

Solutions: Linear Algebra by Hoffman and Kunze



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If

$$x = 0, y = 0 \in Q, a = 0 + \sqrt{2}.0 = 0, 0 \in S$$
(1.1.1.4)

$$x = 1, y = 0, a = 1 + \sqrt{2}.0 = 1, 1 \in S$$
(1.1.1.5)

c)

$$a + b = x + y\sqrt{2} + w + z\sqrt{2} = b + a$$
(1.1.1.6)

d)

$$-a = -x - y\sqrt{2}, x, y \in Qso - x, -y \in Q, a \in S$$
(1.1.1.7)

e)

$$ab = (x + y\sqrt{2})(w + z\sqrt{2}) = ba, ab \in S$$
(1.1.1.8)

f)

$$a^{-1}a = (x + y\sqrt{2})^{-1}(x + y\sqrt{2}) = 1, a^{-1} \in S$$
(1.1.1.9)

Hence (1.1.1.1) ,(1.1.1.2) is verified. Therefore by considering the (1.1.1.1) and (1.1.1.2) we can say set complex numbers of given form $x + y\sqrt{2}$ is subfield of C.

1.1.2. Let \mathbb{F} be the field of complex numbers. Are the following two systems of linear equations equivalent? If so, express each equation in each system as a linear combination of the equations in the other system.

$$x_1 - x_2 = 0$$
$$2x_1 + x_2 = 0$$

and

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

Solution: The given system of linear equations

can be written as,

$$\mathbf{A}\mathbf{x} = 0 \tag{1.1.2.1}$$

$$\implies \begin{pmatrix} 1 & -1 \\ 2 & 1 \end{pmatrix} \mathbf{x} = 0 \tag{1.1.2.2}$$

$$\mathbf{B}\mathbf{x} = 0 \tag{1.1.2.3}$$

$$\Longrightarrow \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \mathbf{x} = 0 \tag{1.1.2.4}$$

Now we can obtain \mathbf{B} from matrix \mathbf{A} by performing elementary row operations given as,

$$\mathbf{B} = \mathbf{C}\mathbf{A} \tag{1.1.2.5}$$

$$\begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} = \mathbf{C} \begin{pmatrix} 1 & -1 \\ 2 & 1 \end{pmatrix} \tag{1.1.2.6}$$

where C is product of elementary matrices given as,

$$\mathbf{C} = (\mathbf{E}_{7}\mathbf{E}_{6}\mathbf{E}_{5}\mathbf{E}_{4}\mathbf{E}_{3}\mathbf{E}_{2}\mathbf{E}_{1})$$

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

$$= \begin{pmatrix} \frac{1}{3} & \frac{4}{3} \\ \frac{-1}{2} & \frac{2}{3} \end{pmatrix} \quad (1.1.2.7)$$

Now, performing elementary operations on the right side of A we obtain matrix B given as,

$$\mathbf{B} = \mathbf{AP} \tag{1.1.2.8}$$

$$\begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & -1 \\ 2 & 1 \end{pmatrix} \mathbf{P} \tag{1.1.2.9}$$

where, **P** is product of elementary matrices given by,

$$\mathbf{P} = (\mathbf{E}_{1}\mathbf{E}_{2}\mathbf{E}_{3}\mathbf{E}_{4}\mathbf{E}_{5})$$

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

$$= \begin{pmatrix} \frac{4}{3} & \frac{2}{3} \\ \frac{-5}{3} & \frac{-1}{3} \end{pmatrix} \quad (1.1.2.10)$$

Similarly, \mathbf{A} can be obtained from matrix \mathbf{B} from (1.1.2.5) as,

$$\mathbf{A} = \mathbf{C}^{-1}\mathbf{B} \tag{1.1.2.11}$$

$$\begin{pmatrix} 1 & -1 \\ 2 & 1 \end{pmatrix} = \mathbf{C}^{-1} \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \tag{1.1.2.12}$$

Matrix C is product of elementary matrices and

hence invertible and is given as,

$$\mathbf{C}^{-1} = \begin{pmatrix} \mathbf{E_1}^{-1} \mathbf{E_2}^{-1} \mathbf{E_3}^{-1} \mathbf{E_4}^{-1} \mathbf{E_5}^{-1} \mathbf{E_6}^{-1} \mathbf{E_7}^{-1} \end{pmatrix}$$

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

$$\begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & \frac{3}{2} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ -\frac{1}{3} & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & -2 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} \quad (1.1.2.13)$$

Matrix A can also be obtained from (1.1.2.8) given as,

$$\mathbf{A} = \mathbf{B}\mathbf{P}^{-1} \tag{1.1.2.14}$$

$$\begin{pmatrix} 1 & -1 \\ 2 & 1 \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \mathbf{P}^{-1} \tag{1.1.2.15}$$

where,

$$\mathbf{P}^{-1} = \left(\mathbf{E_5}^{-1} \mathbf{E_4}^{-1} \mathbf{E_3}^{-1} \mathbf{E_2}^{-1} \mathbf{E_1}^{-1}\right)$$

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

$$= \begin{pmatrix} \frac{-1}{2} & -1 \\ \frac{5}{2} & 2 \end{pmatrix} \quad (1.1.2.16)$$

Thus (1.1.2.4) can be obtained from (1.1.2.2) by multiplying it with matrix \mathbb{C} , and by inverse row operations (1.1.2.2) can be obtained back from (1.1.2.4) since \mathbb{C} is product of elementary matrices and hence invertible.

Thus the two given homogeneous systems are row equivalent.

Now writing equations in matrix-vector form as.

$$3x_1 + x_2 = \begin{pmatrix} 3 & 1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.17)

$$\Rightarrow \begin{pmatrix} 3 & 1 \end{pmatrix} \mathbf{x} = \frac{1}{3} \begin{pmatrix} 1 & -1 \end{pmatrix} \mathbf{x} + \frac{4}{3} \begin{pmatrix} 2 & 1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.18)

$$x_1 + x_2 = \begin{pmatrix} 1 & 1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.19)

$$\implies \begin{pmatrix} 1 & 1 \end{pmatrix} \mathbf{x} = \frac{-1}{3} \begin{pmatrix} 1 & -1 \end{pmatrix} \mathbf{x} + \frac{2}{3} \begin{pmatrix} 2 & 1 \end{pmatrix} \mathbf{x}$$

(1.1.2.18), (1.1.2.20) is same as multiplying C with A as it takes the linear combination of each rows of matrix A i.e, (1.1.2.6)

$$x_{1} - x_{2} = \begin{pmatrix} 1 & -1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.21)

$$\Rightarrow \begin{pmatrix} 1 & -1 \end{pmatrix} \mathbf{x} = (1) \begin{pmatrix} 3 & 1 \end{pmatrix} \mathbf{x} + (-2) \begin{pmatrix} 1 & 1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.22)

$$2x_{1} + x_{2} = \begin{pmatrix} 2 & 1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.23)

$$\Rightarrow \begin{pmatrix} 2 & 1 \end{pmatrix} \mathbf{x} = \frac{1}{2} \begin{pmatrix} 3 & 1 \end{pmatrix} \mathbf{x} + \frac{1}{2} \begin{pmatrix} 1 & 1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.24)

(1.1.2.22), (1.1.2.24) is same as multiplying C^{-1} with **B** as it takes the linear combination of each rows of matrix **B** i.e, (1.1.2.12)

Thus each equation in each system can be expressed as a linear combination of the equations in the other system when they are equivalent.

1.1.3. Are the following two systems of linear equations equivalent?

$$-x_1 + x_2 + 4x_3 = 0$$

$$x_1 + 3x_2 + 8x_3 = 0$$

$$\frac{1}{2}x_1 + x_2 + \frac{5}{2}x_3 = 0$$
(1.1.3.1)

Solution:

$$\begin{aligned}
 x_1 - x_3 &= 0 \\
 x_2 + 3x_3 &= 0
 \end{aligned}
 \tag{1.1.3.2}$$

System of linear equations in (1.1.3.1) can be expressed in matrix form as,

$$\mathbf{A}\mathbf{x} = 0 \tag{1.1.3.3}$$

$$\begin{pmatrix} -1 & 1 & 4 \\ 1 & 3 & 8 \\ \frac{1}{2} & 1 & \frac{5}{2} \end{pmatrix} \mathbf{x} = 0 \tag{1.1.3.4}$$

System of linear equations in (1.1.3.2) can be expressed in matrix form as,

$$\mathbf{B}\mathbf{x} = 0 \tag{1.1.3.5}$$

$$\begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 3 \end{pmatrix} \mathbf{x} = 0 \tag{1.1.3.6}$$

Two system of linear equations are equivalent if one system can be expressed as a linear combination of other system. Matrix **B** can be obtained from matrix **A** as,

$$\mathbf{B} = \mathbf{C}\mathbf{A} \tag{1.1.3.7}$$

$$\begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 3 \end{pmatrix} = \mathbf{C} \begin{pmatrix} -1 & 1 & 4 \\ 1 & 3 & 8 \\ \frac{1}{2} & 1 & \frac{5}{2} \end{pmatrix}$$
 (1.1.3.8)

$$\mathbf{C} = \begin{pmatrix} -1 & 1 & -2\\ \frac{1}{2} & -\frac{1}{2} & 2 \end{pmatrix} \tag{1.1.3.9}$$

Now, writing equations in matrix-vector form,

$$x_1 - x_3 = \begin{pmatrix} 1 & 0 & -1 \end{pmatrix} \mathbf{x}$$

$$\implies (1 \quad 0 \quad -1)\mathbf{x} = -1(-1 \quad 1 \quad 4)\mathbf{x} + 1(1 \quad 3 \quad 8)\mathbf{x} - 2(\frac{1}{2} \quad 1 \quad \frac{5}{2})\mathbf{x} \quad (1.1.3.10)$$

$$x_2 + 3x_3 = \begin{pmatrix} 0 & 1 & 3 \end{pmatrix} \mathbf{x}$$

$$\implies \begin{pmatrix} 0 & 1 & 3 \end{pmatrix} \mathbf{x} = \frac{1}{2} \begin{pmatrix} -1 & 1 & 4 \end{pmatrix} \mathbf{x} \\ -\frac{1}{2} \begin{pmatrix} 1 & 3 & 8 \end{pmatrix} \mathbf{x} + 2 \begin{pmatrix} \frac{1}{2} & 1 & \frac{5}{2} \end{pmatrix} \mathbf{x} \quad (1.1.3.11)$$

Equations (1.1.3.10) and (1.1.3.11) is same as multiplying C with A which is the linear combination of rows of matrix A.

Thus each equation in second system can be expressed as linear combination of the equations in first system.

Therefore, the two system of linear equations are equivalent.

1.1.4. Let \mathbb{F} be the field of complex numbers. Are the following two systems of linear equations equivalent? If so, express each equation in each system as a linear combination of equations in other system. First system of equations:

$$2x_1 + (-1+i)x_2 + x_4 = 0 (1.1.4.1)$$

$$3x_2 - 2ix_3 + 5x_4 = 0 (1.1.4.2)$$

The second system of equations:

$$(1 + \frac{i}{2})x_1 + 8x_2 - ix_3 - x_4 = 0 (1.1.4.3)$$
$$\frac{2}{3}x_1 - \frac{1}{2}x_2 + x_3 + 7x_4 = 0 (1.1.4.4)$$

Solution: Let R_1 and R_2 be the reduced row 1.1.5. Let $\mathbb F$ be a set which contains exactly two echelon forms of the augumented matrices of the following systems of homogeneous equa-

tions respectively.

$$\mathbf{AX} = \mathbf{0} \tag{1.1.4.5}$$

$$\mathbf{BX} = \mathbf{0} \tag{1.1.4.6}$$

Where A and B as follows

$$\mathbf{A} = \begin{pmatrix} 2 & -1+i & 0 & 1 \\ 0 & 3 & -2i & 5 \end{pmatrix} \tag{1.1.4.7}$$

$$\mathbf{B} = \begin{pmatrix} 1 + \frac{i}{2} & 8 & -i & -1 \\ \frac{2}{3} & -\frac{1}{2} & 1 & 7 \end{pmatrix}$$
 (1.1.4.8)

On performing elementary row operations on (1.1.4.7),

$$\mathbf{R_1} = \mathbf{CA}$$
 (1.1.4.9)

where C is the product of all elementary matrices. Reducing the first system of linear equations, we get,

$$\mathbf{C} = \begin{pmatrix} 1 & \frac{1-i}{2} \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & \frac{1}{3} \end{pmatrix} \begin{pmatrix} \frac{1}{2} & 0 \\ 0 & 1 \end{pmatrix}$$
 (1.1.4.10)

$$\mathbf{R_1} = \begin{pmatrix} 1 & 0 & \frac{-1-i}{3} & \frac{4}{3} - \frac{5i}{6} \\ 0 & 1 & \frac{-2i}{3} & \frac{5}{3} \end{pmatrix}$$
 (1.1.4.11)

On performing elementary row operations on (1.1.4.8),

$$\mathbf{R}_2 = \mathbf{D}\mathbf{A} \tag{1.1.4.12}$$

where **D** is the product of all elementary matrices. Reducing the second system of linear equations, we get,

$$\mathbf{D} = \begin{pmatrix} \frac{4}{5}(1 - \frac{i}{2}) & 0\\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0\\ \frac{-2}{3} & 1 \end{pmatrix} \begin{pmatrix} 1 & 0\\ 0 & \frac{-6(143 + 43i)}{4909} \end{pmatrix} \begin{pmatrix} 1 & \frac{16(-2+i)}{5}\\ 0 & 1 \end{pmatrix}$$
(1.1.4.13)

$$\mathbf{R_2} = \begin{pmatrix} 1 & 0 & \frac{6702}{4909} - \frac{708i}{4909} & \frac{46620}{4909} - \frac{1998i}{4909} \\ 0 & 1 & \frac{-2(441 + 472i)}{4909} & \frac{-2(3283 + 1332i)}{4909} \end{pmatrix}$$
(1.1.4.14)

From the equations (1.1.4.11) and (1.1.4.14), we can say that

$$\mathbf{R}_1 \neq \mathbf{R}_2$$
 (1.1.4.15)

Hence the given systems of linear equations are not equivalent.

elements,0 and 1.Define an addition and multiplication by tables. Verify that the set \mathbb{F} , together with these two operations, is a field.

+	0	1
0	0	1
1	1	0

•	0	1
0	0	0
1	0	1

Solution:

To prove that $(\mathbb{F},+,\cdot)$ is a field we need to satisfy the following,

- a) + and \cdot should be closed
 - For any a and b in \mathbb{F} , $a+b \in \mathbb{F}$ and $a \cdot b \in \mathbb{F}$. For example 0+0=0 and $0\cdot 0=0$.
- b) + and \cdot should be commutative
 - For any a and b in F, a+b = b+a and a ·
 b = b · a. For example 0+1=1+0 and 0 ·
 1=1 · 0.
- c) + and \cdot should be associative
 - For any a and b in \mathbb{F} , a+(b+c)=(a+b)+c and $a\cdot (b\cdot c)=(a\cdot b)\cdot c$. For example 0+(1+0)=(0+1)+0 and $0\cdot (1\cdot 0)=(0\cdot 1)\cdot 0$.
- d) + and · operations should have an identity element
 - If we perform a + 0 then for any value of a from F the result will be a itself. Hence 0 is an identity element of + operation. If we perform a ⋅ 1 then for any value of a from F the result will be a itself. Hence 1 is an identity element of ⋅ operation.
- e) \forall a \in \mathbb{F} there exists an additive inverse
 - For additive inverse to exist, ∀ a in F a+(-a)=0. For example. 1-1=0 and 0-0=0.
- f) \forall a \in \mathbb{F} such that a is non zero there exists a multiplicative inverse
 - For multiplicative inverse to exist, \forall a such that a is non zero in \mathbb{F} , $a \cdot a^{-1} = 1$. For example $1 \cdot 1^{-1} = 1$.
- g) + and \cdot should hold distributive property
 - For any a,b and c in \mathbb{F} the property $a \cdot (b+c) = a \cdot b + a \cdot c$ should always hold true. For example $0 \cdot (1+1) = 0 \cdot 1 + 0 \cdot 1$.

Since the above properties are satisfied we can say that $(\mathbb{F},+,\cdot)$ is a field.

1.1.6. Prove that if two homogenous systems of linear equations in two unknowns have the same solutions, then they are equivalent.

