{R}^m$  then the equation  $\mathbf{A}\mathbf{x} = \mathbf{b}$ is consistent for each **b** in  $\mathbb{R}^m$ .

The **null space** of **A** is defined to be

$$Null(\mathbf{A}) = \{ \mathbf{x} \in \mathbf{R}^n \,|\, \mathbf{A}\mathbf{x} = 0 \} \tag{4.5.1}$$

$$\mathbf{A} = \begin{pmatrix} -3 & -2 & 4\\ 14 & 8 & -18\\ 4 & 2 & -4 \end{pmatrix} \tag{4.5.2}$$

Reduced Row Echelon form is

$$RREF(\mathbf{A}) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{4.5.3}$$

: the only possible nullspace of the matrix A

Let **B** be given as

$$\mathbf{B} = \begin{pmatrix} -3 & -2 & 4\\ 14 & 8 & -18\\ 4 & 2 & -4\\ 28 & 16 & -36\\ 8 & 4 & -8 \end{pmatrix} \tag{4.5.4}$$

Reduced Row Echelon form is

$$RREF(\mathbf{B}) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \tag{4.5.5}$$

 $\therefore$  the rank of matrix **B** = 3.

4.6. Let  $\mathbf{M} = \left\{ \begin{pmatrix} a & b \\ c & d \end{pmatrix} : a, b, c, d \in \mathbb{Z} \text{ and eigen values of } \mathbf{A} \in \mathbb{Q} \right\}$ 

a) M is empty

b) 
$$\mathbf{M} = \left\{ \begin{pmatrix} a & b \\ c & d \end{pmatrix} : a, b, c, d \in \mathbb{Z} \right\}$$
  
c) If  $\mathbf{A} \in \mathbf{M}$  then the eigen values of  $\mathbf{A} \in \mathbb{Z}$ 

- d) If  $A,B \in M$  such that AB=I then  $|A| \in \{+1,-1\}$ **Solution:** See Table 4.6.1.

Options	Observations
m = r	The rank of any matrix $A$ is the dimension of its column space. When the number of rows $(m)$ is equal to the rank $(r)$ of the matrix, then their linear combination gives us span of $\mathbf{R}^m$ . $\therefore$ This statement is <b>True</b> .
the column space of <b>A</b> is a proper subspace of <b>R</b> <sup>m</sup>	Any subspace of a vector space $V$ other than $V$ itself is considered a proper subspace of $V$ . Which means that linear combination of $A$ will span less than $m$ . That will make the resultant $b$ span strictly less than $m$ . But it is given that $b \in \mathbb{R}^m$ , which is contradicting. $\therefore$ This statement is <b>False</b> .
the null space of <b>A</b> is a non-trivial subsapce of $\mathbf{R}^n$ whenever $m = n$	From (4.5.2) we see that even when $m = n$ then also we are getting a trivial nullspace. $\therefore$ This statement is <b>False</b> .
$m \ge n$ implies $m = n$	It is given that the number of rows are greater than the column, and it is given that there exists a solution. If we refer to theorem (4.1) we see that the corresponding system will be <b>Exactly Determined</b> system.  As an example, it will look like (4.5.4).  ∴ This statement is <b>True</b> .

TABLE 4.5.1: Solution

M is empty	Consider $\mathbf{A} = \mathbf{I} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ . The elements of $\mathbf{A} \in \mathbb{Z}$ and it's eigen values $1 \in \mathbb{Q}$ . So, $\mathbf{M}$ is not empty.
$\mathbf{M} = \left\{ \begin{pmatrix} a & b \\ c & d \end{pmatrix} : a, b, c, d \in \mathbb{Z} \right\}$	Let $\mathbf{A} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$ where elements of $\mathbf{A} \in \mathbb{Z}$ . The characteristic equation can be written as:
	$\lambda^2 + 1 = 0 \implies \lambda = \pm i$

	We see that $\lambda \in \mathbb{C}$ which is contradicting the main definition of $M$ .So,this is not correct.
Eigen values of $\mathbf{A} \in \mathbb{Z}$	Given $A \in M$ .Let $\lambda_1, \lambda_2$ be the eigen values of $A$ .The characteristic polynomial can be written as:
	$\lambda^2 - tr(\mathbf{A}) \lambda + \det \mathbf{A} = 0 \text{ where } tr(\mathbf{A}) = \lambda_1 + \lambda_2, \det \mathbf{A} = \lambda_1 \lambda_2$
	Given the eigen values $\lambda_1, \lambda_2 \in \mathbb{Q}$ , For this to be possible the discriminant of above equation should $\in \mathbb{Z}$ $\frac{\sqrt{(\lambda_1 + \lambda_2)^2 - 4\lambda_1\lambda_2}}{\sqrt{(\lambda_1 - \lambda_2)^2} \in \mathbb{Z}}$ $\implies \lambda_1 - \lambda_2 \in \mathbb{Z} \text{ This is possible when both } \lambda_1, \lambda_2 \in \mathbb{Z}.$
If $AB=I$ then $ A  \in \{+1,-1\}$	As $\mathbf{A}, \mathbf{B} \in \mathbf{M}$ , $\Longrightarrow  \mathbf{A} ,  \mathbf{B}  \in \mathbb{Z}$ Given $\mathbf{A}\mathbf{B} = \mathbf{I} \implies  \mathbf{A}   \mathbf{B}  = 1$ This is possible only when $ \mathbf{A}  =  \mathbf{B}  = \pm 1$
Conclusion	options 3) and 4) are correct.

TABLE 4.6.1: Solution

4.7. Let A be a 3×3 matrix with real entries. Identify the correct statements.

- a) A is necessarily diagonalizable over  ${\bf R}$
- b) If A has distinct real eigen values than it is diagonalizable over R
- c) If A has distinct eigen values than it is diagonalizable over C
- d) If all eigen values are non zero than it is diagonalizable over  ${\bf C}$

**Solution:** See Table 4.7.1.

Statement 1.	A is necessarily diagonalizable over <b>R</b>		
False statement  Example:	Matrix A is diagonalizable if and only if there is a basis of $\mathbb{R}^3$ consisting of eigenvectors of A. Consider a matrix		
	$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 4 \end{pmatrix} \tag{4.7.1}$		
	Eigen values are:		
	$\begin{pmatrix} 1 - \lambda & 1 & 0 \\ 0 & 1 - \lambda & 1 \\ 0 & 0 & 4 - \lambda \end{pmatrix} = 0. \implies \lambda_1 = 1, \lambda_2 = 4 $ (4.7.2)		
	$\lambda_1 = 1$ has eigen vector $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ and $\lambda_2 = 4$ has eigen vector $\begin{pmatrix} 1 \\ 3 \\ 9 \end{pmatrix}$ (4.7.3)		
	We have found only two linearly independent eigenvectors for A,not diagonalisable		
Statement 2.	If A has distinct real eigen values than it is diagonalizable over <b>R</b>		
True statement	Distinct real eigenvalues implies linearly independent eigenvectors . and if a matrix has n linearly independent vectors than it is diagonalizable.		
Proof 1:	Distinct eigen values implies linearly independent vectors that spans entire space. Consider 2 eigen vectors $\mathbf{v}$ , $\mathbf{w}$ with eigen values $\lambda$ , $\mu$ respectively. such that $\lambda \neq \mu$		
	$\alpha(\mathbf{v}) + \beta(\mathbf{w}) = 0 \tag{4.7.4}$		
	$\alpha A(\mathbf{v}) + \beta A(\mathbf{w}) = 0 \tag{4.7.5}$		
	$\alpha \lambda \mathbf{v} + \beta \mu \mathbf{w} = 0 \tag{4.7.6}$		
	Multiplying (4.7.4)with $-\lambda$ and subtracting from (4.7.6) we have,		
	$\beta(\mu - \lambda)\mathbf{w} = 0 \tag{4.7.7}$		
Proof 2:	eigen values are distinct $(\mu - \lambda) \neq 0$ . From equation (4.7.7) we have, $\beta = 0$ substituting $\beta = 0$ in equation (4.7.4) we have, $\alpha = 0$ . As, $\mathbf{v} \neq 0$ which proves that vectors are linearly independent.  If a matrix has n linearly independent vectors than it is diagonalizable If $(\mathbf{p_1}  \mathbf{p_2}  \cdots  \mathbf{p_n})$ are n independent eigen vectors then, $A\mathbf{p_1} = \lambda \mathbf{p_1}, \cdots, A\mathbf{p_n} = \lambda \mathbf{p_n}$		
	$D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix} P = (\mathbf{P_1} \ \mathbf{P_2} \ \cdots \ \mathbf{P_n}) $ $Now, A\mathbf{P_i} = \lambda_i \mathbf{P_i} \implies AP = PD$ $(4.7.8)$		

	$so, P^{-1}AP = D$ is a diagonal matrix.	
Statement 3.	If A has distinct real eigen values than it is diagonalizable overC	
True statement	If A is an $N \times N$ complex matrix with n distinct eigenvalues, then any set of n corresponding eigenvectors form a basis for $\mathbb{C}^n$	
Proof:	It is sufficient to prove that the set of eigenvectors is linearly independent which is proved in statement 2.	
Example:	$A = \begin{pmatrix} 4 & 0 & -2 \\ 2 & 5 & 4 \\ 0 & 0 & 5 \end{pmatrix} \tag{4.7.9}$	
	Eigen values of A are:	
	$\lambda_1 = 2, \lambda_2 = 3, \lambda_3 = 6 \tag{4.7.10}$	
	Eigen vectors are:	
	$x_1 = \begin{pmatrix} -1\\1\\0 \end{pmatrix}, x_2 = \begin{pmatrix} 1\\1\\1 \end{pmatrix}, x_3 = \begin{pmatrix} -1\\-1\\2 \end{pmatrix} $ (4.7.11)	
	Matrix A is diagonalizable because there is a basis of $\mathbb{C}^3$ consisting of eigenvectors of A.	
Statement 4.	If all eigen values are non zero than it is diagonalizable over C	
False Statement:	Matrix would be diagonalizable if and only if it has linearly independent eigenvectors.	
Example:	Consider a matrix	
	$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 4 \end{pmatrix} \tag{4.7.12}$	
	Eigen values are:	
	$\begin{pmatrix} 1 - \lambda & 1 & 0 \\ 0 & 1 - \lambda & 1 \\ 0 & 0 & 4 - \lambda \end{pmatrix} = 0. \implies \lambda_1 = 1, \lambda_2 = 4 \neq 0 $ (4.7.13)	
	$\lambda_1 = 1$ has eigen vector $\begin{pmatrix} 1\\0\\0 \end{pmatrix}$ and $\lambda_2 = 4$ has eigen vector $\begin{pmatrix} 1\\3\\9 \end{pmatrix}$ (4.7.14)	
	We have found only two linearly independent eigenvectors for A,not diagonalisable.	

TABLE 4.7.1: Solution summary

Given

V be a vector space over C of all the polynomials in a variable X of degree atmost 3  $D: P_3 \rightarrow P_3$ 

> $D: V \to V$  be the linear operator given by differentiation wrt X  $D(P(x)) \rightarrow P'(x)$

> > A be the matrix of D wrt some basis for V Assume basis for V be  $\{1, x, x^2, x^3\}$

## **TABLE 4.8.1**

- 4.8. Let V be a vector space over C of all the polynomials in a variable X of degree atmost 3. Let  $D: V \to V$  be the linear operator given by differentiation with respect to X. Let A be the matrix of D with respect to some basis for V. Which of the following are true?
  - a) A is nilpotent matrix
  - b) A is diagonalizable matrix
  - c) the rank of A is 2
  - d) the Jordan canonical form of A is

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

**Solution:** See Tables 4.8.1, 4.8.2 and 4.8.3

- 4.9. For every  $4 \times 4$  real symmetric non-singular matrix **A** there exists a positive integer p such 4.10. Let **A** be an  $m \times n$  matrix of rank m with n > m. that
  - a) pI + A is positive definite
  - b)  $A^p$  is positive definite
  - c)  $A^{-p}$  is positive definite
  - d)  $\exp(p\mathbf{A}) \mathbf{I}$  is positive definite

**Solution:** A matrix is real symmetric implies its eigen values are real and eigen vectors are orthogonal, that is its eigen value decomposition is

$$\mathbf{A} = \mathbf{P}\mathbf{D}\mathbf{P}^T \tag{4.9.1}$$

**D** is the diagonal matrix containing the real eigen values of A

P has the corresponding eigen vectors

$$\mathbf{P}\mathbf{P}^T = \mathbf{P}^T\mathbf{P} = \mathbf{I} \tag{4.9.2}$$

A real matrix is positive definite if

$$\mathbf{x}^T \mathbf{A} \mathbf{x} > 0 \tag{4.9.3}$$

$$\implies \mathbf{x}^T \lambda \mathbf{x} > 0 \tag{4.9.4}$$

$$\implies \lambda \mathbf{x}^T \mathbf{x} > 0 \tag{4.9.5}$$

$$\implies \lambda > 0$$
 (4.9.6)

In other words, all the eigen values of A are positive See Table 4.9.1

Let A be

$$\mathbf{A} = \mathbf{P}\mathbf{D}\mathbf{P}^T \tag{4.9.7}$$

$$\mathbf{D} = \begin{pmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 0 & 0 \\ 0 & 0 & \lambda_3 & 0 \\ 0 & 0 & 0 & \lambda_4 \end{pmatrix} \tag{4.9.8}$$

From the table, the choices would be option 1,2,3

- If for some non-zero real number  $\alpha$ , we have  $\mathbf{x}^{T}\mathbf{A}\mathbf{A}^{T}\mathbf{x} = \alpha\mathbf{x}^{T}\mathbf{x}$ , for all  $x \in \mathbf{R}^{m}$ , then  $\mathbf{A}^{T}\mathbf{A}$ 
  - a) exactly two distinct eigenvalues.
  - b) 0 as an eigenvalue with multiplicity n m.
  - c)  $\alpha$  as a non-zero eigenvalue.
  - d) exactly two non-zero distinct eigenvalues.

**Solution:** Refer Table 4.10.1.

Refer Table 4.10.2.

(4.9.1) 4.11. Consider a Markov chain with five states

 $\{1, 2, 3, 4, 5\}$  and transition matrix

$$P = \begin{pmatrix} \frac{1}{2} & 0 & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{7} & 0 & 0 & \frac{6}{7} \\ \frac{1}{5} & \frac{1}{5} & \frac{1}{5} & \frac{1}{5} & \frac{1}{5} \\ \frac{1}{3} & 0 & 0 & \frac{2}{3} & 0 \\ 0 & \frac{5}{8} & 0 & 0 & \frac{3}{8} \end{pmatrix}$$
(4.11.1)

Which of the following are true?

- a) 3 and 1 are in the same communicating class
- b) 1 and 4 are in the same communicating class
- c) 4 and 2 are in the same communicating class
- d) 2 and 5 are in the same communicating class

**Solution:** See Tables 4.11.1 and 4.11.2

$D(1) = 0 = 0.1 + 0.x + 0.x^{2} + 0.x^{3}$		
$D(1) = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$		
$D(x) = 1 = 1.1 + 0.x + 0.x^{2} + 0.x^{3}$		
$D(x) = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}$		
$D(x^2) = 2x = 0.1 + 2.x + 0.x^2 + 0.x^3$		
$D(x^2) = \begin{pmatrix} 0 \\ 2 \\ 0 \\ 0 \end{pmatrix}$		
$D(x^3) = 3x^2 = 0.1 + 0.x + 3.x^2 + 0.x^3$		
$D(x^3) = \begin{pmatrix} 0 \\ 0 \\ 3 \\ 0 \end{pmatrix}$		
$Matrix A = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$		
An $n \times n$ matrix with $\lambda$ as diagonal elements, ones on the super diagonal and zeroes in all other entries is nilpotent with minimal polynomial $(A - \lambda I)^n$		
$A = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$		
All eigen values of matrix <i>A</i> is 0 Thus, above matrix is nilpotent matrix Thus, above statement is true		

TABLE 4.8.2

Diagonalizable	$A = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$ $Rank(A) + nullity(A) = \text{no of column}$ $Rank(A) = 3, \text{ no of column} = 4$ $nullity(A) = 4 - 3 = 1$ $\text{means there exists only one}$ $\text{linearly independent eigen vector}$ $\text{corresponding to 0 eigen values}$ $\text{Thus, matrix } A \text{ is not Diagonalizable.}$ $\text{Thus, above statement is false}$
Rank	$A = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$ Rank of matrix A is 3 Thus, above statement is false
Jordan CF	Assume characteristic polynomial of matrix $A$ is $c_A(x)$ $c_A(x) = x^4$ Assume minimal polynomial of $A$ is $m_A(x)$ $m_A(x)$ always divide $c_A(x)$ $m_A(x) = \{x, x^2, x^3, x^4\}$ Minimal polynomial always annihilates its matrix. Thus, we see that $m_A(A) = \{A = 0, A^2 = 0, A^3 = 0, A^4 = 0\}$ But we see that neither $A$ is zero matrix nor $A^2$ and $A^3$ equal to zero but $A^4$ is equal to zero. Thus, $x^4$ is minimal polynomial.  Algebraic Multiplicity = $a_M(\lambda = 0) = 4$ Geometric Multiplicity = $g_M(\lambda = 0) = 4$ Geometric Multiplicity = $g_M(\lambda = 0) = 4$ Using Inference, $\mathbf{J} = \begin{pmatrix} \lambda & 1 & 0 & 0 \\ 0 & \lambda & 1 & 0 \\ 0 & 0 & \lambda & 1 \\ 0 & 0 & 0 & \lambda \end{pmatrix}$ $\lambda = 0$ $\begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$ which is same as given in the question. Thus, statement is true