Solution: Let the two systems of homogenous equations be

$$\mathbf{A}\mathbf{x} = \mathbf{0} \tag{1.1.6.1}$$

$$\mathbf{B}\mathbf{y} = \mathbf{0}$$
 (1.1.6.2)

We can write

$$CAx = 0$$
 (1.1.6.3)

$$DBy = 0$$
 (1.1.6.4)

where C and D are product of elementary matrices that reduce A and B into their reduced row echelon forms R_1 and R_2

(1.1.6.3) and (1.1.6.4) imply

$$\mathbf{R_1} \mathbf{x} = 0 \tag{1.1.6.5}$$

$$\mathbf{R}_2 \mathbf{v} = 0 \tag{1.1.6.6}$$

Given that they have same solution, we can write

$$\mathbf{R_1} \mathbf{x} = 0 \tag{1.1.6.7}$$

$$\mathbf{R_2} \mathbf{x} = 0 \tag{1.1.6.8}$$

$$\implies (\mathbf{R_1} - \mathbf{R_2})\mathbf{x} = 0 \tag{1.1.6.9}$$

Note that for a solution to exist, R_1 and R_2 can be either of matrices

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \text{ or } \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{1.1.6.10}$$

Case 1 Let us assume that the solution is unique. The unique solution is

$$\mathbf{x} = \mathbf{0} \tag{1.1.6.11}$$

Since they have the same solution, both R_1 , R_2 must have their rank as 2. So,

$$\mathbf{R_1} = \mathbf{R_2} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{1.1.6.12}$$

Case 2 Let us assume that (1.1.6.3),(1.1.6.4) have infinitely many solutions So,

$$rank(A) = rank(B) = 1$$
 (1.1.6.13)

equation (1.1.6.9) for solutions other than zero solution implies

$$\mathbf{R_1} = \mathbf{R_2} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{1.1.6.14}$$

So, in both the cases, we have

$$\mathbf{R_1} = \mathbf{R_2} \tag{1.1.6.15}$$

$$\implies$$
 CA = **DB** (1.1.6.16)

Since C, D are product of elementary matrices, they are invertible.

$$\implies \mathbf{A} = \mathbf{C}^{-1}\mathbf{DB} \tag{1.1.6.17}$$

$$\mathbf{B} = \mathbf{D}^{-1}\mathbf{C}\mathbf{A} \tag{1.1.6.18}$$

Let
$$\mathbf{C}^{-1}\mathbf{D} = \mathbf{E}$$
 (1.1.6.19)

where E is also a product of elementary matrices

(1.1.6.17) and (1.1.6.18) hence become

$$\mathbf{A} = \mathbf{EB} \tag{1.1.6.20}$$

$$\mathbf{B} = \mathbf{E}^{-1}\mathbf{A} \tag{1.1.6.21}$$

Hence the two systems of equations are equivalent.

1.1.7. Prove that each subfield of the field of complex number contains every rational number

Solution:

Complex Numbers: A complex number is a number that can be expressed in the form a + bi, where a and b are real numbers, and i represents the imaginary unit, satisfying the equation $i^2 = -1$. The set of complex numbers is denoted by C

$$\mathbb{C} = \{ (a, b) : a, b \in \mathbb{R} \}$$
 (1.1.7.1)

Rational Numbers: A number in the form $\frac{p}{a}$, where both p and q(non-zero) are integers, is called a rational number. The set of rational numbers is dentoed by Q Let Q be the set of rational numbers.

$$\mathbb{Q} = \left\{ \frac{p}{q} : p \in \mathbb{Z}, q \in \mathbb{Z}_{\neq 0} \right\}$$
 (1.1.7.2)

Let \mathbb{C} be the field of complex numbers and given F be the subfield of field of complex numbers $\mathbb C$ Since $\mathbb F$ is the subfield , we could say that

$$0 \in \mathbb{F} \tag{1.1.7.3}$$

$$1 \in \mathbb{F} \tag{1.1.7.4}$$

Closed under addition: Here \mathbb{F} is closed under

addition since it is subfield

$$1 + 1 = 2 \in \mathbb{F}$$
 (1.1.7.5)

$$1 + 1 + 1 = 3 \in \mathbb{F}$$
 (1.1.7.6)

$$1 + 1 + \dots + 1$$
(p times) = $p \in \mathbb{F}$ (1.1.7.7)

$$1 + 1 + \dots + 1$$
(q times) = $q \in \mathbb{F}$ (1.1.7.8)

By using the above property we could say that zero and other positive integers belongs to \mathbb{F} . Since p and q are integers we say,

$$p \in \mathbb{Z} \tag{1.1.7.9}$$

$$q \in \mathbb{Z} \tag{1.1.7.10}$$

Additive Inverse: Let x be the positive integer belong \mathbb{F} and by additive inverse we could say,

$$\forall x \in \mathbb{F} \tag{1.1.7.11}$$

$$(-x) \in \mathbb{F} \tag{1.1.7.12}$$

Therefore field F contains every integers. Let n be a integer then,

$$n \in \mathbb{Z} \implies n \in \mathbb{F}$$
 (1.1.7.13)

$$\mathbb{Z} \subseteq \mathbb{F} \tag{1.1.7.14}$$

Where \mathbb{Z} is subset of \mathbb{F} Multiplicative Inverse: Every element except zero in the subfield \mathbb{F} has an multiplicative inverse. From equation (1.1.7.8), since $q \in \mathbb{F}$ we could say,

$$\frac{1}{q} \in \mathbb{F} \quad \text{and } q \neq 0 \tag{1.1.7.15}$$

Closed under multiplication: Also, F is closed under multiplication and thus, from equation (1.1.7.7) and (1.1.7.15) we get,

$$p \cdot \frac{1}{q} \in \mathbb{F} \tag{1.1.7.16}$$

$$p \cdot \frac{1}{q} \in \mathbb{F}$$
 (1.1.7.16)

$$\implies \frac{p}{q} \in \mathbb{F}$$
 (1.1.7.17)

where , $p \in \mathbb{Z}$ and $q \in \mathbb{Z}_{\neq 0}$ (from equation (1.1.7.3) and (1.1.7.15)) Conclusion From (1.1.7.2) and (1.1.7.17) we could say,

$$\mathbb{Q} \subseteq \mathbb{F} \tag{1.1.7.18}$$

From equation (1.1.7.18) we could say that each subfield of the field of complex number contains every rational number

Hence Proved

1.1.8. Prove that, each field of the characteristic zero contains a copy of the rational number field.

Solution: The characteristic of a field is defined to be the smallest number of times one must use the field's multiplicative identity (1) in a sum to get the additive identity. If this sum never reaches the additive identity (0), then the field is said to have characteristic zero.

Let \mathbb{Q} be the rational number field. Hence,

$$0 \in \mathbb{Q}$$
 [Additive Identity] (1.1.8.1)

$$1 \in \mathbb{Q}$$
 [Multiplicative Identity] $(1.1.8.2)$ 1.2.2. If

As addition is defined on $\mathbb Q$ hence we have,

$$1 \neq 0$$
 (1.1.8.3)

$$1 + 1 = 2 \neq 0 \tag{1.1.8.4}$$

And so on,

$$1 + 1 + \dots + 1 = n \neq 0 \tag{1.1.8.5}$$

From the definition of characteristic of a field and from (1.1.8.3), (1.1.8.4) and so on upto (1.1.8.5), the rational number field, \mathbb{Q} has characteristic 0.

$$\left(1 \quad \frac{1-i}{2}\right)\mathbf{x} = 0 \tag{1.2.1.6}$$

$$\left(1 \quad \frac{1-i}{2}\right) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0 \tag{1.2.1.7}$$

$$x_1 = -\frac{1-i}{2}x_2 \tag{1.2.1.8}$$

$$\mathbf{x} = \begin{pmatrix} -\frac{1-i}{2}x_2 \\ x_2 \end{pmatrix} \tag{1.2.1.9}$$

$$\implies \mathbf{x} = x_2 \begin{pmatrix} -\frac{1-i}{2} \\ 1 \end{pmatrix} \tag{1.2.1.10}$$

$$A = \begin{pmatrix} 3 & -1 & 2 \\ 2 & 1 & 1 \\ 1 & -3 & 0 \end{pmatrix} \tag{1.2.2.1}$$

Find all solutions of AX = 0 by row reducing A.

Solution: For the given equation AX = 0 can be defined as follows:

$$\begin{pmatrix} 3 & -1 & 2 \\ 2 & 1 & 1 \\ 1 & -3 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$
 (1.2.2.2)

Now, we can apply Row Reduction Methodol-

1.2 Matrices and Elementary Row Operations

1.2.1. Find all solutions to the system of equations

$$(1-i) x_1 - ix_2 = 0$$

2x₁ + (1-i) x₂ = 0 (1.2.1.1)

Solution: System of Linear Equations (1.2.1.1) can be expressed in matrix form as,

$$\mathbf{A}\mathbf{x} = 0 \tag{1.2.1.2}$$

$$\begin{pmatrix} 1 - i & -i \\ 2 & 1 - i \end{pmatrix} \mathbf{x} = 0 \tag{1.2.1.3}$$

By row reduction,

$$\begin{pmatrix} 1-i & -i \\ 2 & 1-i \end{pmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{pmatrix} 1 & \frac{1-i}{2} \\ 1-i & -i \end{pmatrix} \quad (1.2.1.4)$$

$$\stackrel{R_2 \leftarrow R_2 - (1-i)R_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{1-i}{2} \\ 0 & 0 \end{pmatrix} \quad (1.2.1.5)$$

ogy of matrix A:

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

$$(1.2.2.3)$$

$$\xrightarrow{R_2 = R_2 - 2R_3}
\begin{pmatrix}
5 & 0 & 3 & 0 \\
0 & 7 & 1 & 0 \\
1 & -3 & 0 & 0
\end{pmatrix}$$

$$(1.2.2.4)$$

$$\xrightarrow{R_3 = R_3 - \frac{1}{5}R_1}
\begin{pmatrix}
5 & 0 & 3 & 0 \\
0 & 7 & 1 & 0 \\
0 & -3 & -\frac{3}{5} & 0
\end{pmatrix}$$

$$(1.2.2.5)$$

$$\xrightarrow{R_1 = \frac{1}{5}R_1}
\begin{pmatrix}
1 & 0 & \frac{3}{5} & 0 \\
0 & 7 & 1 & 0 \\
0 & -3 & -\frac{3}{5} & 0
\end{pmatrix}$$

$$(1.2.2.5)$$

$$\xrightarrow{R_2 = \frac{1}{7}R_2}
\begin{pmatrix}
1 & 0 & \frac{3}{5} & 0 \\
0 & 1 & \frac{1}{7} & 0 \\
0 & -3 & -\frac{3}{5} & 0
\end{pmatrix}$$

$$(1.2.2.7)$$

$$\xrightarrow{R_3 = R_3 + 3R_2}
\begin{pmatrix}
1 & 0 & \frac{3}{5} & 0 \\
0 & 1 & \frac{1}{7} & 0 \\
0 & 0 & -\frac{6}{35} & 0
\end{pmatrix}$$

$$(1.2.2.8)$$

$$\xrightarrow{R_3 = -\frac{35}{6}R_3}
\begin{pmatrix}
1 & 0 & \frac{3}{5} & 0 \\
0 & 1 & \frac{1}{7} & 0 \\
0 & 0 & 1 & 0
\end{pmatrix}$$

$$(1.2.2.9)$$

$$\xrightarrow{R_2 = R_2 - \frac{1}{7}R_3}
\begin{pmatrix}
1 & 0 & \frac{3}{5} & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0
\end{pmatrix}$$

$$(1.2.2.10)$$

$$\xrightarrow{R_1 = R_1 - \frac{3}{5}R_3}
\begin{pmatrix}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0
\end{pmatrix}$$

So, as we can see the only solution we got after row reducing of matrix *A* is zero vector. Thus, the solution is:

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$
 (1.2.2.12)

(1.2.2.11)

1.2.3.

$$\mathbf{A} = \begin{pmatrix} 6 & -4 & 0 \\ 4 & -2 & 0 \\ -1 & 0 & 3 \end{pmatrix} \tag{1.2.3.1}$$

Find all solutions of AX = 2X and all solutions of AX = 3X. The symbol cX denotes the matrix each entry of which is c times corresponding entry.

Solution:

$$\mathbf{A} = \begin{pmatrix} 6 & -4 & 0 \\ 4 & -2 & 0 \\ -1 & 0 & 3 \end{pmatrix} \tag{1.2.3.2}$$

To calculate solution of AX = 2X and all solutions of AX = 3X we calculate eigen values of A:

$$(\mathbf{A} - \lambda \mathbf{I})\mathbf{X} = 0 \tag{1.2.3.3}$$

Substituting values in (1.2.3.3),

$$\begin{pmatrix} 6 - \lambda & -4 & 0 \\ 4 & -2 - \lambda & 0 \\ -1 & 0 & 3 - \lambda \end{pmatrix} \mathbf{X} = 0 \quad (1.2.3.4)$$

Simplifying:

$$\begin{pmatrix} 6 - \lambda & -4 & 0 \\ 4 & -2 - \lambda & 0 \\ -1 & 0 & 3 - \lambda \end{pmatrix} \xrightarrow{R_1 \leftarrow R_1 - R_2}$$

$$\begin{pmatrix} 2 - \lambda & -2 + \lambda & 0 \\ 4 & -2 - \lambda & 0 \\ -1 & 0 & 3 - \lambda \end{pmatrix}$$
 (1.2.3.5)

Taking $(3-\lambda)$ and $(2-\lambda)$ common from C_3 and R_1

$$(3-\lambda)(2-\lambda)\begin{pmatrix} 1 & -1 & 0\\ 4 & -2-\lambda & 0\\ -1 & 0 & 1 \end{pmatrix}$$
 (1.2.3.6)

$$\begin{pmatrix} 1 & -1 & 0 \\ 4 & -2 - \lambda & 0 \\ -1 & 0 & 1 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_2 - 4R_1} \begin{pmatrix} 1 & -1 & 0 \\ 0 & -\lambda + 2 & 0 \\ -1 & 0 & 1 \end{pmatrix}$$
(1.2.3.7)

Taking $(2 - \lambda)$ common from R_2 :

$$(2-\lambda)^2(3-\lambda)\begin{pmatrix} 1 & -1 & 0\\ 0 & 1 & 0\\ -1 & 0 & 1 \end{pmatrix}$$
 (1.2.3.8)

Eigen values are:

$$\lambda_1 = 2 \tag{1.2.3.9}$$

$$\lambda_2 = 3$$
 (1.2.3.10)

solution to AX = 2X is eigen vector corresponding to $\lambda = 2$

$$(\mathbf{A} - 2\mathbf{I})\mathbf{X} = 0$$
 (1.2.3.11)

Substituting values:

$$\begin{pmatrix} 4 & -4 & 0 & 0 \\ 4 & -4 & 0 & 0 \\ -1 & 0 & 1 & 0 \end{pmatrix} \xleftarrow{R_1 \leftarrow \frac{R_1}{4}} \begin{pmatrix} 1 & -1 & 0 & 0 \\ 4 & -4 & 0 & 0 \\ -1 & 0 & 1 & 0 \end{pmatrix} \xleftarrow{R_2 \leftarrow R_2 - 4R_1}$$

$$\begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 \end{pmatrix} \xleftarrow{R_3 \leftarrow R_3 - R_1} \begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 \end{pmatrix} \xleftarrow{R_3 \longleftrightarrow R_2}$$

$$\begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_2 \leftarrow -R_2} \begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_1 \leftarrow R_1 + R_2} \stackrel{1.2}{\longleftrightarrow}$$

$$\begin{pmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \tag{1.2.3.12}$$

So, x_3 is a free variable: Let $x_3 = c$.

$$x_2 - x_3 = 0 \implies x_2 = x_3 = c$$
 (1.2.3.13)

$$x_1 - x_3 = 0 \implies x_1 = x_3 = c$$
 (1.2.3.14)