OPTIONS	DERIVATIONS	
	$p\mathbf{I} + \mathbf{A} = \mathbf{P}(p\mathbf{I})\mathbf{P}^T + \mathbf{P}\mathbf{D}\mathbf{P}^T$	(4.9.9)
	$= \mathbf{P} \mathbf{D}_1 \mathbf{P}^T$	(4.9.10)
Choice 1	$\mathbf{D}_1 = \begin{pmatrix} \lambda_1 + p & 0 & 0 & 0 \\ 0 & \lambda_2 + p & 0 & 0 \\ 0 & 0 & \lambda_3 + p & 0 \\ 0 & 0 & 0 & \lambda_4 + p \end{pmatrix}$	(4.9.11)
	Some of the eigen values of $A$ may be negative. All the eigen values in $D_1$ are positive only if	
	$p >  \lambda_i  \ \forall i \in [1, 4]$	(4.9.12)
	$\mathbf{A}^2 = \mathbf{A}\mathbf{A}$	(4.9.13)
	$= (\mathbf{P}\mathbf{D}\mathbf{P}^T)(\mathbf{P}\mathbf{D}\mathbf{P}^T)$	(4.9.14)
	$=\mathbf{P}\mathbf{D}^2\mathbf{P}^T$	(4.9.15)
Choice 2	Similarly, $\mathbf{A}^p = \mathbf{P}\mathbf{D}^p\mathbf{P}^T$	(4.9.16)
	$\mathbf{D}^{p} = \begin{pmatrix} \lambda_{1}^{p} & 0 & 0 & 0 \\ 0 & \lambda_{2}^{p} & 0 & 0 \\ 0 & 0 & \lambda_{3}^{p} & 0 \\ 0 & 0 & 0 & \lambda_{4}^{p} \end{pmatrix}$	(4.9.17)
	$\mathbf{A}^p$ is positive definite only if $p$ is even.	
	$\mathbf{A}^{-p} = \mathbf{P}\mathbf{D}^{-p}\mathbf{P}^T$	(4.9.18)
Choice 3	$\mathbf{D}^{-p} = \begin{pmatrix} \lambda_1^{-p} & 0 & 0 & 0\\ 0 & \lambda_2^{-p} & 0 & 0\\ 0 & 0 & \lambda_3^{-p} & 0\\ 0 & 0 & 0 & \lambda_4^{-p} \end{pmatrix}$	(4.9.19)
	$\mathbf{A}^{-p}$ is positive definite only if p is even.	
	$\exp(p\mathbf{A}) = \sum_{k=0}^{\infty} \frac{(p\mathbf{A})^k}{k!}$	(4.9.20)
	$\implies \exp(p\mathbf{A}) - \mathbf{I} = \mathbf{P}\exp(p\mathbf{D})\mathbf{P}^T - \mathbf{P}\mathbf{I}\mathbf{P}^T$	(4.9.21)
Choice 4	$\Rightarrow \exp(p\mathbf{A}) - \mathbf{I} = \mathbf{P}\exp(p\mathbf{D})\mathbf{P}^{T} - \mathbf{P}\mathbf{I}\mathbf{P}^{T}$ $= \mathbf{P}(\exp(p\mathbf{D}) - \mathbf{I})\mathbf{P}^{T}$	(4.9.22)
	$= \mathbf{P}(\exp(p\mathbf{D}) - \mathbf{I})\mathbf{P}^{T}$ $\exp(p\mathbf{D}) - \mathbf{I} = \begin{pmatrix} e^{\lambda_{1}} - 1 & 0 & 0 & 0\\ 0 & e^{\lambda_{2}} - 1 & 0 & 0\\ 0 & 0 & e^{\lambda_{3}} - 1 & 0\\ 0 & 0 & 0 & e^{\lambda_{4}} - 1 \end{pmatrix}$	(4.9.23)
	A is non-singular	
	$\implies \forall i \in [1,4], \lambda_i \neq 0$	(4.9.24)
	$e^{\lambda_i} < 1$	(4.9.25)
	So, $\exp(p\mathbf{A}) - \mathbf{I}$ is not positive definite.	. ,

TABLE 4.9.1: Solution

Given	Derivation	
Given	A is a $m \times n$ matrix of rank $m$ with $n > m$ .	
	A non-zero real number α.	
	To find eigenvalues of A <sup>T</sup> A.	
Eigenvalues of AAT	$AA^T$ is a $m \times m$ matrix and $A^TA$ is a $n \times n$ matrix.	
	Let, $\lambda$ be a non-zero eigen value of $\mathbf{A}^{\mathrm{T}}\mathbf{A}$ .	
	$\mathbf{A}^{\mathbf{T}}\mathbf{A}\mathbf{v} = \lambda \mathbf{v}  \mathbf{v} \in \mathbf{R}^{\mathbf{n}} \tag{4.10.1}$	
	$\mathbf{A}\mathbf{A}^{T}\mathbf{A}\mathbf{v} = \lambda \mathbf{A}\mathbf{v} \tag{4.10.2}$	
	Let, $\mathbf{x} = \mathbf{A}\mathbf{v}  \mathbf{x} \in \mathbf{R}^{\mathbf{m}}$ (4.10.3)	
	$\mathbf{A}\mathbf{A}^{\mathrm{T}}\mathbf{x} = \lambda \mathbf{x} \tag{4.10.4}$	
	$\mathbf{x}^{T} \mathbf{A} \mathbf{A}^{T} \mathbf{x} = \lambda \mathbf{x}^{T} \mathbf{x} \tag{4.10.5}$	
	Given, $\mathbf{x}^{\mathrm{T}} \mathbf{A} \mathbf{A}^{\mathrm{T}} \mathbf{x} = \alpha \mathbf{x}^{\mathrm{T}} \mathbf{x}$ (4.10.6)	
	$\implies \alpha \mathbf{x}^{T} \mathbf{x} = \lambda \mathbf{x}^{T} \mathbf{x} \tag{4.10.7}$	
	From equation (4.10.7), $\lambda = \alpha$ as $\ \mathbf{x}\  \neq 0$	
	As $rank(\mathbf{A}^T\mathbf{A}) = rank(\mathbf{A}) = m$ and equation (4.10.7) satisfies the condition in question.	
	Therefore the only non-zero eigen value is $\alpha$	
	$A^TA$ has an eigenvalue $\alpha$ with multiplicity $m$ .	
Eigenvalues of A <sup>T</sup> A	$A^{T}A$ is a $n \times n$ matrix. Given $n > m$ ,	
	TV I I I I I I I I I I I I I I I I I I I	
	We know that, A <sup>T</sup> A and AA <sup>T</sup> have same number of non-zero eigenvalues	
	and if one of them has more number of eigenvalues than the other	
	then these eigenvalues are zero.  1. From above, as $\alpha$ is non-zero, $A^TA$ has $\alpha$ as its eigenvalue with multiplicity $m$	
	1. From above, as $\alpha$ is non-zero, $\mathbf{A}^{-}\mathbf{A}$ has $\alpha$ as its eigenvalue with multiplicity $m$ 2. $\mathbf{A}^{T}\mathbf{A}$ has 0 as its eigenvalue with multiplicity $n-m$	
	2. A A has 0 as its eigenvalue with multiplicity $n-m$ 3. Therefore, the two distinct eigenvalues of $A^TA$ are $\alpha$ and 0.	
	5. Therefore, the two distinct eigenvalues of A A are a and 0.	

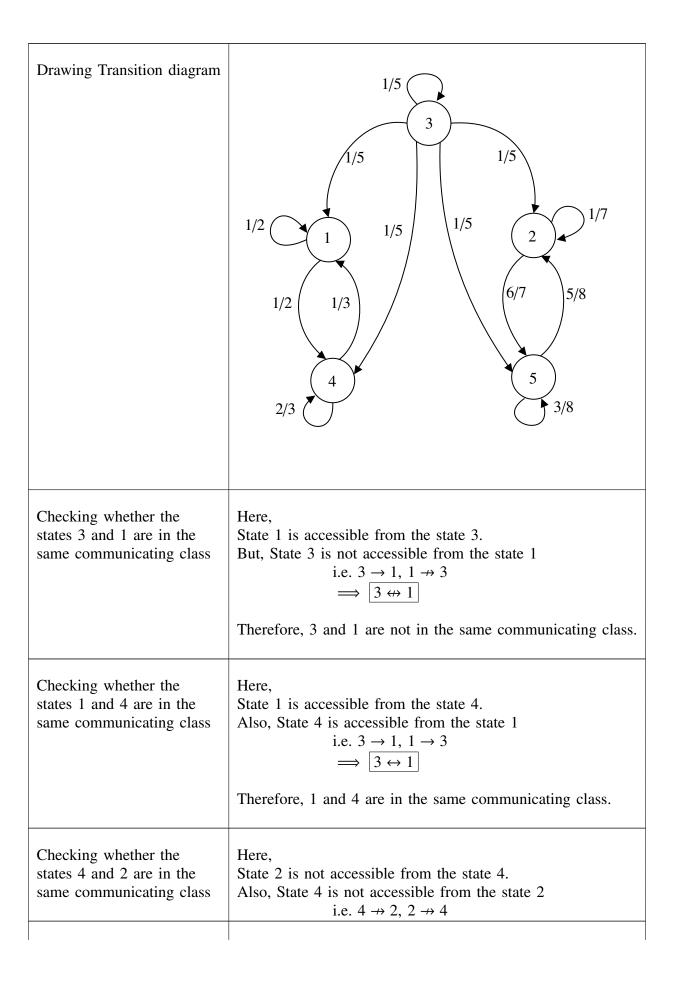
TABLE 4.10.1: Explanation

$\mathbf{A}^{T}\mathbf{A}$ has exactly two distinct eigenvalues.	True statement
$\mathbf{A}^{T}\mathbf{A}$ has 0 as an eigenvalue with multiplicity $n-m$	True statement
${f A}^{ m T}{f A}$ has $lpha$ as a non-zero eigenvalue	True statement
<b>A</b> <sup>T</sup> <b>A</b> has exactly two non-zero distinct eigenvalues.	False statement

TABLE 4.10.2: Solution

Accessibility of states in Markov's chain	We say that state $j$ is accessible from state $i$ , written as $i \to j$ , if $p_{ij}^{(n)} > 0$ for some n. Every state is accessible from itself since $p_{ii}^{(0)} = 1$
Communication between states	Two states $i$ and $j$ are said to communicate, written as $i \leftrightarrow j$ , if they are accessible from each other. In other words, $i \leftrightarrow j \text{ means } i \to j \text{ and } j \to i.$
Communicating class	For each Markov chain, there exists a unique decomposition of the state space $S$ into a sequence of disjoint subsets $C_1, C_2,,$ $S = \bigcup_{i=1}^{\infty} C_i$ in which each subset has the property that all states within it communicate. Each such subset is called a communication class of the Markov chain.

TABLE 4.11.1: Definition and Result used



	$\implies \boxed{4 \leftrightarrow 2}$ Therefore, 4 and 2 are not in the same communicating class.
Checking whether the states 2 and 5 are in the same communicating class	Here, State 2 is accessible from the state 5. Also, State 5 is accessible from the state 2 i.e. $5 \rightarrow 2$ , $2 \rightarrow 5$ $\Rightarrow 2 \leftrightarrow 5$ Therefore, 2 and 5 are in the same communicating class.
Conclusion	Communication classes are: $S = \{1, 4\} \cup \{3\} \cup \{2, 5\}$ Option 2) and 4) are true.

TABLE 4.11.2: Solution

#### 5 June 2017

5.1. Let **A** be an  $n \times n$  self-adjoint matrix with eigenvalues  $\lambda_1, \dots, \lambda_2$ . Let,

$$\|\mathbf{X}\|_2 = \sqrt{|\mathbf{X}_1^2| + \dots + |\mathbf{X}_n^2|}$$
 (5.1.1)

for  $\mathbf{X}=(\mathbf{X}_1,\cdots,\mathbf{X}_n)\in\mathbb{C}^n$ . If

$$p(\mathbf{A}) = a_0 \mathbf{I} + a_1 \mathbf{A} + \dots + a_n \mathbf{A}^n \qquad (5.1.2)$$

then  $\sup_{\|\mathbf{X}\|_2=1} \|p(\mathbf{A})\mathbf{X}\|_2$  is equal to

**Solution:** We know that **A** is a self adjoint matrix and hence  $\mathbf{A} = \mathbf{A}^*$  with eigen values  $\lambda_1, \lambda_2, \dots, \lambda_n$ . Now as we are given,

$$p(\mathbf{A}) = a_0 \mathbf{I} + a_1 \mathbf{A} + \dots + a_n \mathbf{A}^n \qquad (5.1.3)$$

then,

$$(p(\mathbf{A}))^* = a_0 \mathbf{I}^* + a_1 \mathbf{A}^* + \dots + a_n (\mathbf{A}^*)^n \quad (5.1.4)$$

Since,  $A = A^*$  we can state that,

$$p(\mathbf{A})(p(\mathbf{A}))^* = (p(\mathbf{A}))^* p(\mathbf{A})$$
 (5.1.5)

Hence p(A) is a normal matrix. Now using spectral theorem for a normal matrix,

$$||p(\mathbf{A})||_2 = \rho(p(\mathbf{A}))$$
 (5.1.6)

sup refers to the smallest element that is greater than or equal to every number in the set. Hence, sup of  $||p(\mathbf{A})||_2$  will be,

= max { $|\alpha|$  :  $\alpha$  is the eigen value of p(A)} (5.1.7)

= 
$$\max\{|p(\lambda_j)| : j = 1, 2, \dots n\}$$

(5.1.8)

$$= \max\{|a_0 + a_1\lambda_j + \dots + a_n\lambda_j^n| : j = 1, 2, \dots n\}$$
(5.1.9)

Now, to find  $\sup ||p(\mathbf{A})\mathbf{X}||_2$ ,

$$= \max\{|a_0 + a_1\lambda_j + \dots + a_n\lambda_j^n| : j = 1, 2, \dots n\} \|\mathbf{X}\|_2$$
(5.1.10)

Since, we have to find  $\sup_{\|\mathbf{X}\|_2=1}$  i.e,

$$\|\mathbf{X}\|_2 = \sqrt{|\mathbf{X}_1^2| + \dots + |\mathbf{X}_n^2|} = 1$$
 (5.1.11)

Hence the final answer will be,

$$= \max\{|a_0 + a_1\lambda_j + \dots + a_n\lambda_j^n| : j = 1, 2, \dots n\}$$
(5.1.12)

5.2. Let  $p(x) = \alpha x^2 + \beta x + \gamma$  be a polynomial, where

 $\alpha, \beta, \gamma \in R$ . Fix  $X_0 \in R$ . Let  $S = \{(a, b, c) \in R^3 : p(x) = a(x - x_0)^2 + b(x - x_0) + c\}$  for all  $x \in R$ . Find the number of elements in S is

- a) 0
- b) 1
- c) Strictly greater than 1 but finite
- d) Infinite

**Solution:** 

$$p(x) = \alpha x^2 + \beta x + \gamma \tag{5.2.1}$$

$$\implies p(x) = (\alpha\beta\gamma)(x^2x1)^T$$
 (5.2.2)

$$S = \{(a, b, c) \in \mathbb{R}^3 : p(x) = a(x - x_0)^2 + b(x - x_0) + c\},\$$

$$\forall \mathbf{x} \in R (FixX_0) \tag{5.2.3}$$

$$p(x) = (abc) ((x - x_0)^2 (x - x_0)1)^T$$
 (5.2.4)

$$= a(x^2 - 2x_0x + x_0^2) + b(x - x_0) + c \quad (5.2.5)$$

$$= ax^{2} + (b - 2ax_{0})x + (ax_{0}^{2} - bx_{0} + c)$$
(5.2.6)

Refer (5.2.2) and (5.2.6) and comparing the cocoefficients of powers of x,

$$\alpha = a, \beta = b - 2ax_0, \gamma = ax_0^2 - bx_0 + c$$
(5.2.7)

$$a = \alpha, b = \beta + 2\alpha x_0, c = \gamma - \alpha {x_0}^2 + (\beta + 2\alpha x_0) x_0$$
(5.2.8)

Here  $\alpha, \beta, \gamma$  and  $x_0$  are the real fixed numbers. So a, b, c have unique values.

Hence S contain only 1 element. So option 2 is correct

5.3. Let

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 2 \\ 1 & -2 & 0 \\ 0 & 0 & -3 \end{pmatrix} \tag{5.3.1}$$

and I be the  $3 \times 3$  identity matrix. If

$$6\mathbf{A}^{-1} = a\mathbf{A}^2 + b\mathbf{A} + c\mathbf{I} \tag{5.3.2}$$

for  $a, b, c \in \mathbb{R}$  then (a,b,c) equals a) (1,2,1)

- b) (1,-1,2)
- c) (4,1,1)

d) (1,4,1)

**Solution:** Finding the characteristic equation,

$$\begin{vmatrix} \mathbf{A} - \lambda \mathbf{I} \end{vmatrix} = \begin{vmatrix} 1 - \lambda & 0 & 2 \\ 1 & -2 - \lambda & 0 \\ 0 & 0 & -3 - \lambda \end{vmatrix}$$
 (5.3.3)

$$\implies (1 - \lambda)(-2 - \lambda)(-3 - \lambda) = 0 \quad (5.3.4)$$

$$\implies (\lambda^2 + \lambda - 2)(-3 - \lambda) = 0 \quad (5.3.5)$$

$$\implies \lambda^3 + 4\lambda^2 + \lambda - 6 = 0 \quad (5.3.6)$$

Using Cayley-Hamilton Theorem we get,

$$\mathbf{A}^3 + 4\mathbf{A}^2 + \mathbf{A} - 6\mathbf{I} = 0 \tag{5.3.7}$$

$$\implies \mathbf{A}^3 + 4\mathbf{A}^2 + \mathbf{A} = 6\mathbf{I} \tag{5.3.8}$$

$$\implies \mathbf{A}(\mathbf{A}^2 + 4\mathbf{A} + \mathbf{I}) = 6\mathbf{I} \tag{5.3.9}$$

 $|\mathbf{A}| = 6 \neq 0$  hence inverse exists. Hence (5.3.9) we get,

$$6\mathbf{A}^{-1} = \mathbf{A}^2 + 4\mathbf{A} + \mathbf{I}$$
 (5.3.10)

Comparing (5.3.2) and (5.3.10) we get,

$$a = 1$$
  $b = 4$   $c = 1$  (5.3.11)

Hence (a, b, c) = (1, 4, 1)

5.4. Find the Eigenvalues of the matrix,

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 2 \\ 1 & -2 & 5 \\ 2 & 5 & -3 \end{pmatrix} \tag{5.4.1}$$

- a) -4, 3, -3
- b) 4, 3, 1
- c) 4,  $-4 \pm \sqrt{13}$
- d) 4,  $-2 \pm \sqrt{7}$

**Solution:** Using the characteristic equation of the matrix can find the Eigenvalues,

$$\left| \lambda \mathbf{I} - \mathbf{A} \right| = 0 \qquad (5.4.2)$$

$$\implies \begin{vmatrix} \lambda - 1 & -1 & -2 \\ -1 & \lambda + 2 & -5 \\ -2 & -5 & \lambda + 3 \end{vmatrix} = 0 \quad (5.4.3)$$

The expression that is obtained after expanding the determinant and simplifying it is,

$$(\lambda - 1)(\lambda^2 + 5\lambda - 19) - (5\lambda + 31) = 0 \quad (5.4.4)$$

Further simplifying this we obtain the cubic

equation,

$$\lambda^3 + 4\lambda^2 - 29\lambda - 12 = 0 \tag{5.4.5}$$

Solving this equation, the Eigenvalues obtained are,

$$\lambda_1 = -7.605$$
,  $\lambda_2 = -0.394$  and  $\lambda_3 = 4$  (5.4.6)

Therefore, the Eigenvalues of the given matrix are 4,  $-4 \pm \sqrt{13}$  (Option 3)

5.5. Consider the vector space V of real polynomials of degree less than or equal to n. Fix distinct real numbers  $a_0, a_1, \dots, a_k$ . For  $p \in V$ 

$$max\{|p(a_j)|: 0 \le j \le k\}$$
 (5.5.1)

defines a norm on V

- a) only if k < n
- b) only if  $k \ge n$
- c) if  $k + 1 \le n$
- d) if  $k \ge n + 1$

**Solution:** Options 2 and 4 are correct as verified in the table 5.5.2

The scalar multiplication and triangle inequality properties holds true for all k.

$$max\{|\alpha p(a_j)|\} = |\alpha| max\{|p(a_j)|\}$$
 (5.5.4)

$$max\{|p(a_i) + p(a_j)|\} \le max\{|p(a_i)|\} + max\{|p(a_j)|\}$$
(5.5.5)

The positivity property holds true only if  $k \ge n$  as more than n roots are possible when,

$$p(x) = 0 \implies |p(a_j)|_{0 \le j \le k} = 0$$
 (5.5.6)

$$\implies \max\{|p(a_j)|: 0 \le j \le k\} = 0 \quad (5.5.7)$$

5.6. Let **V** be the vector space of polynomials of degree at most 3 in a variable x with coefficients in  $\mathbb{R}$ . Let  $\mathbf{T}=d/dx$  be the linear transformation of **V** to itself given by differentiation.