So, the solution to AX = 2Xis

$$\mathbf{X} = c \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \tag{1.2.3.15}$$

solution of $\mathbf{AX} = 3\mathbf{X}$ is eigen vector corresponding to $\lambda = 3$

$$(\mathbf{A} - 3\mathbf{I})\mathbf{X} = 0$$
 (1.2.3.16)

substituting we have:

$$\begin{pmatrix} 3 & -4 & 0 & 0 \\ 1 & -2 & 0 & 0 \\ -1 & 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_1 \leftarrow \frac{R_1}{3}} \begin{pmatrix} 1 & -\frac{4}{3} & 0 & 0 \\ 4 & -5 & 0 & 0 \\ -1 & 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_2 - 4R_1}$$

$$\begin{pmatrix} 1 & -\frac{4}{3} & 0 & 0 \\ 0 & \frac{1}{3} & 0 & 0 \\ -1 & 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_3 \leftarrow R_3 + R_1} \begin{pmatrix} 1 & -\frac{4}{3} & 0 & 0 \\ 0 & \frac{1}{3} & 0 & 0 \\ 0 & -\frac{4}{3} & 0 & 0 \end{pmatrix} \xrightarrow{R_2 \leftarrow \frac{R_2}{3}}$$

$$\begin{pmatrix} 1 & \frac{-4}{3} & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -\frac{4}{3} & 0 & 0 \end{pmatrix} \longleftrightarrow \xrightarrow{R_3 \leftarrow R_3 - \frac{4}{3}R_2} \begin{pmatrix} 1 & \frac{4}{3} & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \longleftrightarrow \xrightarrow{R_3 \leftarrow R_1 + \frac{4}{3}R_2}$$

$$\begin{pmatrix}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$
(1.2.3.17)

So x_3 is a free variable:

$$x_1 = 0 \tag{1.2.3.18}$$

$$x_2 = 0 (1.2.3.19)$$

$$x_3 = c$$
 (1.2.3.20)

So, the solution to AX = 3X is,

$$\mathbf{X} = c \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \tag{1.2.3.21}$$

1.2.4. Find a row-reduced matrix which is row equivalent to,

$$\mathbf{A} = \begin{pmatrix} i & -(1+i) & 0 \\ 1 & -2 & 1 \\ 1 & 2i & -1 \end{pmatrix}$$
 (1.2.4.1)

Solution: Step 1: Performing scaling operation to matrix **A** as $R_1 \leftarrow \frac{1}{i}R_1$ by scaling matrix D_1 given as,

$$\mathbf{D_1} = \begin{pmatrix} \frac{1}{i} & 0 & 0\\ 0 & 1 & 0\\ 0 & 0 & 1 \end{pmatrix} \ (1.2.4.2)$$

$$\mathbf{D_1A} = \begin{pmatrix} \frac{1}{i} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} i & -(1+i) & 0 \\ 1 & -2 & 1 \\ 1 & 2i & -1 \end{pmatrix} (1.2.4.3)$$

$$\implies \mathbf{D_1 A} = \begin{pmatrix} 1 & -1 + i & 0 \\ 1 & -2 & 1 \\ 1 & 2i & -1 \end{pmatrix} (1.2.4.4)$$

Step 2: Performing $R_2 \leftarrow R_2 - R_1$ and $R_3 \leftarrow R_3 - R_1$ given by elementary matrix $\mathbf{E_{31}E_{21}}$ on

equation (1.2.4.4),

$$\mathbf{E_{31}E_{21}} = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix}$$
(1.2.4.5)

$$\mathbf{E_{31}E_{21}D_{1}A} = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1+i & 0 \\ 1 & -2 & 1 \\ 1 & 2i & -1 \end{pmatrix}$$
(1.2.4.6)

$$\implies \mathbf{A_1} = \mathbf{E_{31}} \mathbf{E_{21}} \mathbf{D_1} \mathbf{A} = \begin{pmatrix} 1 & -1+i & 0 \\ 0 & -1-i & 1 \\ 0 & 1+i & -1 \end{pmatrix}$$
(1.2.4.7)

Step 3: Performing $R_2 \leftarrow \frac{-1}{1+i}R_2$ given by $\mathbf{D_2}$ on equation (1.2.4.7),

$$\mathbf{D_2} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2}(-1+i) & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
(1.2.4.8)

$$\mathbf{D_2A_1} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2}(-1+i) & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1+i & 0 \\ 0 & -1-i & 1 \\ 0 & 1+i & -1 \end{pmatrix}$$
(1.2.4.9)

$$\implies \mathbf{A_2} = \mathbf{D_2} \mathbf{A_1} = \begin{pmatrix} 1 & -1+i & 0\\ 0 & 1 & \frac{1}{2}(-1+i)\\ 0 & 1+i & -1 \end{pmatrix}$$
(1.2.4.10)

Step 4: Performing $R_3 \leftarrow R_3 - (1+i)R_2$ given by E_{32} on equation (1.2.4.10),

$$\mathbf{E_{32}} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -(1+i) & 1 \end{pmatrix} \tag{1.2.4.11}$$

$$\mathbf{E}_{32}\mathbf{A}_{2} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 - i & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 + i & 0 \\ 0 & 1 & \frac{-1 + i}{2} \\ 0 & 1 + i & -1 \end{pmatrix}$$
(1.2.4.12)

$$\implies \mathbf{A_3} = \mathbf{E_{32}A_2} = \begin{pmatrix} 1 & -1+i & 0\\ 0 & 1 & \frac{-1+i}{2}\\ 0 & 0 & 1 \end{pmatrix}$$
(1.2.4.13)

Step 5: Performing $R_1 \leftarrow R_1 - (-1+i)R_2$ given

by E_{12} on equation (1.2.4.13),

$$\mathbf{E_{12}} = \begin{pmatrix} 1 & 1 - i & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{1.2.4.14}$$

$$\mathbf{E}_{12}\mathbf{A}_{3} = \begin{pmatrix} 1 & 1-i & 0\\ 0 & 1 & 0\\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1+i & 0\\ 0 & 1 & \frac{-1+i}{2}\\ 0 & 0 & 1 \end{pmatrix}$$
(1.2.4.15)

$$\implies \mathbf{A_4} = \mathbf{E_{12}A_3} = \begin{pmatrix} 1 & 0 & i \\ 0 & 1 & \frac{-1+i}{2} \\ 0 & 0 & 1 \end{pmatrix} \quad (1.2.4.16)$$

Step 6: Performing $R_1 \leftarrow R_1 - iR_3$ and $R_2 \leftarrow R_2 - \frac{-1+i}{2}R_3$ given by $\mathbf{E_{13}E_{23}}$ on equation (1.2.4.16),

$$\mathbf{E_{13}E_{23}} = \begin{pmatrix} 1 & 0 & -i \\ 0 & 1 & -\left(\frac{-1+i}{2}\right) \\ 0 & 0 & 1 \end{pmatrix}$$
 (1.2.4.17)

$$\mathbf{E_{13}E_{23}A_{4}} = \begin{pmatrix} 1 & 0 & -i \\ 0 & 1 & -\left(\frac{-1+i}{2}\right) \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & i \\ 0 & 1 & \frac{-1+i}{2} \\ 0 & 0 & 1 \end{pmatrix}$$

$$(1.2.4.18)$$

$$\implies \mathbf{A_5} = \mathbf{E_{13}E_{23}A_{4}} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$(1.2.4.19)$$

 \therefore Row-reduced matrix of **A** given by equation (1.2.4.1) is,

$$\mathbf{A} = \begin{pmatrix} i & -1 - i & 0 \\ 1 & -2 & 1 \\ 1 & 2i & -1 \end{pmatrix} \longleftrightarrow \begin{pmatrix} RREF \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \mathbf{I}$$
(1.2.4.20)

1.2.5. Prove that the following two matrices are not row equivalent

$$\begin{pmatrix} 2 & 0 & 0 \\ a & -1 & 0 \\ b & c & 3 \end{pmatrix}, \begin{pmatrix} 1 & 1 & 2 \\ -2 & 0 & -1 \\ 1 & 3 & 5 \end{pmatrix}$$
 (1.2.5.1)

Solution: Call the first matrix **A** and the second matrix **B**.

$$\mathbf{A}^T = \begin{pmatrix} 2 & a & b \\ 0 & -1 & c \\ 0 & 0 & 3 \end{pmatrix} \tag{1.2.5.2}$$

 A^T is a upper triangular matrix with non-zero diagonal. Hence it has full rank = 3.

$$\mathbf{B}^{T} = \begin{pmatrix} 1 & -2 & 1 \\ 1 & 0 & 3 \\ 2 & -1 & 5 \end{pmatrix} \xrightarrow{R_{2} \to R_{2} - R_{1}} \begin{pmatrix} 1 & -2 & 1 \\ 0 & 2 & 2 \\ 0 & 3 & 3 \end{pmatrix}$$

$$(1.2.5.3)$$

$$\xrightarrow{R_{3} \leftarrow R_{3}/3} \begin{pmatrix} 1 & -2 & 1 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

$$(1.2.5.4)$$

$$\xrightarrow{R_{3} \leftarrow R_{3} - R_{2}} \begin{pmatrix} 1 & -2 & 1 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

$$(1.2.5.5)$$

 \mathbf{B}^T is a upper triangular matrix with zero diagonal. Hence it doesn't have full rank. Therefore both matrices have different rank, so it cannot be row equivalent.

1.2.6. Let

$$\mathbf{A} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \tag{1.2.6.1}$$

be a 2×2 matrix with complex entries. Suppose A is row-reduced and also that a+b+c+d=0. Prove that there are exactly three such matrices. **Solution:** A matrix is in row echelon form if it follows the following conditions

- 1. All nonzero rows are above any rows of all zeros.
- 2. Each leading entry (i.e. left most nonzero entry) of a row is in a column to the right of the leading entry of the row above it.
- 3. All entries in a column below a leading entry are zero Row Reduced Echelon Form A matrix is in row reduced echelon form if it follows the following conditions
- 1. The matrix should be row echelon form
- 2. The leading entry in each nonzero row is 1.
- 3. Each leading 1 is the only nonzero entry in its column. Proof

Given,

$$\mathbf{A} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \tag{1.2.6.2}$$

Condition 1: Matrix **A** should be in row-reduced echelon form

Condition 2 : a + b + c + d = 0 where a,b,c and d are the elements of the matrix **A** Reducing the matrix **A** from equation (1.2.6.2)

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \stackrel{R_1 = \frac{1}{a}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{b}{a} \\ c & d \end{pmatrix}$$
 (1.2.6.3)

$$\stackrel{R_2=R_2-cR_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{b}{a} \\ 0 & \frac{ad^a-bc}{a} \end{pmatrix}$$
 (1.2.6.4)

$$\stackrel{R_2 = \frac{a}{ad - bc} R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{b}{a} \\ 0 & 1 \end{pmatrix}$$
(1.2.6.5)

$$\stackrel{R_1 = R_1 - \frac{b}{a}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{1.2.6.6}$$

Case 1: Matrix A of Rank 2

From the equation (1.2.6.4), for the matrix to be in row reduced echelon form,

$$b = 0$$

$$a \neq 0$$

$$d = 1$$

$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$
(1.2.6.7)

For the condition 2 to get satisfied,

$$a + 0 + c + 1 = 0 ag{1.2.6.8}$$

$$\implies a = -(c+1) \tag{1.2.6.9}$$

$$\implies c \neq -1$$
 (1.2.6.10)

Both the condition gets satisfied and so exactly one matrix **A** can be formed of Rank 2 with given conditions

Case 2: Matrix A of Rank 1

From the equation (1.2.6.4), for the matrix to be in row reduced echelon form,

$$a \neq 0$$
$$d = 0$$
$$c = 0$$

For the condition 2 to get satisfied,

$$a + b + 0 + 0 = 0 (1.2.6.11)$$

$$\implies b = -a \tag{1.2.6.12}$$

$$\mathbf{A} = \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \tag{1.2.6.13}$$

Both the condition gets satisfied and so exactly one matrix **A** can be formed of Rank 1 with given conditions

Case 3: Matrix A of Rank 0

From equation (1.2.6.2), for the matrix to be in row reduced echelon form,

$$a = 0$$

$$b = 0$$

$$c = 0$$

$$d = 0$$

$$\mathbf{A} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$
(1.2.6.14)

Both the condition gets satisfied and so exactly one matrix **A** can be formed of Rank 0 with given conditions

Therefore matrix **A** shown in equation (1.2.6.7),(1.2.6.13) and (1.2.6.14) are the exactly three such matrices that can be formed with given conditions.

1.2.7. Prove that the interchange of two rows of a matrix can be accomplished by a finite sequence of elementary row operations of the other two types.

Solution: Let **A** be a 3×3 matrix with having row vectors $\mathbf{a}_1, \mathbf{a}_2$ and \mathbf{a}_3 .

$$\mathbf{A} = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \end{pmatrix} \tag{1.2.7.1}$$

Let's exchange row \mathbf{a}_1 and \mathbf{a}_2 . Let's call this elementary operation \mathbf{E}_1 .

$$\mathbf{E}_1 = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{1.2.7.2}$$

(1.2.7.3)

Now performing operation E_1

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \end{pmatrix} = \begin{pmatrix} \mathbf{a}_2 \\ \mathbf{a}_1 \\ \mathbf{a}_3 \end{pmatrix}$$
 (1.2.7.4)

Now, to prove that same matrix can be obtained by elementary operations let's call them $\mathbf{E_2}$ and $\mathbf{E_3}$. Now performing operation $\mathbf{E_2}$ by adding

row 2 to row 1.

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \end{pmatrix} = \begin{pmatrix} \mathbf{a}_1 + \mathbf{a}_2 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \end{pmatrix}$$
 (1.2.7.5)

Using elementary operation E_2 we will subtract row 1 from row 2.

$$\begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 + \mathbf{a}_2 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \end{pmatrix} = \begin{pmatrix} \mathbf{a}_1 + \mathbf{a}_2 \\ -\mathbf{a}_1 \\ \mathbf{a}_3 \end{pmatrix} \quad (1.2.7.6)$$

Using elementary operation E_2 we will add row 2 to row 1.

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 + \mathbf{a}_2 \\ -\mathbf{a}_1 \\ \mathbf{a}_3 \end{pmatrix} = \begin{pmatrix} \mathbf{a}_2 \\ -\mathbf{a}_1 \\ \mathbf{a}_3 \end{pmatrix} \tag{1.2.7.7}$$

Using elementary operation E_3 we will multiply row 2 by -1.

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{a}_2 \\ -\mathbf{a}_1 \\ \mathbf{a}_3 \end{pmatrix} = \begin{pmatrix} \mathbf{a}_2 \\ \mathbf{a}_1 \\ \mathbf{a}_3 \end{pmatrix} \tag{1.2.7.8}$$

Hence, we can say that,

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \end{pmatrix} =$$

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \mathbf{a}_3 \end{pmatrix} (1.2.7.9)$$

Let us assume a matrix A

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} \tag{1.2.7.10}$$

Let's exchange row \mathbf{a}_1 and \mathbf{a}_2 by applying operation \mathbf{E}_1 .

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 2 & 3 \\ 1 & 1 & 0 \end{pmatrix}$$
 (1.2.7.11)

Now, to prove that same matrix can be obtained by other two elementary operations. We will first perform elementary operation $\mathbf{E_2}$ by adding row 2 to row 1.

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 3 & 3 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} \quad (1.2.7.12)$$

Using elementary operation E_2 we will subtract row 1 from row 2.

$$\begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 3 & 3 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 3 & 3 \\ -1 & -2 & -3 \\ 1 & 1 & 0 \end{pmatrix}$$
(1.2.7.13)

Using elementary operation E_2 we will add row 2 to row 1.

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

$$(1.2.7.14)$$

Using elementary operation E_3 we will multiply row 2 by -1.