Which of the following are correct?

- a) T is invertible
- b) 0 is an eigenvalue of **T**
- c) There is a basis with respect to which the matrix of **T** is nilpotent.
- d) The matrix of **T** with respect to the basis

Properties	<b>Norm</b> $\forall x \in V$
Positivity	$  x   \ge 0,   x   = 0 \iff x = 0$
Scalar Multiplication	$  \alpha x   =  \alpha     x   , \alpha \in F$
Triangle Inequality	$  x + y   \le   x   +   y  $

TABLE 5.5.1: Properties of Norm

For $p \in V$ then the norm, $max\{ p(a_j) : 0 \le j \le k\} = 0 \iff  p(a_j) _{0 \le j \le k} = 0$	
Conditions	Explanation
only if $k < n$	A polynomial doesn't necessarily have k distinct real roots,
	i.e., it may have repeated, complex roots.
Example:	let $p$ be polynomial of degree $n = 2$ and $k = 1$ given by:-
	$p(x) = x^2 + 4x + 4   (5.5.2)$
	$ p(a_j) _{0 \le j \le 1} = 0 \implies a_0 = -2, a_1 = -2$ (5.5.3)
	but $a_0, a_1, \dots, a_k$ should be distinct real numbers.
	This contradicts the property of Norm. Thus condition fails.
only if $k \ge n$	p is a polynomial of degree ≤n,
	it can't have more than $n$ roots and is only possible when,
	$p(x) = 0 \implies \left  p(a_j) \right _{0 < j < k} = 0$
	hence $p$ is identically zero. Thus condition satisfies.
if $k + 1 \le n$	Not a norm for $k < n$ . Hence incorrect.
if $k \ge n + 1$	Norm for $k \ge n$ . Hence correct.

TABLE 5.5.2: Verifying Positivity Property of Norm

$$(1, 1 + x, 1 + x + x^2, 1 + x + x^2 + x^3)$$
 is diagonal.

**Solution:** See Tables 5.6.1 , 5.6.2 and 5.6.3.

Nilpotent Matrix	<ol> <li>If all the eigen values of matrix is zero then it is said to nilpotent matrix</li> <li>Determinant and trace of nilpotent matrix are always zero.</li> </ol>
Invertible Matrix	A matrix is said to be invertible matrix if its determinant is non zero.
Diagonal matrix	diagonal matrix is a matrix in which the entries outside the main diagonal are all zero.

TABLE 5.6.1: Definition

Given 
$$T: P_3 \to P_3$$
  
 $T: V \to V$  be the linear operator given by differentiation wrt  $x$   
 $T(P(x)) \to P'(x)$   
A be the matrix of  $T$  wrt some basis for  $V$   
Assume basis for  $V$  be  $\{1, x, x^2, x^3\}$ 

TABLE 5.6.2: Result used

Checking whether matrix of $T$ is nilpotent  Checking eigen value of matrix $T$	$T: V \to V$ $TP(x) = P'(x)$ Differentiating wrt x to find matrix A; $T(1) = 0 = a_1x + b_1x + c_1x^2 + d_1x^3$ $T(x) = 1 = a_2 + b_2x + c_2x^2 + d_2x^3$ $T(x^2) = 2x = a_3 + b_3x + c_3x^2 + d_3x^3$ $T(x^3) = 3x^2 = a_4 + b_4x + c_4x^2 + d_4x^3$ Representing A in matrix form; $A = \begin{cases} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{cases}$ from the above matrix of T we can say it is nilpotent matrix. $A = \begin{pmatrix} 0 - \lambda & 1 & 0 & 0 \\ 0 & 0 - \lambda & 2 & 0 \\ 0 & 0 & 0 - \lambda & 3 \\ 0 & 0 & 0 & 0 - \lambda \end{pmatrix}$
Checking whether matrix of <i>T</i> is invertible	$\Rightarrow \lambda = 0$ Since det $A = 0$ . Therefore matrix of $T$ is not invertible
Checking whether Matrix of <i>T</i> is diagonal matrix	Let basis be $B' = \{1, 1 + x, 1 + x + x^2, 1 + x + x^2 + x^3\}$ Differentiating wrt $x$ ;

	$T(1) = 0 = a_1x + b_1(1+x) + c_1(1+x+x^2) + d_1(1+x+x^2+x^3)$ $T(1+x) = 1 = a_2 + b_2(1+x) + c_2(1+x+x^2) + d_2(1+x+x^2x^3)$ $T(1+x+x^2) = 1 + 2x = a_3 + b_3(1+x) + c_3(1+x+x^2)$ $+d_3(1+x+x^2+x^3)$ $T(1+x+x^2+x^3) = 1 + 2x + 3x^2 = a_4 + b_4(1+x) + c_4(1+x+x^2)$ $+d_4(1+x+x^2+x^3)$ $B = \begin{pmatrix} 0 & 1 & -1 & -1 \\ 0 & 0 & 2 & -1 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$ above matrix is not a diagonal matrix
Conclusion	Thus we can conclude Option 2) and 3) are correct.

TABLE 5.6.3: Solution

- 5.7. Let m, n, r be natural numbers. Let A be an  $m \times n$  matrix with real entries such that  $(AA^t)^r = I$ , where I is the  $m \times m$  is identity matrix and  $A^t$  is the transpose of the matrix A. We can conclude that
  - a) m = n
  - b)  $AA^{t}$  is invertible
  - c)  $A^tA$  is invertible
  - d) if m = n, then A is invertible

**Solution:** Options 2) and 4) are correct. See Table 5.7.1

- 5.8. For any  $n \times n$  matrix B, let  $N(B) = \{X \in \mathbb{R}^n : BX = 0\}$  be the null space of B. Let A be a  $4 \times 4$  matrix with dim(N(A 4I)) = 2, dim(N(A 2I)) = 1 and rank(A) = 3 Which of the following are true?
  - a) 0,2 and 4 are eigenvalues of A
  - b) determinant(A)=0
  - c) A is not diagonalizable
  - d) trace(A)=8

**Solution:** See Table 5.8.1.

Given	A is a $4 \times 4$ matrix. dim(N(A-2I)) = 2, dim(N(A-4I)) = 1, and rank(A) = 3
-------	--

Eigenvalues of a matrix

The number  $\lambda$  is an eigenvalue of a matrix A if and only if  $A - \lambda I$  is singular, i.e.  $|A - \lambda I| = 0$ 

For  $\lambda = 2$ Given, dim(N(A - 2I)) = 2  $\implies nullity(A - 2I) = 2$  rank(A) + nullity(A) = n  $\implies rank(A - 2I) = 4 - 2 = 2$   $\implies (A - 2I)$  is not a full rank matrix Therefore |A - 2I| = 0

Also,

 $\Rightarrow N(A - 2I) = \{X \in \mathbb{R}^4 : (A - 2I)X = 0\}$   $\Rightarrow (A - 2I)X = 0 \text{ gives two eigen vectors}$   $\Rightarrow 2 \text{ is an eigenvalue of } A \text{ with multiplicity } 2.$ 

Similarly, for  $\lambda = 4$ Given, dim(N(A - 4I)) = 1  $\implies rank(A - 4I) = 4 - 1 = 3$  $\implies (A - 4I)$  is not a full rank matrix

	Therefore $ A - 4I  = 0$ $\Rightarrow 4$ is an eigenvalue of $A$ with multiplicity 1. For $\lambda = 0$ Given that $rank(A) = 3$ $\Rightarrow A$ is not a full rank matrix Therefore $ A  = 0$ $\Rightarrow 0$ is an eigenvalue of $A$ with multiplicity 1.
Determinant	Given that $rank(A) = 3$ $\implies A$ is not a full rank matrix Therefore $ A  = 0$
Diagonalizability	An $n \times n$ matrix $A$ is diagonalizable if and only if $A$ has n linearly independent eigen vectors. $rank(A) + nullity(A) = n$ $\implies$ for $\lambda = 0$ , $nullity(A - \lambda I) = nullity(A) = 4 - 3 = 1$ $\implies$ There exists only one linearly independent eigen vector corresponding to 0 eigen value Thus, matrix $A$ is not diagonalizable.
Trace	Trace(A)=sum of eigen values $\implies Trace(A) = 0 + 2 + 2 + 4 = 8$
Conclusion	Option (1), (2) and (4) are correct

TABLE 5.8.1: Solution

5.9. Which of the following 3x3 matrices are diagonizable over  $\mathbb{R}$ ?

a) 
$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{pmatrix}$$
b) 
$$\begin{pmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
c) 
$$\begin{pmatrix} 1 & 2 & 3 \\ 2 & 1 & 4 \\ 3 & 4 & 1 \end{pmatrix}$$
d) 
$$\begin{pmatrix} 0 & 1 & 2 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

Option	Answer
1) <i>m</i> = <i>n</i>	Let $\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$ and $r = 1$ $(\mathbf{A}\mathbf{A}^{\mathrm{T}})^r = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = I$ Since $m \neq n$ Option 1 is False.
2) $AA^t$ is invertible	w.k.t $det(A^n) = (det(A))^n$ Since $(AA^t)^r = I$ So $det((AA^T)^r) = det(I)$ $(det(AA^T))^r = 1$ $\implies det(AA^T) \neq 0$ Hence $AA^T$ is invertible Option 2 is True.
3) A <sup>t</sup> A is invertible	Let $\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$ and $r = 1$ $(\mathbf{A}^T \mathbf{A})^r = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$ But $\det(AA^T) = 0$ . $\implies AA^T \text{ is not invertible.}$ Hence Option 3 is False
4) if $m = n$ then $A$ is invertible	Since $det(AA^T) \neq 0$ $det(A).det(A^T) \neq 0$ $det(A).det(A) \neq 0$ $\implies A$ is invertible. Hence Option 4 is True

TABLE 5.7.1

**Solution:** See Tables 5.9.1 and 5.9.2

Test for diagonalizability	Let $\mathbf{W}_i$ be the eigenspace corresponding to eigenvalue $\lambda_i$ of $\mathbf{A}$
	1) <b>A</b> is diagonalizable
	2) characteristic polynomial of <b>A</b> is
	$f = (\mathbf{x} - \lambda_1)^{d_1}(\mathbf{x} - \lambda_k)^{d_k}$ and $dim(\mathbf{W}_i) = d_i$
	$3) \sum_{i=1}^k \mathbf{W_i} = n$
Concept	A linear operator <b>A</b> on a <i>n</i> -dimensional space $\mathbb{V}$ is
for diagonalization	diagonalizable, if and only if $A$ has $n$ distinct
	characteristic vectors or null spaces corresponding to the characteristic values

TABLE 5.9.1: Illustration of theorem.