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ -1 & -2 & -3 \\ 1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 2 & 3 \\ 1 & 1 & 0 \end{pmatrix}$$
(1.2.7.15)

Hence, we can say that,

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

$$\times \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix}$$

$$(1.2.7.16)$$

1.2.8. Consider the system of equations AX = 0 where

$$\mathbf{A} = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

is a 2×2 matrix over the field F. Prove the following -

- If every entry of **A** is 0, then every pair x_1 and x_2 is a solution of $\mathbf{AX} = 0$.
- If $ad bc \neq 0$, then the system $\mathbf{AX} = 0$ has only the trivial solution $x_1 = x_2 = 0$
- If ad bc = 0 and some entry of **A** is different from 0, then there is a solution x_1^0 and x_2^0 such that x_1 and x_2 is a solution if and only if there is a scalar y such that $x_1 = yx_1^0$ and $x_2 = yx_2^0$

Solution: Solution 1 If every entry of **A** is 0

then the equation AX = 0 becomes,

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0$$
 (1.2.8.1)

$$\implies 0.x_1 + 0.x_2 = 0 \forall x_1, x_2 \in F$$
 (1.2.8.2)

Hence proved, every pair x_1 and x_2 is a solution for the equation $\mathbf{AX} = 0$. Solution 2 **Case 1:** Let a = 0. Since $ad - bc \neq 0$. As $bc \neq 0$ therefore $b \neq 0$ and $c \neq 0$. Hence, we can perform row reduction on the augmented matrix of equation $\mathbf{AX} = 0$ as follows,

Equation
$$AX = 0$$
 as follows,

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & b & 0 \\ c & d & 0 \end{pmatrix} = \begin{pmatrix} \frac{1}{c} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} c & d & 0 \\ 0 & b & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & \frac{1}{b} \end{pmatrix} \begin{pmatrix} 1 & \frac{d}{c} & 0 \\ 0 & b & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & -\frac{d}{c} \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{d}{c} & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

Case 2: Let $a, b, c, d \neq 0$. Considering the following case,

$$\mathbf{AX} = \mathbf{u}$$
 (1.2.8.7)

$$\implies \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix}$$
 (1.2.8.8)

Row Reducing the augmented matrix of (1.2.8.8) we get,

$$\begin{pmatrix} \frac{1}{a} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} a & b & u_1 \\ c & d & u_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ -c & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{b}{a} & \frac{u_1}{a} \\ c & d & u_2 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & \frac{a}{ad-bc} \end{pmatrix} \begin{pmatrix} 1 & \frac{b}{a} & \frac{u_1}{a} \\ 0 & \frac{ad-bc}{a} & \frac{au_2-cu_1}{a} \\ (1.2.8.10) \end{pmatrix}$$

$$= \begin{pmatrix} 1 & -\frac{b}{a} \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{b}{a} & \frac{u_1}{a} \\ 0 & 1 & \frac{au_2-cu_1}{ad-bc} \\ (1.2.8.11) \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 & \frac{du_1-bu_2}{ad-bc} \\ 0 & 1 & \frac{au_2-cu_1}{ad-bc} \\ \end{pmatrix}$$

$$(1.2.8.12)$$

From (1.2.8.12) we get,

$$x_1 = \frac{du_1 - bu_2}{ad - bc} \tag{1.2.8.13}$$

$$x_2 = \frac{aa - bc}{au_2 - cu_1}$$

$$x_2 = \frac{aa - bc}{ad - bc}$$
(1.2.8.14)

Since $u_1 = 0$ and $u_2 = 0$ then from (1.2.8.13) and (1.2.8.14),

$$x_1 = 0 \tag{1.2.8.15}$$

$$x_2 = 0 (1.2.8.16)$$

Hence we get,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{1.2.8.17}$$

In (1.2.8.6) and (1.2.8.17), we can see that $\mathbf{AX} = 0$ has only one trivial solution i.e $x_1 = x_2 = 0$ in all cases. Hence proved, the equation $\mathbf{AX} = 0$ has only one trivial solution $x_1 = x_2 = 0$ Solution 3 **Case 1:** Let, $a \neq 0$ for **A**. Given ad - bc = 0, we can perform row reduction on augmented matrix of equation $\mathbf{AX} = 0$ as follows,

$$\begin{pmatrix} \frac{1}{a} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} a & b & 0 \\ c & d & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ -c & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{b}{a} & 0 \\ c & d & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & \frac{b}{a} & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad [\because ad - bc = 0]$$

$$(1.2.8.19)$$

Hence from (1.2.8.19), AX = 0 if and only if

$$x_1 = -\frac{b}{a}x_2 \qquad [a \neq 0] \tag{1.2.8.20}$$

Letting $x_1^0 = -\frac{b}{a}$ and $x_2^0 = 1$ we get for y = 1,

$$x_1 = yx_1^0 (1.2.8.21)$$

$$x_2 = yx_2^0 (1.2.8.22)$$

which is a solution of the equation $\mathbf{AX} = 0$. **Case 2:** Let, $b \neq 0$ for **A**. Given ad - bc = 0, at first we multiply by elementary matrix to change the columns and the we can perform row reduction on augmented matrix of equation $\mathbf{AX} = 0$ as follows,

$$\begin{pmatrix} a & b & 0 \\ c & d & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} b & a & 0 \\ d & c & 0 \end{pmatrix}$$
 (1.2.8.23)

Hence using the result obtained from (1.2.8.19)

we can conclude for (1.2.8.23), $\mathbf{AX} = 0$ if and only if

$$x_2 = -\frac{a}{b}x_1 \qquad [b \neq 0] \tag{1.2.8.24}$$

Letting $x_2^0 = -\frac{a}{b}$ and $x_1^0 = 1$ we get for y = 1,

$$x_1 = yx_1^0 (1.2.8.25)$$

$$x_2 = yx_2^0 (1.2.8.26)$$

which is a solution of the equation $\mathbf{AX} = 0$. **Case 3:** Let, $c \neq 0$ for **A**. Given ad - bc = 0, we can perform row reduction on augmented matrix of equation $\mathbf{AX} = 0$ as follows,

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} a & b & 0 \\ c & d & 0 \end{pmatrix} = \begin{pmatrix} \frac{1}{c} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} c & d & 0 \\ a & b & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ -a & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{d}{c} & 0 \\ a & b & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & \frac{d}{c} & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad [\because ad - bc = 0]$$

$$(1.2.8.29)$$

Hence from (1.2.8.29), $\mathbf{AX} = 0$ if and only if

$$x_1 = -\frac{d}{c}x_2 \qquad [a \neq 0] \tag{1.2.8.30}$$

Letting $x_1^0 = -\frac{d}{c}$ and $x_2^0 = 1$ we get for y = 1,

$$x_1 = yx_1^0 (1.2.8.31)$$

$$x_2 = yx_2^0 (1.2.8.32)$$

which is a solution of the equation $\mathbf{AX} = 0$. **Case 4:** Let, $d \neq 0$ for **A**. Given ad - bc = 0, at first we multiply by elementary matrix to change the columns and then we can perform row reduction on augmented matrix of equation $\mathbf{AX} = 0$ as follows,

$$\begin{pmatrix} a & b & 0 \\ c & d & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} b & a & 0 \\ d & c & 0 \end{pmatrix}$$
 (1.2.8.33)
$$= \begin{pmatrix} d & c & 0 \\ b & a & 0 \end{pmatrix}$$
 (1.2.8.34)

Hence using the result from (1.2.8.29) we can conclude for (1.2.8.34), $\mathbf{AX} = 0$ if and only if

$$x_2 = -\frac{c}{d}x_1 \qquad [a \neq 0] \tag{1.2.8.35}$$

Letting $x_2^0 = -\frac{c}{d}$ and $x_1^0 = 1$ we get for y = 1,

$$x_1 = yx_1^0 \tag{1.2.8.36}$$

$$x_2 = yx_2^0 (1.2.8.37)$$

which is a solution of the equation AX = 0.

1.3 Row Reduced Echelon Matrices

1.3.1. Find all solutions to the following system of equations by row-reducing the co-efficient matrix:

$$\frac{1}{3}x_1 + 2x_2 - 6x_3 = 0 (1.3.1.1)$$

$$-4x_1 + 5x_3 = 0 (1.3.1.2)$$

$$-3x_1 + 6x_2 - 13x_3 = 0 (1.3.1.3)$$

$$-\frac{7}{3}x_1 + 2x_2 - \frac{8}{3}x_3 = 0 (1.3.1.4)$$

Solution: The coefficient matrix is:

$$A = \begin{pmatrix} \frac{1}{3} & 2 & -6 \\ -4 & 0 & 5 \\ -3 & 6 & -13 \\ -\frac{7}{2} & 2 & -\frac{8}{2} \end{pmatrix}$$
 (1.3.1.5)

The number of rows of this coefficient matrix is m = 4 and the number of columns is n = 3, So in this case, n < m. Now the row operations

are:

$$\begin{pmatrix} \frac{1}{3} & 2 & -6 \\ -4 & 0 & 5 \\ -3 & 6 & -13 \\ -\frac{7}{3} & 2 & -\frac{8}{3} \end{pmatrix} \xrightarrow{R_4 \leftarrow R_4 \times 3} \begin{pmatrix} 1 & 6 & -18 \\ -4 & 0 & 5 \\ -3 & 6 & -13 \\ -7 & 6 & -8 \end{pmatrix}$$

$$(1.3.1.6)$$

$$\stackrel{R_3 \leftarrow R_2 + R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 6 & -18 \\
-4 & 0 & 5 \\
-7 & 6 & -8 \\
-7 & 6 & -8
\end{pmatrix}
\stackrel{R_4 \leftarrow R_4 - R_3}{\longleftrightarrow}$$

(1.3.1.7)

$$\begin{pmatrix} 1 & 6 & -18 \\ -4 & 0 & 5 \\ -7 & 6 & -8 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_2 + 4R_1} \begin{pmatrix} 1 & 6 & -18 \\ 0 & 24 & -67 \\ R_3 \leftarrow R_3 + 7R_1 \end{pmatrix} \begin{pmatrix} 1 & 6 & -18 \\ 0 & 24 & -67 \\ 0 & 48 & -138 \\ 0 & 0 & 0 \end{pmatrix}$$

$$(1.3.1.8)$$

$$\stackrel{R_3 \leftarrow R_3/2}{\longleftrightarrow} \begin{pmatrix}
1 & 6 & -18 \\
0 & 24 & -67 \\
0 & 24 & -69 \\
0 & 0 & 0
\end{pmatrix}
\stackrel{R_3 \leftarrow R_3 - R_2}{\longleftrightarrow}$$

(1.3.1.9)

$$\begin{pmatrix} 1 & 6 & -18 \\ 0 & 24 & -67 \\ 0 & 0 & -2 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_3 \leftarrow \frac{R_3}{(-2)}} \begin{pmatrix} 1 & 6 & -18 \\ 0 & 24 & -67 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$
(1.3.1.10)

$$\stackrel{R_2 \leftarrow \frac{R_2}{4}}{\longleftrightarrow} \begin{pmatrix} 1 & 6 & -18 \\ 0 & 6 & -\frac{67}{4} \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \stackrel{R_1 \leftarrow R_1 - R_2}{\longleftrightarrow}$$

(1.3.1.11)

$$\begin{pmatrix}
1 & 0 & -\frac{5}{4} \\
0 & 6 & -\frac{67}{4} \\
0 & 0 & 1 \\
0 & 0 & 0
\end{pmatrix}
\xrightarrow{R_2 \leftarrow \frac{R_2}{6}}
\begin{pmatrix}
1 & 0 & -\frac{5}{4} \\
0 & 1 & -\frac{67}{24} \\
0 & 0 & 1 \\
0 & 0 & 0
\end{pmatrix}$$
(1.3.1.12)

$$\xrightarrow{R_2 \leftarrow R_2 + \frac{67R_3}{24}} \xrightarrow{R_1 \leftarrow R_1 + \frac{5R_3}{4}} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \tag{1.3.1.13}$$

Now,

$$A\mathbf{x} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = 0 \tag{1.3.1.14}$$

So,

$$\mathbf{I_3x} = 0 \tag{1.3.1.15}$$

$$\implies \mathbf{x} = 0 \tag{1.3.1.16}$$

1.3.2. Find a row-reduced matrix which is row equivalent to A.What are the solutions of $\mathbf{A}\mathbf{x} = \mathbf{0}$?

$$\mathbf{A} = \begin{pmatrix} 1 & -i \\ 2 & 2 \\ i & 1+i \end{pmatrix} \tag{1.3.2.1}$$

Solution: Let R be a row-reduced echelon matrix which is row equivalent to A. Then the systems

$$Ax = 0, Rx = 0$$
 (1.3.2.2)

have the same solutions. On performing elementary row operations on (1.3.2.1),

$$\mathbf{R} = \mathbf{B}\mathbf{A} \tag{1.3.2.3}$$

where **B** is the product of all elementary matrices. Reducing the given matrix, we get

$$\begin{aligned} \mathbf{B} &= (\mathbf{E_5} \mathbf{E_4} \mathbf{E_3} \mathbf{E_2} \mathbf{E_1}) \\ &= \begin{pmatrix} 1 & i & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -i & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{4}(1-i) & 0 \\ 0 & 0 & 1 \end{pmatrix} \\ &\qquad \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -i & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \\ &= \begin{pmatrix} \frac{1}{2}(1-i) & \frac{1}{4}(1+i) & 0 \\ \frac{1}{2}(-1+i) & \frac{1}{4}(1-i) & 0 \\ \frac{1}{2}(1-i) & \frac{1}{4}(-1-i) & 1 \end{pmatrix} \quad (1.3.2.4) \end{aligned}$$

$$\mathbf{R} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \tag{1.3.2.5}$$

:. Row-reduced matrix of A is,

$$\mathbf{A} = \begin{pmatrix} 1 & -i \\ 2 & 2 \\ i & 1+i \end{pmatrix} \xrightarrow{RREF} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \tag{1.3.2.6}$$

From(1.3.2.2) and (1.3.2.6),

$$A\mathbf{x} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0 \tag{1.3.2.7}$$

The solution of Ax = 0 is,

$$I_2 \mathbf{x} = 0 \tag{1.3.2.8}$$

$$\implies \mathbf{x} = 0 \tag{1.3.2.9}$$

As I_2 is invertible.

alent to A.What are the solutions of $\mathbf{A}\mathbf{x} = \mathbf{0}$? 1.3.3. Describe explicitly all 2x2 row-reduced echelon matrices.

Solution:

2x2 matrices which are row-reduced echelon matrix can be represented as a linear combination of three matrices:-

$$c_1 \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + c_2 \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c_3 \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
 (1.3.3.1)

1.3.4. Consider the system of the equations

$$x_1 - x_2 + 2x_3 = 1 (1.3.4.1)$$

$$x_1 - 0x_2 + 2x_3 = 1 (1.3.4.2)$$

$$x_1 - 3x_2 + 4x_3 = 2 ag{1.3.4.3}$$

Does this system have a solution? If so describe explicitly all solutions.

Solution: Let **V** is the set of all $(x_1, x_2, x_3) \in \mathbb{R}^3$ which satisfy the (1.3.4.1), (1.3.4.2) and (1.3.4.3)

From equation (1.3.4.1) to (1.3.4.3) we can write,

$$\begin{pmatrix} 1 & -1 & 2 \\ 1 & 0 & 2 \\ 1 & -3 & 4 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} (1.3.4.4)$$

$$\implies \mathbf{A}\mathbf{x} = \mathbf{b} \quad (1.3.4.5)$$

Where,

$$\mathbf{A} = \begin{pmatrix} 1 & -1 & 2 \\ 1 & 0 & 2 \\ 1 & -3 & 4 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}, \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$
 (1.3.4.7)

Solving the matrix A for rank we get,

$$\begin{pmatrix} 1 & -1 & 2 \\ 2 & 0 & 2 \\ 1 & -3 & 4 \end{pmatrix} \xrightarrow{R_2 = R_1 - 2R_1} \begin{pmatrix} 1 & -1 & 2 \\ 0 & 2 & -2 \\ 1 & -3 & 4 \end{pmatrix} (1.3.4.8)$$

$$\xrightarrow{R_3 = R_3 - R_1} \begin{pmatrix} 1 & -1 & 2 \\ 0 & 2 & -2 \\ 0 & -2 & 2 \end{pmatrix} (1.3.4.9)$$

$$\xrightarrow{R_3 = R_3 + R_2} \begin{pmatrix} 1 & -1 & 2 \\ 0 & 2 & -2 \\ 0 & 0 & 0 \end{pmatrix} (1.3.4.10)$$

Hence, rank (A) = 2. Now solving the augmented matrix of (1.3.4.5) we get,

$$\begin{pmatrix}
1 & -1 & 2 & 1 \\
2 & 0 & 2 & 1 \\
1 & -3 & 4 & 2
\end{pmatrix}
\xrightarrow{R_2=R_1-2R_1}
\begin{pmatrix}
1 & -1 & 2 & 1 \\
0 & 2 & -2 & -1 \\
1 & -3 & 4 & 2
\end{pmatrix}$$

$$(1.3.4.11)$$

$$\xrightarrow{R_3=R_3-R_1}
\begin{pmatrix}
1 & -1 & 2 & 1 \\
0 & 2 & -2 & -1 \\
0 & -2 & 2 & 1
\end{pmatrix}$$

$$(1.3.4.12)$$

$$\xrightarrow{R_3=R_3+R_2}
\begin{pmatrix}
1 & -1 & 2 & 1 \\
0 & 2 & -2 & -1 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$(1.3.4.13) 1.3.5$$

We have rank $(\mathbf{A}) = \text{rank } (\mathbf{A} : \mathbf{b}) = 2 < n$, where n = 3. Hence we have infinite no of solutions for given system of equations.