Option A	Given matrix is $\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{pmatrix}$
Finding Characteristics polynomial	Characteristics polynomial of the matrix <b>A</b> is $det(x\mathbf{I} - \mathbf{A})$ $det(x\mathbf{I} - \mathbf{A}) = \begin{vmatrix} (x-1) & -3 & -2 \\ 0 & (x-4) & -5 \\ 0 & 0 & x-6 \end{vmatrix}$ Characteristic Polynomial = $(x-1)(x-4)(x-6)$
Testing diagonalizability over ℝ	1) As the characteristics polynomial is product of linear factors over $\mathbb{R}$ .  2) To find characteristic values of the operator $\det(xI-A)=0$ which gives $\lambda_1=1, \lambda_2=4, \lambda_3=6$ Thus over $\mathbb{R}$ matrix $\mathbf{A}$ has three distinct characteristic values. There will be atleast one characteristics vector i.e., one dimension with each characteristics value . Thus $\dim \mathbf{W}_i=d_i$ 3) $\sum_i \mathbf{W}_i=n=3$ , which is equal to $\dim \mathbf{A}$ .

Conclusion on Option A	Option A satisfy all three condition of Diagonalizability over $\mathbb{R}$ .
Option B	Given matrix is $\mathbf{A} = \begin{pmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
Finding Characteristics polynomial	Characteristics polynomial of the matrix $det(x\mathbf{I} - \mathbf{A})$ $det(x\mathbf{I} - \mathbf{A}) = \begin{vmatrix} x & -1 & 0 \\ 1 & x & 0 \\ 0 & 0 & x - 1 \end{vmatrix}$ Characteristic Polynomial = $(x - 1)(x + i)(x - i)$
Testing diagonalizability over $\mathbb R$	1) As the characteristics polynomial is not the product of linear factors over $\mathbb R$ beacuse roots of characteristic eq are complex . Thus $\mathbf A$ is not diagonalizable over $\mathbb R$ .
Conclusion on Option B	Option B does not satisfy condition 1.
Option C	Given matrix is $ \mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 1 & 4 \\ 3 & 4 & 1 \end{pmatrix} $
Finding Characteristics polynomial	Characteristics polynomial of the matrix <b>A</b> is $det(x\mathbf{I} - \mathbf{A})$ $det(x\mathbf{I} - \mathbf{A}) = \begin{vmatrix} (x-1) & -2 & -3 \\ -2 & (x-1) & -4 \\ -3 & -4 & x-1 \end{vmatrix}$ Characteristic Polynomial = $(x + 3.19)(x + 0.877)(x - 7.07)$
Testing diagonalizability over R	<ol> <li>As the characteristics polynomial are product of linear factors over ℝ.</li> <li>To find characteristic values of the operator det(x<b>I</b> – <b>A</b>) = 0 which gives λ<sub>1</sub> = -3.19, λ<sub>2</sub> = -0.887, λ<sub>3</sub> = 7.07</li> </ol>

	Thus over $\mathbb{R}$ matrix $\mathbf{A}$ has three distinct characteristic values. There will be at least one characteristics vector i.e., one dimension with each characteristics value. Thus $dim\mathbf{W}_i = d_i$ 3) $\sum_i \mathbf{W}_i = n = 3$ , which is equal to $dim$ of $\mathbf{A}$ .
Conclusion on Option C	Option C satisfy all three condition of Diagonalizability over $\mathbb{R}$ .
Option D	Given matrix is $ A = \begin{pmatrix} 0 & 1 & 2 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} $
Finding Characteristics polynomial	Characteristics polynomial of the matrix <b>A</b> is $det(x\mathbf{I} - \mathbf{A})$ $det(x\mathbf{I} - \mathbf{A}) = \begin{vmatrix} x & -1 & -2 \\ 0 & x & -1 \\ 0 & 0 & x \end{vmatrix}$ Characteristic Polynomial = $(x)(x)(x) = x^3$
Testing diagonalizability over $\mathbb{R}$	1) As the characteristics polynomial is product of linear factors over $\mathbb{R}$ .  2) To find characteristic values of the operator $\det(x\mathbf{I} - \mathbf{A}) = 0$ $\lambda_1 = 0$ $d_1 = 3$ $\mathbf{W}_1 = \mathbf{A} - \lambda_1 \mathbf{I} \implies \begin{pmatrix} 0 & 1 & 2 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} - 0 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 2 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$ $dim \mathbf{W}_1 = 2$ $dim \mathbf{W}_i \neq d_i$ Algebric Multiplicity is not equal to Geometric Multiplicity.
Conclusion on Option D	Option D does not satisfy second condition of Diagonalizability.
Answer	Option A and Option C are Diagonalizable over $\mathbb{R}$ .

TABLE 5.9.2: Option Checking Table

Positive Semi Definite Matrix	A $n \times n$ symmetric real matrix $\mathbf{M}$ is said to be positive semi definite if $\mathbf{x}^T \mathbf{M} \mathbf{x} \ge 0$ for all non-zero $\mathbf{x}$ in $\mathbb{R}^n$ . Formally $\mathbf{M}$ is positive semi-definite $\Leftrightarrow \mathbf{x}^T \mathbf{M} \mathbf{x} \ge 0 \ \forall \ \mathbf{x} \in \mathbb{R}^n \setminus \{0\}$
Theorem	For a symmetric $n \times n$ matrix $\mathbf{M} \in \mathbf{L}(\mathbf{V})$ , following are equivalent. 1). $\mathbf{x}^{\mathbf{T}} \mathbf{M} \mathbf{x} \ge 0 \ \forall \ \mathbf{x} \in \mathbf{V}$ . 2). All the eigenvalues of $\mathbf{M}$ are non-negative.

TABLE 5.10.1: Definition and Result used

Calculating eigen values of A	Given $\mathbf{A} = \begin{pmatrix} 3 & 1 & 2 \\ 1 & 2 & 3 \\ 2 & 3 & 1 \end{pmatrix}$ Calculating, eigen values of $\mathbf{A}$ , ie $\det(\mathbf{A} - \lambda \mathbf{I}) = 0$ $\Rightarrow \begin{pmatrix} 3 - \lambda & 1 & 2 \\ 1 & 2 - \lambda & 3 \\ 2 & 3 & 1 - \lambda \end{pmatrix} = 0$ $\Rightarrow (3 - \lambda)((2 - \lambda)((1 - \lambda) - 9) - 1(1 - \lambda - 6) + 2(3 - 2(2 - \lambda))) = 0$
	$\Rightarrow \lambda^3 - 6\lambda^2 - 3\lambda + 18 = 0$ $\Rightarrow \lambda_1 = 6, \lambda_2 = \sqrt{3} \text{ and } \lambda_3 = -\sqrt{3}$ Hence, A has exactly two positive eigen values.
Proving $\mathbf{x}^T \mathbf{A} \mathbf{x} < 0$ for some $\mathbf{x} \in \mathbb{R}^3$ using contradiction	Suppose $\mathbf{x}^T \mathbf{A} \mathbf{x} \geq 0$ for all $\mathbf{x} \in \mathbb{R}^3$ . Then, by theorem above in definition section, matrix $\mathbf{A}$ is positive semi definite. Hence, all the eigen values of $\mathbf{A}$ non-negative, but this is not the case as one of eigen value is $\lambda_3 = -\sqrt{3}$ . So, $\mathbf{x}^T \mathbf{A} \mathbf{x} \geq 0$ is not true for all $\mathbf{x} \in \mathbb{R}^3$ . Similarly, as $\lambda_1 \leq 0$ , $\forall i$ is also not true, so $\mathbf{x}^T \mathbf{A} \mathbf{x} \leq 0$ is not true for all $\mathbf{x} \in \mathbb{R}^3$ . Thus, $\mathbf{x}^T \mathbf{A} \mathbf{x} < 0$ for some $\mathbf{x} \in \mathbb{R}^3$ .
Correct Options	Hence, correct options are (1) and (4).