Using Gauss - Jordan elimination method to getting the solution,

$$\begin{pmatrix} 1 & -1 & 2 & 1 \\ 2 & 0 & 2 & 1 \\ 1 & -3 & 4 & 2 \end{pmatrix} \xrightarrow{R_2 = R_1 - 2R_1} \begin{pmatrix} 1 & -1 & 2 & 1 \\ 0 & 2 & -2 & -1 \\ 1 & -3 & 4 & 2 \end{pmatrix}$$
(1.3.4.14)

$$\stackrel{R_3=R_3-R_1}{\longleftrightarrow} \begin{pmatrix} 1 & -1 & 2 & 1\\ 0 & 2 & -2 & -1\\ 0 & -2 & 2 & 1 \end{pmatrix}$$
 (1.3.4.15)

$$\stackrel{R_2 = \frac{R_2}{2}}{\longleftrightarrow} \begin{pmatrix} 1 & -1 & 2 & 1 \\ 0 & 1 & -1 & -\frac{1}{2} \\ 0 & -2 & 2 & 1 \end{pmatrix}$$
(1.3.4.16)

$$\stackrel{R_3=R_3+2R_2}{\longleftrightarrow} \begin{pmatrix} 1 & -1 & 2 & 1 \\ 0 & 1 & -1 & -\frac{1}{2} \\ 0 & 0 & 0 & 0 \end{pmatrix}$$
 (1.3.4.17)

$$\stackrel{R_1=R_1+R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 1 & \frac{1}{2} \\ 0 & 1 & -1 & -\frac{1}{2} \\ 0 & 0 & 0 & 0 \end{pmatrix}$$
 (1.3.4.18)

$$\implies x_1 + x_3 = \frac{1}{2}, x_2 - x_3 = -\frac{1}{2}$$
 (1.3.4.19)

$$\implies x_2 = -\frac{1}{2} + x_3, x_1 = \frac{1}{2} - x_3 \quad (1.3.4.20)$$

From equation (1.3.4.19) and (1.3.4.20)

$$\mathbf{x} = \begin{pmatrix} \frac{1}{2} - x_3 \\ -\frac{1}{2} + x_3 \\ x_3 \end{pmatrix}$$
 (1.3.4.21)

which can be written as,

$$\mathbf{x} = x_3 \begin{pmatrix} -1\\1\\1 \end{pmatrix} + \begin{pmatrix} -\frac{1}{2}\\-\frac{1}{2}\\0 \end{pmatrix}$$
 (1.3.4.22)

from 1.3.4.22 we can say that for any value x_3 , **V** will no be gives zero vector. Hence the given solution space will not span of the vector space **V**

1.3.4.13) 1.3.5. Give an example of a system of two linear equations in two unknowns which has no solution.

Solution: Let us assume two equations as given below $(5 2)\mathbf{x} = 7$ and $(10 4)\mathbf{x} = -3$

Let the coefficient matrix be given as

$$\mathbf{A} = \begin{pmatrix} 5 & 2 \\ 10 & 4 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 7 \\ -3 \end{pmatrix} \tag{1.3.5.1}$$

the augmented matrix be given as matrix be given as

$$\mathbf{A}|\mathbf{B} = \begin{pmatrix} 5 & 2 & 7 \\ 10 & 4 & -3 \end{pmatrix} \tag{1.3.5.2}$$

Applying row reduction

$$\begin{pmatrix} 5 & 2 & 7 \\ 10 & 4 & -3 \end{pmatrix} \xrightarrow{R_2 = R_2 - 2R_1} \begin{pmatrix} 5 & 2 & 7 \\ 0 & 0 & -17 \end{pmatrix} (1.3.5.3)$$

$$\xrightarrow{R_1 = \frac{R_1}{5}} \begin{pmatrix} 1 & \frac{2}{5} & \frac{7}{5} \\ 0 & 0 & -17 \end{pmatrix} (1.3.5.4)$$

$$\xrightarrow{R_2 = \frac{R_2}{-17}} \begin{pmatrix} 1 & \frac{2}{5} & \frac{7}{5} \\ 0 & 0 & 1 \end{pmatrix} (1.3.5.5)$$

$$\xrightarrow{R_1 = R_1 - \frac{7}{5}R_2} \begin{pmatrix} 1 & \frac{2}{5} & 0 \\ 0 & 0 & 1 \end{pmatrix} (1.3.5.6)$$

$$(1.3.5.7)$$

Clearly, On comparing the ranks of matrix A and A|B, we find that rank of matrix $A|B \neq A$ Hence the system of linear equation have no solutions. Consider the system Ax = b, with coefficient matrix A and augmented matrix A|B.

As above, the sizes of **b**, **A**, and A|B are m \times 1, m \times n, and m \times (n + 1), respectively; in addition, the number of unknowns is n.

Ax is inconsistent (i.e., no solution exists) if 1.3.7. Find all solutions of and only if rank A < rank A | B.

1.3.6. Find all solutions of

$$x_1 - 2x_2 + x_3 + 2x_4 = 1$$

$$x_1 + x_2 - x_3 + x_4 + x_5 = 2$$

$$x_1 + 7x_2 - 5x_3 - x_4 = 3$$

Solution: The given equations can be written as,

$$Ax = B$$
 (1.3.6.1)

$$\begin{pmatrix} 1 & -2 & 1 & 2 \\ 1 & 1 & -1 & 1 \\ 1 & 7 & -5 & -1 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$
 (1.3.6.2)

Now, we form the augmented matrix and per-

form Row reduction,

$$\begin{pmatrix}
1 & -2 & 1 & 2 & | & 1 \\
1 & 1 & -1 & 1 & | & 2 \\
1 & 7 & -5 & -1 & | & 3
\end{pmatrix}$$

$$(1.3.6.3)$$

$$\xrightarrow{R_2 = R_2 - R_1, R_3 = R_3 - R_1}
\begin{pmatrix}
1 & -2 & 1 & 2 & | & 1 \\
0 & 3 & -2 & -1 & | & 1 \\
0 & 9 & -6 & -3 & | & 2
\end{pmatrix}$$

$$(1.3.6.4)$$

$$\xrightarrow{R_2 = \frac{1}{3}R_2}
\begin{pmatrix}
1 & -2 & 1 & 2 & | & 1 \\
0 & 1 & \frac{-2}{3} & \frac{-1}{3} & | & \frac{1}{3} \\
0 & 9 & -6 & -3 & | & 2
\end{pmatrix}$$

$$(1.3.6.5)$$

$$\xrightarrow{R_3 = R_3 - 9R_1}
\begin{pmatrix}
1 & -2 & 1 & 2 & | & 1 \\
0 & 1 & \frac{-2}{3} & \frac{-1}{3} & | & \frac{1}{3} \\
0 & 0 & 0 & 0 & | & -1
\end{pmatrix}$$

$$(1.3.6.6)$$

Rank of A is less than rank of the augmented matrix. Hence, the given system has no solution.

$$2x_1 - 3x_2 - 7x_3 + 5x_4 + 2x_5 = -2 (1.3.7.1)$$

$$x_1 - 2x_2 - 4x_3 + 3x_4 + x_5 = -2 (1.3.7.2)$$

$$2x_1 - 4x_3 + 2x_4 + x_5 = 3 (1.3.7.3)$$

$$x_1 - 5x_2 - 7x_3 + 6x_4 + 2x_5 = -7 (1.3.7.4)$$

Solution: The given equations can be written as,

$$\begin{pmatrix} 2 & -3 & -7 & 5 & 2 \\ 1 & -2 & -4 & 3 & 1 \\ 2 & 0 & -4 & 2 & 1 \\ 1 & -5 & -7 & 6 & 2 \end{pmatrix} \mathbf{x} = \begin{pmatrix} -2 \\ -2 \\ 3 \\ 7 \end{pmatrix}$$
 (1.3.7.5)

Now, we form the augmented matrix and per-

form Row reduction,

$$\begin{pmatrix}
2 & -3 & -7 & 5 & 2 & | & -2 \\
1 & -2 & -4 & 3 & 1 & | & -2 \\
2 & 0 & -4 & 2 & 1 & | & 3 \\
1 & -5 & -7 & 6 & 2 & | & 7
\end{pmatrix}$$

$$\stackrel{R_3=R_3-R_1}{\longleftrightarrow} \begin{pmatrix}
2 & -3 & -7 & 5 & 2 & | & -2 \\
1 & -2 & -4 & 3 & 1 & | & -2 \\
0 & 3 & 3 & -3 & -1 & | & 5 \\
1 & -5 & -7 & 6 & 2 & | & 7
\end{pmatrix}$$
(1.3.7.7)

$$\stackrel{R_1 = \frac{1}{2}R_1}{\longleftrightarrow} \begin{pmatrix}
1 & \frac{-3}{2} & \frac{-7}{2} & \frac{5}{2} & 1 & | & -1 \\
1 & -2 & -4 & 3 & 1 & | & -2 \\
0 & 3 & 3 & -3 & -1 & | & 5 \\
1 & -5 & -7 & 6 & 2 & | & 7
\end{pmatrix}$$
(1.3.7.8)

$$\stackrel{R_2=R_2-R_1,R_4=R_4-R_1}{\longleftrightarrow} \begin{pmatrix}
1 & \frac{-3}{2} & \frac{-7}{2} & \frac{5}{2} & 1 & | & -1\\ 0 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} & 0 & | & -1\\ 0 & 3 & 3 & -3 & -1 & | & 5\\ 0 & -\frac{7}{2} & -\frac{7}{2} & \frac{7}{2} & 1 & | & -6\end{pmatrix}$$
(1.3.7.9)

$$\stackrel{R_1=R_1-3R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -2 & 1 & 1 & 2 \\
0 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} & 0 & -1 \\
0 & 3 & 3 & -3 & -1 & 5 \\
0 & -\frac{7}{2} & -\frac{7}{2} & \frac{7}{2} & 1 & -6
\end{pmatrix}$$

$$\stackrel{R_3=R_3+6R_2,R_4=R_4-7R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -2 & 1 & 1 & 2 \\
0 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} & 0 & -1 \\
0 & 0 & 0 & 0 & -1 & -1 \\
0 & 0 & 0 & 0 & 1 & 1
\end{pmatrix}$$
(1.3.7.11)

$$\xrightarrow{R_2 = -2R_2} \begin{pmatrix} 1 & 0 & -2 & 1 & 1 & 2 \\ 0 & 1 & 1 & -1 & 0 & 2 \\ 0 & 0 & 0 & 0 & -1 & -1 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix} 1.3.8. \text{ Let}$$

$$(1.3.7.12)$$

$$\stackrel{R_1=R_1+R_3,R_4=R_4+R_3,R_3=-R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -2 & 1 & 0 & | & 1 \\
0 & 1 & 1 & -1 & 0 & | & 2 \\
0 & 0 & 0 & 0 & 1 & | & 1 \\
0 & 0 & 0 & 0 & 0 & | & 0
\end{pmatrix}$$
(1.3.7.13)

So,

$$x_1 - 2x_3 + x_4 = 1 (1.3.7.14)$$

$$x_2 + x_3 - x_4 = 2$$
 (1.3.7.15)
 $x_5 = 1$ (1.3.7.16)

$$x_5 = 1 \tag{1.3.7.16}$$

Solving the equations we get,

$$x_1 = 1 + 2x_3 - x_4 \tag{1.3.7.17}$$

$$x_2 = 2 - x_3 + x_4 \tag{1.3.7.18}$$

$$x_5 = 1 \tag{1.3.7.19}$$

which can be written as,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 (1.3.7.20)

$$\implies \mathbf{x} = \begin{pmatrix} 1 + 2x_3 - x_4 \\ 2 - x_3 + x_4 \\ x_3 \\ x_4 \\ 1 \end{pmatrix}$$
 (1.3.7.21)

We can express (1.3.7.21) as a sum of linear combination of vectors,

$$\mathbf{x} = \begin{pmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 1 \end{pmatrix} + \begin{pmatrix} 2 \\ -1 \\ 1 \\ 0 \\ 0 \end{pmatrix} \mathbf{x_3} + \begin{pmatrix} -1 \\ 1 \\ 0 \\ 1 \\ 0 \end{pmatrix} \mathbf{x_4}$$
 (1.3.7.22)

where $x_3, x_4 \in \mathbb{R}$.

Note that the above solution space is not closed on vector addition and scalar multiplication. As $x_5 = 1$, the zero vector is not included in the solution space. Hence, x is not a vector space. Since, x is not a vector space, it cannot be expressed in the form of linear combination of basis vectors.

$$\mathbf{A} = \begin{pmatrix} 3 & -1 & 2 \\ 2 & 1 & 1 \\ 1 & -3 & 0 \end{pmatrix} \tag{1.3.8.1}$$

For which triples (y_1, y_2, y_3) does the system AX = Y have a solution?

Solution:

Given,

$$\mathbf{AX} = \mathbf{Y} \tag{1.3.8.2}$$

$$\begin{pmatrix} 3 & -1 & 2 \\ 2 & 1 & 1 \\ 1 & -3 & 0 \end{pmatrix} \mathbf{X} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix}$$
 (1.3.8.3)

Now we try to find the matrix **B** such that **BA** gives the row echelon form of matrix A.

Here, **B** is given by,

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 \\ -\frac{2}{3} & 1 & 0 \\ -\frac{7}{5} & \frac{8}{5} & 1 \end{pmatrix} \tag{1.3.8.4}$$

$$\implies \mathbf{BA} = \begin{pmatrix} 3 & -1 & 2 \\ 0 & \frac{5}{3} & -\frac{1}{3} \\ 0 & 0 & -\frac{6}{5} \end{pmatrix}$$
 (1.3.8.5)

Therefore, from (1.3.8.5) rank of matrix **A** is 3 and it is a full rank matrix.

Hence the columns of **A** are linearly independent.

Therefore, the triples (y_1, y_2, y_3) are linear combination of columns of matrix **A**.

$$\implies \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = a \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix} + b \begin{pmatrix} -1 \\ 1 \\ -3 \end{pmatrix} + c \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix} \quad (1.3.8.6)$$

where a,b,c can be any real value.