TABLE 5.10.2: Solution

5.10. Let 
$$\mathbf{A} = \begin{pmatrix} 3 & 1 & 2 \\ 1 & 2 & 3 \\ 2 & 3 & 1 \end{pmatrix}$$
 and  $\mathbf{Q}(\mathbf{X}) = \mathbf{X}^{T} \mathbf{A} \mathbf{X}$  for  $\mathbf{X} \in$ 

 $\mathbb{R}^3$ . Then

- a) A has exactly two positive eigen values.
- b) all the eigen values of A are positive.
- c)  $\mathbf{Q}(\mathbf{X}) \geq 0 \ \forall \ \mathbf{X} \in \mathbb{R}^3$
- d)  $\mathbf{Q}(\mathbf{X}) < 0$  for some  $\mathbf{X} \in \mathbb{R}^3$

**Solution:** See Tables 5.10.1 and 5.10.2

5.11. Consider the matrix

$$A(x) = \begin{pmatrix} 1 + x^2 & 7 & 11 \\ 3x & 2x & 4 \\ 8x & 17 & 13 \end{pmatrix}; x \in \mathbf{R}.$$
 (5.11.1)

Then,

- a) A(x) has eigenvalue 0 for some  $x \in \mathbf{R}$ .
- b) 0 is not an eigenvalue of A(x) for any  $x \in \mathbf{R}$ .
- c) A(x) has eigenvalue  $0 \ \forall x \in \mathbf{R}$ .
- d) A(x) is invertible  $\forall x \in \mathbf{R}$ .

**Solution:** Let  $\lambda = 0$  be an eigenvalue. Hence,

$$|A - \lambda I| = 0 (5.11.2)$$

$$\implies |A| = 0 (5.11.3)$$

$$\implies |A| = \begin{vmatrix} 1 + x^2 & 7 & 11 \\ 3x & 2x & 4 \\ 8x & 17 & 13 \end{vmatrix} = 0 (5.11.4)$$

Performing row reduction we get,

$$\begin{vmatrix} 1+x^2 & 7 & 11\\ 0 & \frac{2x^3-19x}{1+x^2} & \frac{4x^2-33x+4}{1+x^2}\\ 0 & 0 & \frac{26x^3-244x^2+538x-68}{2x^3-19x} \end{vmatrix} = 0$$
(5.11.5)

$$\implies 26x^3 - 244x^2 + 538x - 68 = 0 \quad (5.11.6)$$

$$\implies x_1 = 6.01, x_2 = 3.23, x_3 = 0.13 \quad (5.11.7)$$

See Table 5.11.1

# **6** December 2016

6.1. The matrix

$$\mathbf{A} = \begin{pmatrix} 3 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 3 \end{pmatrix} \tag{6.1.1}$$

is

- a) positive definite.
- b) non-negative definite but not positive definite.
- c) negative definite.
- d) neither negative definite nor positive definite.

### **Solution:**

a) For a real symmetric matrix to be positive definite the eigen values of the matrix should

OPTIONS	Explanation
Option (b)	At the Values of x given by (5.11.7), eigen value $\lambda = 0$ . Hence option (b) can't be correct.
Option (c)	If one of the eigenvalue is 0 for A(x) then, $ A(x)  = 0 \forall x \in R$ . But from (5.11.7) we have concluded that $ A  = 0$ only for, $x_1 = 6.01, x_2 = 3.23, x_3 = 0.13$ . Hence, Option (c) is incorrect.
Option (d)	Now for the values of x given by (5.11.7), $ A  = 0$ . Hence it is not invertible $\forall x \in \mathbf{R}$ Hence Option (d) is incorrect.
Option (a)	Now clearly from above arguments $A(x)$ has eigenvalue 0 for some $x \in R$ Hence Option (a) is Correct.

TABLE 5.11.1

be positive.

b) For a real symmetric matrix to be negative definite the eigen values of the matrix should be negative.

$$\mathbf{A} = \begin{pmatrix} 3 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 3 \end{pmatrix}$$

The characteristic equation of the matrix **A**is given by

$$\begin{vmatrix} V - \lambda \mathbf{I} \end{vmatrix} = \begin{vmatrix} 3 - \lambda & -1 & 0 \\ -1 & 2 - \lambda & -1 \\ 0 & -1 & 3 - \lambda \end{vmatrix} = 0$$

$$\implies \lambda^3 - 8\lambda^2 + 19\lambda - 12 = 0$$
(6.1.2)

The Eigen values of A are:

$$\lambda_1 = 5/2$$
 $\lambda_2 = 3/2$ 
 $\lambda_3 = 4$ 
(6.1.3)

Since all the eigen values of matrix **A** are positive, Therefore the matrix **A** is positive definite.

- 6.2. Let  $\mathbf{A} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$  and let  $\alpha_n$  and  $\beta_n$  denote the two eigenvalues of  $\mathbf{A}^n$  such that  $|\alpha_n| \ge |\beta_n|$ . Then
  - a)  $\alpha_n \to \infty$  as  $n \to \infty$
  - b)  $\beta_n \to 0$  as  $n \to \infty$
  - c)  $\beta_n$  is positive if n is even.
  - d)  $\beta_n$  is negative if n is odd.

**Solution:** See Table 6.2.1.

- 6.3. Let  $M_n$  denote the vector space of all  $n \times n$  real matrices. Which of the following is a linear subspaces of  $M_n$ :
  - a)  $V_1 = \{A \in M_n : A \text{ is nonsingular}\}\$
  - b)  $V_2 = \{A \in M_n : det(A) = 0\}$
  - c)  $V_3 = \{A \in M_n : trace(A) = 0\}$
  - d)  $V_4 = \{BA : A \in M_n\}$ , where *B* is some fixed matrix in  $M_n$

**Solution:** See Table 6.3.1

- 6.4. If **P** and **Q** are invertible matrices such that PQ = -QP, then we can conclude that
  - a)  $Tr(\mathbf{P}) = Tr(\mathbf{Q}) = 0$
  - b)  $Tr(\mathbf{P}) = Tr(\mathbf{Q}) = 1$
  - c)  $Tr(\mathbf{P}) = -Tr(\mathbf{Q})$
  - d)  $Tr(\mathbf{P}) \neq Tr(\mathbf{Q})$

**Solution:** See Table 6.4.1

Options	Solutions	True/False
1.	Given	
	$\mathbf{A} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$	
	Now lets find the eigen values of matrix A	
	$ \mathbf{A} - \lambda \mathbf{I}  = 0$	
	$\implies \begin{vmatrix} 1 - \lambda & 1 \\ 1 & -\lambda \end{vmatrix} = 0$	
	$\implies \lambda^2 - \lambda - 1 = 0$	True
	On solving we get 2 eigen values	
	$\alpha_1 = \frac{1+\sqrt{5}}{2}$ $\beta_1 = \frac{1-\sqrt{5}}{2}$	
	We know that if eigenvalue of <b>A</b> is $\lambda$ then eigenvalue of <b>A</b> <sup>n</sup> is $\lambda$ <sup>n</sup> .	
	In this problem we can say that the eigenvalues $\alpha_n$ and $\beta_n$ of $\mathbf{A}^n$ are	
	$\alpha_n = \alpha_1^n  \beta_n = \beta_1^n$	
	Since $\alpha_1 > 1$ we can say that $\alpha_n \to \infty$ as $n \to \infty$ .	
2.	We got $\beta_1 = \frac{1-\sqrt{5}}{2}$ and $\beta_n = \beta_1^n$ .	
	Since $-1 < \beta_1 < 0$ , we can say that $\beta_n \to 0$ as $n \to \infty$ .	True
3.	We got $\beta_1 = \frac{1-\sqrt{5}}{2}$ and $\beta_n = \beta_1^n$ .	
	Since $\beta_1$ is negative because $-1 < \beta_1^2 < 0$ , if n is even then $\beta_n$ is positive.	True
4.	We got $\beta_1 = \frac{1-\sqrt{5}}{2}$ and $\beta_n = \beta_1^n$ .	
	Since $\beta_1$ is negative, if n is odd then $\beta_n$ is negative.	True

TABLE 6.2.1

Vector space	Is it subspace to $M_n$ ?
1) $V_1$ : All non-singular matrices of $n \times n$	The matrices $I_{n\times n}$ and $-I_{n\times n}$ are non-singular matrices, but the sum $I_{n\times n} - I_{n\times n}$ is zero matrix and it is singular.
	$\therefore V_1$ does not form subspace of $M_n$ .
2) $V_2$ : All singular matrices of $n \times n$	The matrices $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ and $\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$ are singular matrices, but the sum is a non-singular matrix.
	$\therefore V_2$ does not form subspace $M_n$ .
3) $V_3$ : All matrices of $n \times n$ with trace =0	Let $\mathbf{v_1}$ and $\mathbf{v_2}$ be matrices with Trace = 0.
	$Tr(\mathbf{v}_1 + \alpha \mathbf{v}_2) = Tr(\mathbf{v}_1) + \alpha Tr(\mathbf{v}_2) = 0.$
	$\therefore$ the vector space $V_3$ forms linear subspace of $M_n$ .
4) $V_4$ : $F_A$ = BA, where B is some fixed matrix in $M_n$	Let $\mathbf{v_1}$ and $\mathbf{v_2}$ be matrices in the vector space $V_4$ .
	$F_{v_1+\alpha v_2}=B(\mathbf{v_1}+\alpha \mathbf{v_2})$
	$=B\mathbf{v}_1 + \alpha B\mathbf{v}_2 =$
	$F_{ u_1} + lpha F_{ u_2}.$
	$\therefore V_4$ forms linear subspace of $M_n$ .