1.3.9. Let

$$\mathbf{A} = \begin{pmatrix} 3 & -6 & 2 & -1 \\ -2 & 4 & 1 & 3 \\ 0 & 0 & 1 & 1 \\ 1 & -2 & 1 & 0 \end{pmatrix} \tag{1.3.9.1}$$

For which (y_1, y_2, y_3, y_4) does the system of equations $\mathbf{AX} = \mathbf{Y}$ have a solution? **Solution:** Given,

$$\mathbf{AX} = \mathbf{Y} \tag{1.3.9.2}$$

$$\begin{pmatrix}
3 & -6 & 2 & -1 \\
-2 & 4 & 1 & 3 \\
0 & 0 & 1 & 1 \\
1 & -2 & 1 & 0
\end{pmatrix} \mathbf{X} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$
(1.3.9.3)

Now we try to find the matrix B such that BA gives the row echelon form of matrix A Here,B is given by ,

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ \frac{2}{3} & 1 & 0 & 0 \\ -\frac{2}{7} & -\frac{3}{7} & 1 & 0 \\ 0 & \frac{1}{2} & -\frac{3}{2} & 1 \end{pmatrix}$$
 (1.3.9.4)

$$\mathbf{BA} = \begin{pmatrix} 3 & -6 & 2 & -1 \\ 0 & 0 & \frac{7}{3} & \frac{7}{3} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \tag{1.3.9.5}$$

Therefore, rank of matrix A is 2 Now B is

expressed in terms of two block matrices

$$\mathbf{B} = \begin{pmatrix} \mathbf{B_1} \\ \mathbf{B_2} \end{pmatrix} \tag{1.3.9.6}$$

$$\mathbf{B_1} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ \frac{2}{3} & 1 & 0 & 0 \end{pmatrix} \tag{1.3.9.7}$$

$$\mathbf{B_2} = \begin{pmatrix} -\frac{2}{7} & -\frac{3}{7} & 1 & 0\\ 0 & \frac{1}{2} & -\frac{3}{2} & 1 \end{pmatrix}$$
 (1.3.9.8)

Multiplying matrix \mathbf{B} to both sides on the equation (1.3.9.2), we get,

$$\begin{pmatrix} \mathbf{B_1} \\ \mathbf{B_2} \end{pmatrix} \mathbf{AX} = \begin{pmatrix} \mathbf{B_1} \\ \mathbf{B_2} \end{pmatrix} \mathbf{Y} \tag{1.3.9.9}$$

We know that , matrix A is of rank 2 The augumented matrix of (1.3.9.9) is given by

$$\begin{pmatrix}
\mathbf{B_1 A} & \mathbf{B_1 Y} \\
\mathbf{B_2 A} & \mathbf{B_2 Y}
\end{pmatrix}$$
(1.3.9.10)

$$\mathbf{B_1A} = \begin{pmatrix} 3 & -6 & 2 & -1 \\ 0 & 0 & \frac{7}{3} & \frac{7}{3} \end{pmatrix} \tag{1.3.9.11}$$

$$\mathbf{B_2A} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \tag{1.3.9.12}$$

Since B_2A is zero matrix and for the given system AX = Y to have a solution,

$$\mathbf{B_2Y} = 0 \qquad (1.3.9.13)$$

$$\begin{pmatrix} -\frac{2}{7} & -\frac{3}{7} & 1 & 0 \\ 0 & \frac{1}{2} & -\frac{3}{2} & 1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} = 0$$
 (1.3.9.14)

The augumented matrix of (1.3.9.14) is given by,

$$\begin{pmatrix} -\frac{2}{7} & -\frac{3}{7} & 1 & 0 & 0 \\ 0 & \frac{1}{2} & -\frac{3}{2} & 1 & 0 \end{pmatrix}$$
 (1.3.9.15)

By row reduction technique,

$$\stackrel{R_1 = -\frac{7}{2}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{3}{2} & -\frac{7}{2} & 0 \\ 0 & \frac{1}{2} & -\frac{3}{2} & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$
 (1.3.9.16)

$$\stackrel{R_2=2R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{3}{2} & -\frac{7}{2} & 0 & | & 0 \\ 0 & 1 & -3 & 2 & | & 0 \end{pmatrix}$$
 (1.3.9.17)

$$\stackrel{R_1 = R_1 - \frac{3}{2}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 1 & -3 & | & 0 \\ 0 & 1 & -3 & 2 & | & 0 \end{pmatrix} \quad (1.3.9.18)$$

Equation (1.3.9.14) can be modified as,

$$\begin{pmatrix} 1 & 0 & 1 & -3 \\ 0 & 1 & -3 & 2 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} = 0$$
 (1.3.9.19)

Here y_3 and y_4 are free variables If $y_3 = a$ and $y_4 = b$, then the solution to the system of equation AX = Y is given by,

$$\begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} = a \begin{pmatrix} -1 \\ 3 \\ 1 \\ 0 \end{pmatrix} + b \begin{pmatrix} 3 \\ -2 \\ 0 \\ 1 \end{pmatrix}$$
 (1.3.9.20)

One of the solution when a = 1 and b = 2 is given by,

$$\begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} = \begin{pmatrix} -1 \\ 3 \\ 1 \\ 0 \end{pmatrix} + 2 \begin{pmatrix} 3 \\ -2 \\ 0 \\ 1 \end{pmatrix}$$
 (1.3.9.21)

1.3.10. Suppose \mathbf{R} and \mathbf{R}' are 2×3 row-reduced echelon matrices and that the system \mathbf{RX} =0 and $\mathbf{R}'\mathbf{X}$ =0 have exactly the same solutions. Prove that $\mathbf{R} = \mathbf{R}'$.

Solution:

Since **R** and **R**' are 2×3 row-reduced echelon matrices they can be of following three types:-

a) Suppose matrix **R** has one non-zero row then **RX**=0 will have two free variables. Since **R**'**X**=0 will have the exact same solution as **RX** = 0, **R**'**X**=0 will also have two free variables. Thus **R**' have one non zero row. Now let's consider a matrix **A** with the first row as the non-zero row **R** and second row as the second row of **R**'.

$$\mathbf{R} = \begin{pmatrix} 1 & a & b \\ 0 & 0 & 0 \end{pmatrix} \tag{1.3.10.1}$$

$$\mathbf{R}' = \begin{pmatrix} 1 & c & d \\ 0 & 0 & 0 \end{pmatrix} \tag{1.3.10.2}$$

(1.3.10.3)

Let X satisfy

$$\mathbf{RX} = 0$$
 (1.3.10.4)

$$(1 \quad \mathbf{a}^T) \begin{pmatrix} x \\ \mathbf{y} \end{pmatrix} = 0$$
 (1.3.10.5)

$$x + \mathbf{a}^T \mathbf{y} = 0 \tag{1.3.10.6}$$

where

$$\mathbf{a} = \begin{pmatrix} a \\ b \end{pmatrix} \tag{1.3.10.7}$$

$$\mathbf{R}'\mathbf{X} = 0 \tag{1.3.10.8}$$

$$\begin{pmatrix} 1 & \mathbf{b}^T \end{pmatrix} \begin{pmatrix} x \\ \mathbf{y} \end{pmatrix} = 0 \tag{1.3.10.9}$$

$$x + \mathbf{b}^T \mathbf{y} = 0 \tag{1.3.10.10}$$

where

$$\mathbf{b} = \begin{pmatrix} c \\ d \end{pmatrix} \tag{1.3.10.11}$$

Subtracting (1.3.10.10) from (1.3.10.6),

$$x + \mathbf{a}^T \mathbf{y} - x - \mathbf{b}^T \mathbf{y} = 0$$
 (1.3.10.12)

$$(\mathbf{a}^T - \mathbf{b}^T)\mathbf{y} = 0 \qquad (1.3.10.13)$$

Since y is a 2×1 vector,

$$\implies y_1 \mathbf{a} - y_2 \mathbf{b} = 0 \tag{1.3.10.14}$$

Which can be written as,

$$\mathbf{a} = k\mathbf{b} \tag{1.3.10.15}$$

where, $k = \frac{y_2}{y_1}$ assuming $\mathbf{y}_1 \neq 0$. Now, Substituting (1.3.10.15) in (1.3.10.6)

$$x + k\mathbf{b}^T \mathbf{y} = 0 ag{1.3.10.16}$$

Comparing (1.3.10.16) with (1.3.10.10)

$$x + \mathbf{b}^T \mathbf{y} = 0 \tag{1.3.10.17}$$

$$x + k\mathbf{b}^T\mathbf{y} = 0 \tag{1.3.10.18}$$

Hence k=1 which means $y_1=y_2$ and from this we can say that $\mathbf{a}=\mathbf{b}$. If in the above case we take $y_1=0$ then

$$y_1 \mathbf{a} - y_2 \mathbf{b} = 0 \tag{1.3.10.19}$$

$$y_2 \mathbf{b} = 0$$
 (1.3.10.20)

Hence for the (1.3.10.20) to be always true **b** should be zero. Now from (1.3.10.15) we will see that **a** will also be 0. Hence, $\mathbf{R} = \mathbf{R}'$

b) Let **R** and **R** have all rows as non zero.

$$\mathbf{R} = \begin{pmatrix} 1 & 0 & b \\ 0 & 1 & c \end{pmatrix} \tag{1.3.10.21}$$

$$\mathbf{R}' = \begin{pmatrix} 1 & 0 & e \\ 0 & 1 & f \end{pmatrix} \tag{1.3.10.22}$$

Let X satisfy

$$\mathbf{RX} = 0$$
 (1.3.10.23)

$$\mathbf{X}^T \mathbf{R}^T = 0 \tag{1.3.10.24}$$

Here,

$$\mathbf{R} = \begin{pmatrix} \mathbf{I} & \mathbf{a} \end{pmatrix}$$
 (1.3.10.25) *1.4 Matrix Multiplication*

$$\mathbf{a} = \begin{pmatrix} b \\ c \end{pmatrix}$$
 (1.3.10.26) 1.4.1. Let

$$\mathbf{R}^T = \begin{pmatrix} \mathbf{I} \\ \mathbf{a}^T \end{pmatrix} \tag{1.3.10.27}$$

Let,

$$\mathbf{X}^T = \begin{pmatrix} \mathbf{y}^T & z \end{pmatrix} \tag{1.3.10.28}$$

where z is a scalar constant. Now, substituting (1.3.10.28) and (1.3.10.25) in (1.3.10.24)

$$(\mathbf{y}^T \quad z) \begin{pmatrix} \mathbf{I} \\ \mathbf{a}^T \end{pmatrix} = 0$$
 (1.3.10.29)

$$\mathbf{y}^T + z\mathbf{a}^T = 0 (1.3.10.30)$$

Now for,

$$\mathbf{R}'\mathbf{X} = 0 \tag{1.3.10.31}$$

$$\mathbf{X}^T \mathbf{R'}^T = 0 \tag{1.3.10.32}$$

Here,

$$\mathbf{R}' = \begin{pmatrix} \mathbf{I} & \mathbf{b} \end{pmatrix} \tag{1.3.10.33}$$

$$\mathbf{b} = \begin{pmatrix} e \\ f \end{pmatrix} \tag{1.3.10.34}$$

Let,

$$\mathbf{X}^T = \begin{pmatrix} \mathbf{y}^T & z \end{pmatrix} \tag{1.3.10.35}$$

where z is a scalar constant. Now, substituting (1.3.10.35) and (1.3.10.33) in (1.3.10.32)

$$\begin{pmatrix} \mathbf{y}^T & z \end{pmatrix} \begin{pmatrix} \mathbf{I} \\ \mathbf{b}^T \end{pmatrix} = 0 \tag{1.3.10.36}$$

$$\mathbf{y}^T + z\mathbf{b}^T = 0 \tag{1.3.10.37}$$

Subtracting (1.3.10.37) from (1.3.10.30)

$$\mathbf{y}^T + z\mathbf{a}^T - \mathbf{y}^T - z\mathbf{b}^T = 0$$
 (1.3.10.38)

$$(\mathbf{a}^T - \mathbf{b}^T)z = 0$$
 (1.3.10.39)

$$\mathbf{a}^T = \mathbf{b}^T \qquad (1.3.10.40)$$

c) Suppose matrix **R** have all the rows as zero

then $\mathbf{R}\mathbf{X}=0$ will be satisfied for all values of \mathbf{X} . We know that $\mathbf{R}'\mathbf{X}=0$ will have the exact same solution as $\mathbf{R}\mathbf{X}=0$ then we can say that for all values of $\mathbf{X}=0$ equation $\mathbf{R}'\mathbf{X}=0$ will be satisfied.Hence, $\mathbf{R}'=\mathbf{R}=0$.

 $\mathbf{A} = \begin{pmatrix} 2 & -1 & 1 \\ 1 & 2 & 1 \end{pmatrix}, \mathbf{B} = \begin{pmatrix} 3 \\ 1 \\ 1 \end{pmatrix}, \quad \mathbf{C} = \begin{pmatrix} 1 & -1 \end{pmatrix}$

(1.4.1.1)

Compute ABC and CAB.

Solution: Given,

$$\mathbf{A} = \begin{pmatrix} 2 & -1 & 1 \\ 1 & 2 & 1 \end{pmatrix} \tag{1.4.1.2}$$

$$\mathbf{B} = \begin{pmatrix} 3 \\ 1 \\ -1 \end{pmatrix} \tag{1.4.1.3}$$

$$\mathbf{C} = \begin{pmatrix} 1 & -1 \end{pmatrix} \tag{1.4.1.4}$$

Take, ABC = (AB) C

$$\mathbf{AB} = \begin{pmatrix} 2 & -1 & 1 \\ 1 & 2 & 1 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \\ -1 \end{pmatrix}$$
 (1.4.1.5)

$$\mathbf{AB} = \begin{pmatrix} 6 - 1 - 1 \\ 3 + 2 - 1 \end{pmatrix} \tag{1.4.1.6}$$

$$\mathbf{AB} = \begin{pmatrix} 4\\4 \end{pmatrix} \tag{1.4.1.7}$$

Now,

$$\mathbf{ABC} = \begin{pmatrix} 4 \\ 4 \end{pmatrix} \begin{pmatrix} 1 & -1 \end{pmatrix} \tag{1.4.1.8}$$

$$\mathbf{ABC} = \begin{pmatrix} 4 & -4 \\ 4 & -4 \end{pmatrix} \tag{1.4.1.9}$$

similarly, CAB = C(AB)

$$\mathbf{CAB} = \begin{pmatrix} 1 & -1 \end{pmatrix} \begin{pmatrix} 4 \\ 4 \end{pmatrix} \tag{1.4.1.10}$$

$$\implies \mathbf{CAB} = 0 \tag{1.4.1.11}$$

therefore,

$$\mathbf{ABC} = \begin{pmatrix} 4 & -4 \\ 4 & -4 \end{pmatrix} \tag{1.4.1.12}$$

$$CAB = 0$$
 (1.4.1.13)

1.4.2. Let

$$A = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix}, B = \begin{pmatrix} 2 & -2 \\ 1 & 3 \\ 4 & 4 \end{pmatrix}$$
 (1.4.2.1)

Verify directly that $A(AB) = A^2B$ Solution:

$$A^{2} = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix}$$
 (1.4.2.2)

$$A^{2} = \begin{pmatrix} 2 & -1 & 1 \\ 5 & -2 & 3 \\ 6 & -3 & 4 \end{pmatrix}$$
 (1.4.2.3)

and

$$AB = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix} \begin{pmatrix} 2 & -2 \\ 1 & 3 \\ 4 & 4 \end{pmatrix}$$
 (1.4.2.4)

$$AB = \begin{pmatrix} 5 & -1 \\ 8 & 0 \\ 10 & -2 \end{pmatrix} \tag{1.4.2.5}$$

Now RHS is

$$A^{2}B = \begin{pmatrix} 2 & -1 & 1 \\ 5 & -2 & 3 \\ 6 & -3 & 4 \end{pmatrix} \begin{pmatrix} 2 & -2 \\ 1 & 3 \\ 4 & 4 \end{pmatrix}$$
 (1.4.2.6)

$$A^2B = \begin{pmatrix} 7 & -3\\ 20 & -4\\ 25 & -5 \end{pmatrix} \tag{1.4.2.7}$$

Now LHS is

$$A(AB) = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix} \begin{pmatrix} 5 & -1 \\ 8 & 0 \\ 10 & -2 \end{pmatrix}$$
 (1.4.2.8)

$$A(AB) = \begin{pmatrix} 7 & -3 \\ 20 & -4 \\ 25 & -5 \end{pmatrix} \tag{1.4.2.9}$$

Hence verified.