TABLE 6.3.1

Given	$\mathbf{P}$ and $\mathbf{Q}$ are invertible matrices. Therefore $\mathbf{P}^{-1}$ and $\mathbf{Q}^{-1}$ exists.	
	PQ = -QP	(6.4.1)
To Prove	$Tr(\mathbf{P})=0$	
Proof 1	Post multiplying equation (6.4.1) by $\mathbf{Q}^{-1}$ we g	get,
	$\mathbf{PQQ}^{-1} = -\mathbf{QPQ}^{-1}$	(6.4.2)
	$\implies \mathbf{PI} = -\mathbf{QPQ}^{-1}$	(6.4.3)
	$\implies \mathbf{P} = -\mathbf{Q}\mathbf{P}\mathbf{Q}^{-1}$	(6.4.4)
	Taking trace on both sides for the equation (6	.4.4),

	$Tr(\mathbf{P}) = Tr(-\mathbf{Q}\mathbf{P}\mathbf{Q}^{-1})$	(6.4.5)
	$\implies Tr(\mathbf{P}) = -Tr(\mathbf{Q}\mathbf{P}\mathbf{Q}^{-1})$	(6.4.6)
	We know that $Tr(\mathbf{AB}) = Tr(\mathbf{BA})$	
	Let $A=Q$ and $B=PQ^{-1}$ From the above property of trace equation (6.	16) can be modified as
	Trom the above property of trace equation (o.	4.0) can be mounted as
	$Tr(\mathbf{P}) = -Tr(\mathbf{P}\mathbf{Q}^{-1}\mathbf{Q})$	(6.4.7)
	$\implies Tr(\mathbf{P}) = -Tr(\mathbf{PI})$	(6.4.8)
	$\implies Tr(\mathbf{P}) = -Tr(\mathbf{P})$	(6.4.9)
	$\implies 2Tr(\mathbf{P}) = 0$	(6.4.10)
	$\implies Tr(\mathbf{P}) = 0$	(6.4.11)
To Prove	$Tr(\mathbf{Q})=0$	
Proof 2	Post multiplying equation (6.4.1) by $\mathbf{P}^{-1}$ we g	get,
	$\mathbf{PQP}^{-1} = -\mathbf{QPP}^{-1}$	(6.4.12)
	$\implies \mathbf{PQP}^{-1} = -\mathbf{QI}$	(6.4.13)
	$\implies \mathbf{PQP}^{-1} = -\mathbf{Q}$	(6.4.14)
	Taking trace on both sides for the equation (6	(4.14).
	$Tr(\mathbf{PQP}^{-1}) = Tr(-\mathbf{Q})$	(6.4.15)
	$\implies Tr(\mathbf{PQP}^{-1}) = -Tr(\mathbf{Q})$	(6.4.16)
	We know that $Tr(AB) = Tr(BA)$	
	Let $A=P$ and $B=QP^{-1}$ From the above property of trace equation (6.	4.16) can be modified as
	$Tr(\mathbf{Q}\mathbf{P}^{-1}\mathbf{P}) = -Tr(\mathbf{Q})$	
	$Tr(\mathbf{QI} - \mathbf{I}) = -Tr(\mathbf{Q})$ $\implies Tr(\mathbf{QI}) = -Tr(\mathbf{Q})$	(6.4.17)
		(6.4.19)
	$\Rightarrow Tr(\mathbf{Q}) = -Tr(\mathbf{Q})$ $\Rightarrow 2Tr(\mathbf{Q}) = 0$	(6.4.20)
	$\Rightarrow Tr(\mathbf{Q}) = 0$ $\Rightarrow Tr(\mathbf{Q}) = 0$	(6.4.21)
~		(0.4.21)
Statement 1 Explanation	$Tr(\mathbf{P})=Tr(\mathbf{Q})=0$ From equation (6.4.11) and (6.4.21) we could	cay that
Explanation	110111 equation (0.4.11) and (0.4.21) we could	say mat,
	$Tr(\mathbf{P}) = Tr(\mathbf{Q}) = 0$	(6.4.22)
	Valid Conclusion	
Statement 2	$Tr(\mathbf{P}) = Tr(\mathbf{Q}) = 1$	•
Explanation	From equation (6.4.11) and (6.4.21) we could	say that,
	$Tr(\mathbf{P}) = Tr(\mathbf{Q}) \neq 1$	(6.4.23)
	Invalid Conclusion	
Statement 3	$Tr(\mathbf{P}) = -Tr(\mathbf{Q})$	
	Substituting the conclusion 1 result equation (	

	$Tr(\mathbf{P}) = -Tr(\mathbf{Q}) \tag{6.4.24}$
	Valid Conclusion
Statement 4	$Tr(\mathbf{P}) \neq Tr(\mathbf{Q})$
Explanation	From equation (6.4.11) and (6.4.21) we could say that,
	$Tr(\mathbf{P}) = Tr(\mathbf{Q}) \tag{6.4.25}$
	Invalid Conclusion

TABLE 6.4.1: Explanation with Proofs

- 6.5. Let  $W_1$ ,  $W_2$ ,  $W_3$  be 3 distinct subspaces of  $\mathbf{R}^{10}$  such that each  $W_i$  has dimension of 9. Let  $\mathbf{W} = \mathbf{W}_1 \cap \mathbf{W}_2 \cap \mathbf{W}_3$ . Then we can conclude that
  - a) W may not be a subspace of  $\mathbf{R}^{10}$
  - b) dim  $\mathbf{W} \le 8$
  - c) dim  $W \ge 7$
  - d) dim  $W \le 3$

**Solution:** See Table 6.5.1

Given	$W_1, W_2, W_3$
Given	are 3 distinct subspaces of
	$\mathbf{R}^{10}$
	Each W <sub>i</sub> has dimension 9
	$\mathbf{W} = \mathbf{W}_1 \cap \mathbf{W}_2 \cap \mathbf{W}_3$
Statement1	W may not be a subspace of $\mathbf{R}^{10}$
Explanation	$As W = W_1 \cap W_2 \cap W_3$
	and $W_1$ , $W_2$ , $W_3$
	are subspaces of W,then W
	must be a subspace of $\mathbf{R}^{10}$ .
	So the first option is false.
Statement2	$\dim \mathbf{W} \leq 8$
Explanation	As W be a subspace of a
	finite dimension vector space $\mathbf{R}^{10}$
	and dim $\mathbf{R}^{10}$ = 10, so $\mathbf{W}$
	is finite dimension and
	$\dim \mathbf{W} \le 10$
Theorem	$\dim (W_1 \cap W_2)$
Theorem	$= \dim(\mathbf{W}_1) + \dim(\mathbf{W}_2) - \dim(\mathbf{W}_1 + \mathbf{W}_2)$
	and
	$\mathbf{W_1} \cap \mathbf{W_2}$ is also a subspace of $\mathbf{R}^{10}$
Proof	The minimum dimension of $W = W_1 \cap W_2 \cap W_3$
Explanation	Let us consider $V = \mathbf{R}^{10}$ and $dim(V) = 10$
	and $\mathbf{U} = \mathbf{W_1} \cap \mathbf{W_2}$
	So, $dim(\mathbf{W_1} \cap \mathbf{W_2} \cap \mathbf{W_3}) = dim(\mathbf{U})$
	$+dim(\mathbf{W_3}) - dim(\mathbf{U} + \mathbf{W_3})$
	or, $dim(\mathbf{W}_1 \cap \mathbf{W}_2 \cap \mathbf{W}_3) = dim(\mathbf{W}_1)$
	$+dim(\mathbf{W_2})+dim(\mathbf{W_3}) - dim(\mathbf{W_1} + \mathbf{W_1})$
	$-dim((\mathbf{W}_1 \cap \mathbf{W}_2) + \mathbf{W}_3)$
	Now (W \circ W) \ W \circ W
	Now, $(\mathbf{W}_1 \cap \mathbf{W}_2) + \mathbf{W}_3 \subseteq \mathbf{V}$ $\implies dim((\mathbf{W}_1 \cap \mathbf{W}_2) + \mathbf{W}_3) \le dim(\mathbf{V})$
	$\implies dim((\mathbf{W}_1 \cap \mathbf{W}_2) + \mathbf{W}_3) \le dim(\mathbf{V})$ $\implies -dim((\mathbf{W}_1 \cap \mathbf{W}_2) + \mathbf{W}_3) \ge -dim(\mathbf{V})$
	Similarly, $(\mathbf{W}_1 + \mathbf{W}_2) \subseteq \mathbf{V}$ $\implies dim(\mathbf{W}_1 + \mathbf{W}_2) \le dim(\mathbf{V})$

	$\implies -dim(\mathbf{W_1} + \mathbf{W_2}) \ge -dim(\mathbf{V})$
	Considering these two inequations, $-dim((\mathbf{W}_1 \cap \mathbf{W}_2) + \mathbf{W}_3) - dim(\mathbf{W}_1 + \mathbf{W}_2)$ $\geq -2dim(\mathbf{V})$
	or, $dim(W_1) + dim(W_2) + dim(W_3)$ $-dim((W_1 \cap W_2) + W_3) - dim(W_1 + W_2)$ $\geq dim(W_1) + dim(W_2) + dim(W_3) - 2dim(V)$
	or, $dim(\mathbf{W}_1 \cap \mathbf{W}_2 \cap \mathbf{W}_3)$ $\geq dim(\mathbf{W}_1) + dim(\mathbf{W}_2) + dim(\mathbf{W}_3) - 2dim(\mathbf{V})$
	$\implies \dim(\mathbf{W}) \ge \dim(\mathbf{W}_1) + \dim(\mathbf{W}_2) + \dim(\mathbf{W}_3) - 2\dim(\mathbf{V})$
Statement 3	dim <b>W</b> ≥ 7
Explanation	As $dim(\mathbf{W}) \ge dim(\mathbf{W}_1) + dim(\mathbf{W}_2)$
	$+dim(\mathbf{W}_3) - 2dim(\mathbf{V})$ $\rightarrow dim(\mathbf{W}) > (0+0+0) - (2\times10)$
	$\implies \dim(\mathbf{W}) \ge (9+9+9) - (2 \times 10)$ $\implies \dim(\mathbf{W}) \ge 7$
Answer	$7 \le \dim(\mathbf{W}) \le 10$

TABLE 6.5.1: Solution summary

Hence, we can conclude that  $dim(\mathbf{W}) \ge 7$ .