1.4.3. Find two different 2×2 matrices **A** such that $\mathbf{A}^2 = 0$ but $\mathbf{A} \neq 0$

Solution: The matrix **A** can be given by,

$$\mathbf{A} = \begin{pmatrix} \mathbf{m} & \mathbf{n} \end{pmatrix} \tag{1.4.3.1}$$

$$\mathbf{m} = \begin{pmatrix} m_1 \\ m_2 \end{pmatrix}, \mathbf{n} = \begin{pmatrix} n_1 \\ n_2 \end{pmatrix} \tag{1.4.3.2}$$

Now,

$$\mathbf{A}^2 = \mathbf{A}\mathbf{A} = \mathbf{0} \tag{1.4.3.3}$$

$$\implies$$
 $\mathbf{A}^2 = (\mathbf{Am} \ \mathbf{An}) = (\mathbf{0} \ \mathbf{0}) \ (1.4.3.4)$

From (1.4.3.4), we say that the the null space of **A** contains columns of matrix **A**. Also atleast one of the columns must be non-zero since given $\mathbf{A} \neq 0$. Thus, the null space of **A** contains non zero vectors, $rank(\mathbf{A}) < 2$. Hence, **A** is a singular matrix. This implies that the columns of **A** are linearly dependent.

$$\mathbf{A}\mathbf{x} = 0 \tag{1.4.3.5}$$

$$\begin{pmatrix} \mathbf{m} & \mathbf{n} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0 \tag{1.4.3.6}$$

$$x_1 \mathbf{m} + x_2 \mathbf{n} = 0 \tag{1.4.3.7}$$

$$\mathbf{n} = \frac{-x_1}{x_2} \mathbf{m} \tag{1.4.3.8}$$

$$\implies$$
 n = k **m** (1.4.3.9)

where $\mathbf{m} \neq 0$ as $\mathbf{A} \neq 0$ Now from (1.4.3.4),

$$\mathbf{Am} = 0$$
 (1.4.3.10)

$$m_1 \mathbf{m} + m_2 \mathbf{n} = 0 \tag{1.4.3.11}$$

$$(m_1 + km_2) \mathbf{m} = 0 (1.4.3.12)$$

Thus we get, $m_1 = -km_2$

$$\mathbf{A} = \begin{pmatrix} -km_2 & -k^2m_2 \\ m_2 & km_2 \end{pmatrix}; m_2 \neq 0 \qquad (1.4.3.13)$$

(1.4.3.9) can be written as,

$$\implies \mathbf{m} = \frac{1}{k}\mathbf{n} \tag{1.4.3.14}$$

$$\implies \mathbf{m} = c\mathbf{n} \tag{1.4.3.15}$$

where $\mathbf{n} \neq 0$ as $\mathbf{A} \neq 0$ From (1.4.3.4),

$$\mathbf{An} = 0$$
 (1.4.3.16)

$$n_1 \mathbf{m} + n_2 \mathbf{n} = 0 \tag{1.4.3.17}$$

$$(cn_1 + n_2)\mathbf{n} = 0 (1.4.3.18)$$

Thus we get, $n_2 = -cn_1$

$$\mathbf{A} = \begin{pmatrix} cn_1 & n_1 \\ -c^2 n_1 & -cn_1 \end{pmatrix}; n_1 \neq 0 \qquad (1.4.3.19)$$

From (1.4.3.13), (1.4.3.19) two different 2×2

matrices A can be given as,

$$\mathbf{A} = \begin{pmatrix} 0 & 0 \\ 2 & 0 \end{pmatrix} \tag{1.4.3.20}$$

$$\mathbf{A} = \begin{pmatrix} 0 & 2 \\ 0 & 0 \end{pmatrix} \tag{1.4.3.21}$$

1.4.4. For the matrix $\mathbf{A} = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix}$, find elementary matrices $\mathbf{E_1}, \mathbf{E_2}, \dots, \mathbf{E_k}$ such that

$$\mathbf{E_k}...\mathbf{E_2}\mathbf{E_1}\mathbf{A} = \mathbf{I}$$
 (1.4.4.1)

Solution: Given,

$$\mathbf{A} = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix} \tag{1.4.4.2}$$

Take,

$$\mathbf{E_1} = \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{1.4.4.3}$$

$$\mathbf{E_2} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -3 & 0 & 1 \end{pmatrix} \tag{1.4.4.4}$$

$$\mathbf{E_3} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{1.4.4.5}$$

$$\mathbf{E_4} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{1.4.4.6}$$

$$\mathbf{E_5} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -3 & 1 \end{pmatrix} \tag{1.4.4.7}$$

$$\mathbf{E_6} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \frac{2}{5} \end{pmatrix} \tag{1.4.4.8}$$

$$\mathbf{E}_7 = \begin{pmatrix} 1 & 0 & \frac{-1}{2} \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{1.4.4.9}$$

$$\mathbf{E_8} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & \frac{1}{2} \\ 0 & 0 & 1 \end{pmatrix} \tag{1.4.4.10}$$

Now, we calculate

$$\mathbf{E_8}\mathbf{E_7}\mathbf{E_6}\mathbf{E_5}\mathbf{E_4}\mathbf{E_3}\mathbf{E_2}\mathbf{E_1} = \begin{pmatrix} 0 & -1 & 1 \\ -1 & 2 & -1 \\ 0 & 3 & -2 \end{pmatrix}$$
(1.4.4.11)

Hence,

$$(\mathbf{E}_{8}\mathbf{E}_{7}\mathbf{E}_{6}\mathbf{E}_{5}\mathbf{E}_{4}\mathbf{E}_{3}\mathbf{E}_{2}\mathbf{E}_{1})\mathbf{A} = \begin{pmatrix} 0 & -1 & 1 \\ -1 & 2 & -1 \\ 0 & 3 & -2 \end{pmatrix} \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

1.4.5. Let $A = \begin{pmatrix} 1 & -1 \\ 2 & 2 \\ 1 & 0 \end{pmatrix}$ and $B = \begin{pmatrix} 3 & 1 \\ -4 & 4 \end{pmatrix}$ Is there any matrix C such that CA = B?

Solution: The matrix B is obtained by multiplying the matrix A with matrix C. B is a 2×2 matrix and A is a 3×2 matrix. so matrix C must be a 2×3 matrix. Let the matrix C is:

$$C = \begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{pmatrix} \tag{1.4.5.1}$$

$$\implies C^T = \begin{pmatrix} a_1 & a_2 \\ b_1 & b_2 \\ c_1 & c_2 \end{pmatrix}$$
 (1.4.5.2)

So, after multiplying with A matrix we get,

$$\begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 2 & 2 \\ 1 & 0 \end{pmatrix} =$$

$$\begin{pmatrix} a_1 + 2b_1 + c_1 & -a_1 + 2b_1 \\ a_2 + 2b_2 + c_2 & -a_2 + 2b_2 \end{pmatrix}$$
 (1.4.5.3)

Matrix A is a rectangular matrix. Now, Considering CA = B and by transposing both side,

$$(CA)^{T} = B^{T}$$

$$(CA)^{T} = B^{T}$$

$$\Rightarrow A^{T}C^{T} = B^{T}$$

$$(1.4.5.4)$$

$$\Rightarrow \left(\begin{array}{ccc} 1 & 2 & 1 \\ -1 & 2 & 0 \end{array}\right) \begin{pmatrix} \mathbf{c_{1}} & \mathbf{c_{2}} \end{pmatrix} = \begin{pmatrix} 3 & -4 \\ 1 & 4 \end{pmatrix}$$

$$(1.4.5.6)$$

We can represent it like this:

$$\begin{pmatrix} 1 & 2 & 1 \\ -1 & 2 & 0 \end{pmatrix} \mathbf{c_1} = \begin{pmatrix} 3 \\ 1 \end{pmatrix}$$
 (1.4.5.7) (1.4.5.8)

Now the augmented matrix is:

$$\begin{pmatrix} 1 & 2 & 1 & 3 \\ -1 & 2 & 0 & 1 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_1 + R_2} \begin{pmatrix} 1 & 2 & 1 & 3 \\ 0 & 4 & 1 & 4 \end{pmatrix}$$

$$\xrightarrow{R_2 \leftarrow R_2/2} \begin{pmatrix} 1 & 2 & 1 & 3 \\ 0 & 2 & \frac{1}{2} & 2 \end{pmatrix} \xrightarrow{R_1 \leftarrow R_1 - R_2}$$

$$\begin{pmatrix} 1 & 0 & \frac{1}{2} & 1 \\ 0 & 2 & \frac{1}{2} & 2 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_2/2} \begin{pmatrix} 1 & 0 & \frac{1}{2} & 1 \\ 0 & 1 & \frac{1}{4} & 1 \end{pmatrix} \quad (1.4.5.9)$$

Similarly,

$$\begin{pmatrix} 1 & 2 & 1 \\ -1 & 2 & 0 \end{pmatrix} \mathbf{c_2} = \begin{pmatrix} -4 \\ 4 \end{pmatrix}$$
 (1.4.5.10) (1.4.5.11)

Now the augmented matrix is:

$$\begin{pmatrix} 1 & 2 & 1 & -4 \\ -1 & 2 & 0 & 4 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_1 + R_2} \begin{pmatrix} 1 & 2 & 1 & -4 \\ 0 & 4 & 1 & 0 \end{pmatrix} \implies CA = \begin{pmatrix} 3 & 1 \\ -4 & 4 \end{pmatrix} + c_1 \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} + c_2 \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\xrightarrow{R_2 \leftarrow R_2/2} \begin{pmatrix} 1 & 2 & 1 & -4 \\ 0 & 2 & \frac{1}{2} & 0 \end{pmatrix} \xleftarrow{R_1 \leftarrow R_1 - R_2} \implies CA = B \quad (1.4.5.18)$$
Hence, it is proved that there there exist a matrix C such that $CA = B$.
$$\begin{pmatrix} 1 & 0 & \frac{1}{2} & -4 \\ 0 & 2 & \frac{1}{2} & 0 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_2/2} \begin{pmatrix} 1 & 0 & \frac{1}{2} & -4 \\ 0 & 1 & \frac{1}{4} & 0 \end{pmatrix} \qquad 1.4.6. \text{ Let } \mathbf{A} \text{ be an } m \times n \text{ matrix and } \mathbf{B} \text{ be an } n \times k \text{ matrix.Show that the columns of } \mathbf{C} = \mathbf{C}$$

From equations 1.4.5.9 and 1.4.5.12, it can be observed that solutions exist and there is a matrix C such that CA = B. Now,

$$\mathbf{c_1} = \begin{pmatrix} 1 - \frac{c_1}{2} \\ 1 - \frac{c_1}{4} \\ c_1 \end{pmatrix} \tag{1.4.5.13}$$

$$\implies \mathbf{c_1} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + c_1 \begin{pmatrix} -\frac{1}{2} \\ -\frac{1}{4} \\ 1 \end{pmatrix} \qquad (1.4.5.14)$$

$$\mathbf{c_2} = \begin{pmatrix} -4 - \frac{c_2}{2} \\ -\frac{c_2}{4} \\ c_2 \end{pmatrix} \tag{1.4.5.15}$$

$$\implies \mathbf{c_2} = \begin{pmatrix} -4\\0\\0\\0 \end{pmatrix} + c_2 \begin{pmatrix} -\frac{1}{2}\\-\frac{1}{4}\\1 \end{pmatrix} \qquad (1.4.5.16)$$

Now,

$$C^{T} = \begin{pmatrix} 1 & -4 \\ 1 & 0 \\ 0 & 0 \end{pmatrix} + c_{1} \begin{pmatrix} -\frac{1}{2} & 0 \\ -\frac{1}{4} & 0 \\ 1 & 0 \end{pmatrix} + c_{2} \begin{pmatrix} 0 & -\frac{1}{2} \\ 0 & -\frac{1}{4} \\ 0 & 1 \end{pmatrix}$$

$$\implies C = \begin{pmatrix} 1 & 1 & 0 \\ -4 & 0 & 0 \end{pmatrix} + c_{1} \begin{pmatrix} -\frac{1}{2} & -\frac{1}{4} & 1 \\ 0 & 0 & 0 \end{pmatrix} + c_{2} \begin{pmatrix} 0 & 0 & 0 \\ -\frac{1}{2} & -\frac{1}{4} & 1 \end{pmatrix} \quad (1.4.5.17)$$

Now,

$$CA = \begin{pmatrix} 1 & 1 & 0 \\ -4 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 2 & 2 \\ 1 & 0 \end{pmatrix}$$

$$+ c_1 \begin{pmatrix} -\frac{1}{2} & -\frac{1}{4} & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 2 & 2 \\ 1 & 0 \end{pmatrix}$$

$$+ c_2 \begin{pmatrix} 0 & 0 & 0 \\ -\frac{1}{2} & -\frac{1}{4} & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 2 & 2 \\ 1 & 0 \end{pmatrix}$$

$$\implies CA = \begin{pmatrix} 3 & 1 \\ -4 & 4 \end{pmatrix} + c_1 \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} + c_2 \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\implies CA = B \quad (1.4.5.18)$$

Hence, it is proved that there there exist a

 $n \times k$ matrix. Show that the columns of $\mathbf{C} =$ **AB** are linear combinations of columns of A.If $\alpha_1, \alpha_2, \ldots, \alpha_n$ are the columns of A and $\gamma_1, \gamma_2, \dots, \gamma_k$ are the columns of C then,

$$\gamma_{\mathbf{j}} = \sum_{r=1}^{n} B_{rj} \alpha_{\mathbf{r}}$$
 (1.4.6.1)

Solution:

$$\mathbf{C} = \mathbf{AB} \tag{1.4.6.2}$$

$$\mathbf{C} = \begin{pmatrix} \gamma_1 & \gamma_2 & \dots & \gamma_k \end{pmatrix} \tag{1.4.6.3}$$

$$\mathbf{A} = \begin{pmatrix} \alpha_1 & \alpha_2 & \dots & \alpha_n \end{pmatrix} \tag{1.4.6.4}$$

$$\mathbf{B} = \begin{pmatrix} \beta_1 & \beta_2 & \dots & \beta_k \end{pmatrix} \tag{1.4.6.5}$$

$$= \begin{pmatrix} B_{11} & B_{12} & \dots & B_{1k} \\ B_{21} & B_{22} & \dots & B_{2k} \\ \vdots & \vdots & \dots & \vdots \\ B_{n1} & B_{n2} & \dots & B_{nk} \end{pmatrix}$$
(1.4.6.6)

By matrix multiplication, we can write

$$(\gamma_1 \quad \gamma_2 \quad \dots \quad \gamma_k) = (\mathbf{A}\beta_1 \quad \mathbf{A}\beta_2 \quad \dots \quad \mathbf{A}\beta_k)$$

$$(1.4.6.7)$$

Consider γ_1

$$\gamma_{1} = \mathbf{A}\beta_{1} \qquad (1.4.6.8)$$

$$= \left(\alpha_{1} \quad \alpha_{2} \quad \dots \quad \alpha_{n}\right) \begin{bmatrix} B_{11} \\ B_{21} \\ \vdots \\ B_{n} \end{bmatrix} \qquad (1.4.6.9)$$

$$= B_{11}\alpha_1 + B_{21}\alpha_2 + \ldots + B_{n1}\alpha_n \qquad (1.4.6.10) \ 1.4.8. \ L$$

Similarly, considering j^{th} column of C

$$\gamma_{\mathbf{j}} = \begin{pmatrix} \alpha_{1} & \alpha_{2} & \dots & \alpha_{\mathbf{n}} \end{pmatrix} \begin{pmatrix} B_{1j} \\ B_{2j} \\ \vdots \\ B_{nj} \end{pmatrix}$$
(1.4.6.11)

$$= B_{1j}\alpha_1 + B_{2j}\alpha_2 + \ldots + B_{nj}\alpha_n \qquad (1.4.6.12)$$

$$\implies \gamma_{\mathbf{j}} = \sum_{r=1}^{n} B_{rj} \alpha_{\mathbf{r}} \qquad (1.4.6.13)$$

which proves that columns of C are linear combinations of columns of A

1.4.7. Let **A** and **B** be $n \times n$ matrices such that $\mathbf{AB} = \mathbf{I}$. Prove that $\mathbf{BA} = \mathbf{I}$. Solution: Let $\mathbf{BX} = 0$ be a system of linear equation with n unknowns and n equations as **B** is $n \times n$ matrix. Hence,

$$\mathbf{BX} = 0 \tag{1.4.7.1}$$

$$\implies \mathbf{A}(\mathbf{BX}) = 0 \tag{1.4.7.2}$$

$$\implies (\mathbf{A}\mathbf{B})\mathbf{X} = 0 \tag{1.4.7.3}$$

$$\implies$$
 IX = 0 [:: **AB** = **I**] (1.4.7.4)

$$\implies \mathbf{X} = 0 \tag{1.4.7.5}$$

From (1.4.7.5) since $\mathbf{X} = 0$ is the only solution of (1.4.7.1), hence $rank(\mathbf{B}) = n$. Which implies all columns of \mathbf{B} are linearly independent. Hence \mathbf{B} is invertible. Therefore, every left inverse of \mathbf{B} is also a right inverse of \mathbf{B} . Hence there exists a $n \times n$ matrix \mathbf{C} such that,

$$BC = CB = I$$
 (1.4.7.6)

Again given that AB = I. Hence,

$$\mathbf{AB} = \mathbf{I} \tag{1.4.7.7}$$

$$\implies ABC = C \tag{1.4.7.8}$$

$$\implies \mathbf{A}(\mathbf{BC}) = \mathbf{C} \tag{1.4.7.9}$$

$$\implies$$
 A = **C** [: **BC** = **I**] (1.4.7.10)

Hence using (1.4.7.10) and (1.4.7.6) we can write,

$$\mathbf{BA} = \mathbf{I} \tag{1.4.7.11}$$

Hence Proved.

.8. Let,

$$\mathbf{C} = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix} \tag{1.4.8.1}$$

be a 2×2 matrix. We inquire when it is possible to find 2×2 matrices **A** and **B** such that C=AB-BA. Prove that such matrices can be found if and only if $C_{11}+C_{22}=0$. Solution: We have to find,

$$tr(\mathbf{C}) = C_{11} + C_{22} = tr(\mathbf{AB} - \mathbf{BA})$$
 (1.4.8.2)

$$\implies tr(\mathbf{C}) = tr(\mathbf{AB}) - tr(\mathbf{BA}) \quad (1.4.8.3)$$

We know that,

$$tr(\mathbf{AB}) = \sum_{i=1}^{2} (\mathbf{AB})_{ii}$$
 (1.4.8.4)

$$\implies \sum_{i=1}^{2} \sum_{j=1}^{2} a_{ij} b_{ji} \qquad (1.4.8.5)$$

$$\implies \sum_{i=1}^{2} \sum_{j=1}^{2} b_{ji} a_{ij} \qquad (1.4.8.6)$$

$$\implies tr(\mathbf{AB}) = \sum_{i=1}^{2} \mathbf{BA}_{ij} \qquad (1.4.8.7)$$

$$\implies tr(\mathbf{AB}) = tr(\mathbf{BA})$$
 (1.4.8.8)

Substituting equation (1.4.8.8) to (1.4.8.3) we get

$$\implies tr(\mathbf{C}) = tr(\mathbf{AB}) - tr(\mathbf{BA}) = 0 \quad (1.4.8.9)$$

1.5 Invertible Matrices

1.5.1. Let

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 1 & 0 \\ -1 & 0 & 3 & 5 \\ 1 & -2 & 1 & 1 \end{pmatrix} \tag{1.5.1.1}$$

Find a row-reduced echelon matrix **R** which is row-equivalent to A and an invertible 3x3 matrix P such that R = P A. Solution: Given

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 1 & 0 \\ -1 & 0 & 3 & 5 \\ 1 & -2 & 1 & 1 \end{pmatrix} \tag{1.5.1.2}$$

Row reduce A by applying the elementary row operations and equivalently at each operations find the elementary matrix E

$$\mathbf{A}|\mathbf{I} = \begin{pmatrix} 1 & 2 & 1 & 0 & 1 & 0 & 0 \\ -1 & 0 & 3 & 5 & 0 & 1 & 0 \\ 1 & -2 & 1 & 1 & 0 & 0 & 1 \end{pmatrix} (1.5.1.3)$$

$$\stackrel{R_2=R_2+R_1}{\longleftrightarrow} \begin{pmatrix} 1 & 2 & 1 & 0 & | & 1 & 0 & 0 \\ 0 & 2 & 4 & 5 & | & 1 & 1 & 0 \\ 1 & -2 & 1 & 1 & | & 0 & 0 & 1 \end{pmatrix} (1.5.1.4)$$

$$\xrightarrow{R_3 = R_3 - R_1} \begin{pmatrix} 1 & 2 & 1 & 0 & | & 1 & 0 & 0 \\ 0 & 2 & 4 & 5 & | & 1 & 1 & 0 \\ 0 & -4 & 0 & 1 & | & -1 & 0 & 1 \end{pmatrix}$$

$$(1.5.1.5)$$

$$\xrightarrow{R_1 = R_1 - R_2} \begin{pmatrix} 1 & 0 & -3 & -5 & 0 & -1 & 0 \\ 0 & 2 & 4 & 5 & 1 & 1 & 0 \\ 0 & -4 & 0 & 1 & -1 & 0 & 1 \end{pmatrix}$$

$$(1.5.1.6)$$

$$\xrightarrow{R_3 = R_3 + 2R_2} \begin{pmatrix} 1 & 0 & -3 & -5 & | & 0 & -1 & 0 \\ 0 & 2 & 4 & 5 & | & 1 & 1 & 0 \\ 0 & 0 & 8 & 11 & | & 1 & 2 & 1 \end{pmatrix}$$

$$(1.5.1.7)$$

$$\stackrel{R_2 = \frac{R_2}{2}}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -3 & -5 & 0 & -1 & 0 \\
0 & 1 & 2 & \frac{5}{2} & \frac{1}{2} & \frac{1}{2} & 0 \\
0 & 0 & 8 & 11 & 1 & 2 & 1
\end{pmatrix}$$
(1.5.1.8)

$$\stackrel{R_3 = \frac{R_3}{8}}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -3 & -5 & 0 & 0 & -1 & 0 \\
0 & 1 & 2 & \frac{5}{2} & \frac{1}{2} & \frac{1}{2} & 0 & 0 \\
0 & 0 & 1 & \frac{11}{8} & \frac{1}{8} & \frac{1}{4} & \frac{1}{8}
\end{pmatrix} (1.5.1.9)$$

$$\xrightarrow{R_1 = R_1 + 3R_3} \begin{pmatrix} 1 & 0 & 0 & -\frac{7}{8} \\ 0 & 1 & 2 & \frac{5}{2} \\ 0 & 0 & 1 & \frac{11}{8} \end{pmatrix} \begin{vmatrix} \frac{3}{8} & -\frac{1}{4} & \frac{3}{8} \\ \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{8} & \frac{1}{4} & \frac{1}{8} \end{pmatrix}$$

$$(1.5.1.10)$$

$$\xrightarrow{R_2 = R_2 - 2R_3} \begin{pmatrix} 1 & 0 & 0 & -\frac{7}{8} & | & \frac{3}{8} & -\frac{1}{4} & \frac{3}{8} \\ 0 & 1 & 0 & -\frac{1}{4} & | & \frac{1}{4} & 0 & -\frac{1}{4} \\ 0 & 0 & 1 & \frac{11}{8} & | & \frac{1}{8} & \frac{1}{4} & \frac{1}{8} \end{pmatrix}$$

$$(1.5.1.11)$$

Hence, row reduced echelon matrix that is row equivalent to A is

$$\mathbf{R} = \begin{pmatrix} 1 & 0 & 0 & -\frac{7}{8} \\ 0 & 1 & 0 & -\frac{1}{4} \\ 0 & 0 & 1 & \frac{11}{8} \end{pmatrix}$$
 (1.5.1.12)

where E is the elementary matrices that transform A to R Thus:-

$$\mathbf{EA} = \mathbf{R} \tag{1.5.1.13}$$

Since elementary matrices is invertible

$$\mathbf{P} = \mathbf{E} \tag{1.5.1.14}$$

is invertible.

From (1.5.1.11)

$$\mathbf{P} = \begin{pmatrix} \frac{3}{8} & -\frac{1}{4} & \frac{3}{8} \\ \frac{1}{4} & 0 & -\frac{1}{4} \\ \frac{1}{8} & \frac{1}{4} & \frac{1}{8} \end{pmatrix}$$
 (1.5.1.15)

$$\mathbf{R} = \begin{pmatrix} 1 & 0 & 0 & -\frac{7}{8} \\ 0 & 1 & 0 & -\frac{1}{4} \\ 0 & 0 & 1 & \frac{11}{9} \end{pmatrix}$$
 (1.5.1.16)

such that
$$\mathbf{R} = \mathbf{PA}$$
.
1.5.2. Let $\mathbf{A} = \begin{pmatrix} 2 & 0 & i \\ 1 & -3 & -i \\ i & 1 & 1 \end{pmatrix}$, find a row-reduced applied matrix \mathbf{P} , which is row equivalent to

echelon matrix R which is row-equivalent to **A** and an invertible 3x3 matrix **P** such that **R** = P A. Solution: Given.

$$\mathbf{A} = \begin{pmatrix} 2 & 0 & i \\ 1 & -3 & -i \\ i & 1 & 1 \end{pmatrix} \tag{1.5.2.1}$$

Row reduce A by applying the elementary row operations and equivalently at each operations find the elementary matrix E

$$[\mathbf{A} \ \mathbf{I}] = \begin{pmatrix} 2 & 0 & i & | & 1 & 0 & 0 \\ 1 & -3 & -i & | & 0 & 1 & 0 \\ i & 1 & 1 & | & 0 & 0 & 1 \end{pmatrix} \quad (1.5.2.2)$$

$$\stackrel{R_1 \leftrightarrow R_2}{\longleftrightarrow} \begin{pmatrix} 1 & -3 & -i & | & 0 & 1 & 0 \\ 2 & 0 & i & | & 1 & 0 & 0 \\ i & 1 & 1 & | & 0 & 0 & 1 \end{pmatrix}$$
(1.5.2.3)

$$\xrightarrow{R_2 \leftarrow R_2 - 2R_1} \begin{pmatrix} 1 & -3 & -i & | & 0 & 1 & 0 \\ 0 & 6 & 3i & | & 1 & -2 & 0 \\ i & 1 & 1 & | & 0 & 0 & 1 \end{pmatrix}$$

$$(1.5.2.4)$$

$$\xrightarrow{R_3 \leftarrow R_3 - iR_1} \begin{pmatrix} 1 & -3 & -i & | & 0 & 1 & 0 \\ 0 & 6 & 3i & | & 1 & -2 & 0 \\ 0 & 1 + 3i & 0 & | & 0 & -i & 1 \end{pmatrix}$$

$$(1.5.2.5)$$

$$\stackrel{R_2 \leftarrow \frac{R_2}{6}}{\longleftrightarrow} \begin{pmatrix} 1 & -3 & -i & | & 0 & 1 & 0 \\ 0 & 1 & \frac{i}{2} & | & \frac{1}{6} & -\frac{1}{3} & 0 \\ 0 & 1 + 3i & 0 & | & 0 & -i & 1 \end{pmatrix}$$
(1.5.2.6)

$$\xrightarrow{R_1 \leftarrow R_1 + 3R_2} \begin{pmatrix} 1 & 0 & \frac{i}{2} & | & \frac{1}{2} & 0 & 0 \\ 0 & 1 & \frac{i}{2} & | & \frac{1}{6} & -\frac{1}{3} & 0 \\ 0 & 1 + 3i & 0 & | & 0 & -i & 1 \end{pmatrix}$$

$$(1.5.2.7)$$

$$\xrightarrow{R_3 \leftarrow R_3/(3-i)/2} \begin{pmatrix} 1 & 0 & \frac{i}{2} & | & \frac{1}{2} & 0 & 0\\ 0 & 1 & \frac{7}{2} & | & \frac{1}{6} & -\frac{1}{3} & 0\\ 0 & 0 & 1 & | & -\frac{(i)}{3} & \frac{3+i}{15} & \frac{3+i}{5} \end{pmatrix}$$

$$(1.5.2.8)$$

$$\stackrel{R_1 \leftarrow R_1 - \frac{i}{2}R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & | & \frac{1}{3} & \frac{1-3i}{30} & \frac{1-3i}{10} \\
0 & 1 & \frac{i}{2} & | & \frac{1}{6} & -\frac{1}{3} & 0 \\
0 & 0 & 1 & | & -\frac{(i)}{3} & \frac{3+i}{15} & \frac{3+i}{5}
\end{pmatrix} (1.5.2.9)$$

$$\stackrel{R_2 \leftarrow R_2 - \frac{i}{2}R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & | & \frac{1}{3} & \frac{1-3i}{30} & \frac{1-3i}{10} \\
0 & 1 & 0 & | & 0 & -\frac{3+i}{10} & \frac{1-3i}{10} \\
0 & 0 & 1 & | & -\frac{i}{3} & \frac{3+i}{15} & \frac{3+i}{5}
\end{pmatrix}$$
(1.5.2.10

$$= [I E]$$

Hence, the row reduced matrix that is row equivalent to A is

$$\mathbf{R} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \mathbf{I} \tag{1.5.2.11}$$

Using Gauss-Jordan Elimination, if there exists an elimentary matrix \mathbf{E} such that $\mathbf{E}[\mathbf{A}\ \mathbf{I}] = [\mathbf{I}\ \mathbf{E}]$ then \mathbf{E} is the inverse of \mathbf{A} i.e

$$\mathbf{E} = \mathbf{A}^{-1}$$

$$\mathbf{E} = \mathbf{A}^{-1} = \begin{pmatrix} \frac{1}{3} & \frac{1-3i}{30} & \frac{1-3i}{10} \\ 0 & -\frac{3+i}{10} & \frac{1-3i}{10} \\ -\frac{i}{2} & \frac{3+i}{3} & \frac{3+i}{3} \end{pmatrix}$$
(1.5.2.12)

Since,

$$\mathbf{R} = \mathbf{P}\mathbf{A} \implies \mathbf{P} = \mathbf{A}^{-1}\mathbf{R} \qquad (1.5.2.13)$$

$$\mathbf{P} = \begin{pmatrix} \frac{1}{3} & \frac{1-3i}{30} & \frac{1-3i}{10} \\ 0 & -\frac{3+i}{10} & \frac{1-3i}{10} \\ -\frac{i}{3} & \frac{3+i}{15} & \frac{3+i}{5} \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
 (1.5.2.14)

Thus,

$$\mathbf{P} = \begin{pmatrix} \frac{1}{3} & \frac{1-3i}{30} & \frac{1-3i}{10} \\ 0 & -\frac{3+i}{10} & \frac{1-3i}{10} \\ -\frac{i}{3} & \frac{3+i}{15} & \frac{3+i}{5} \end{pmatrix}$$

$$\mathbf{R} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

1.5.3. For each of the two matrices use elementary row operations to discover whether it is invertible, and to find the inverse in case it is invertible.

$$\mathbf{A} = \begin{pmatrix} 2 & 5 & -1 \\ 4 & -1 & 2 \\ 6 & 4 & 1 \end{pmatrix}, \mathbf{B} = \begin{pmatrix} 1 & -1 & 2 \\ 3 & 2 & 4 \\ 0 & 1 & -2 \end{pmatrix}$$

Solution: Given

$$\mathbf{A} = \begin{pmatrix} 2 & 5 & -1 \\ 4 & -1 & 2 \\ 6 & 4 & 1 \end{pmatrix}, \mathbf{B} = \begin{pmatrix} 1 & -1 & 2 \\ 3 & 2 & 4 \\ 0 & 1 & -2 \end{pmatrix}$$
(1.5.3.1)

By applying row reductions on A

$$\begin{pmatrix} 2 & 5 & -1 \\ 4 & -1 & 2 \\ 6 & 4 & 1 \end{pmatrix} \xrightarrow{R_2 = R_2 - 2R_1} \mathbf{A} = \begin{pmatrix} 2 & 5 & -1 \\ 0 & -11 & 4 \\ 6 & 4 & 1 \end{pmatrix}$$
(1.5.3.2)

$$\stackrel{R_3 = R_3 - 3R_1}{\longleftrightarrow} \begin{pmatrix} 2 & 5 & -1 \\ 0 & -11 & 4 \\ 0 & -11 & 4 \end{pmatrix}$$
(1.5.3.3)

$$\stackrel{R_1 = \frac{R_1}{2}}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{5}{2} & \frac{-1}{2} \\ 0 & -11 & 4 \\ 0 & -11 & 4 \end{pmatrix}$$
(1.5.3.4)

$$\stackrel{R_3=R_3-R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{5}{2} & \frac{-1}{2} \\ 0 & -11 & 4 \\ 0 & 0 & 0 \end{pmatrix}$$
 (1.5.3.5)

$$\stackrel{R_2 = \frac{-R_2}{11}}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{5}{2} & \frac{-1}{2} \\ 0 & 1 & \frac{-4}{11} \\ 0 & 0 & 0 \end{pmatrix}$$
(1.5.3.6)

$$\stackrel{R_1 = R_1 - \frac{5}{2}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{9}{22} \\ 0 & 1 & \frac{-1}{11} \\ 0 & 0 & 0 \end{pmatrix} \tag{1.5.3.7}$$
 1.5.4. Let

For a matrix to be invertible, it has to be a matrix of full rank. However the matrix A is not of full rank (Rank(A) < 3). Therefore A is not invertible.

Let us now consider augmented matrix B|I, By applying row reductions on B|I

$$\begin{pmatrix}
1 & -1 & 2 & 1 & 0 & 0 \\
3 & 2 & 4 & 0 & 1 & 0 \\
0 & 1 & -2 & 0 & 0 & 1
\end{pmatrix}
\xrightarrow{R_2 = R_2 - 3R_1}$$

$$\begin{pmatrix}
1 & -1 & 2 & 1 & 0 & 0 \\
0 & 5 & -2 & -3 & 1 & 0 \\
0 & 1 & -2 & 0 & 0 & 1
\end{pmatrix}$$
(1.5.3.8)

$$\stackrel{R_2 = \frac{R_2}{5}}{\longleftrightarrow} \begin{pmatrix} 1 & -1 & 2 & 1 & 0 & 0 \\ 0 & 1 & \frac{-2}{5} & \frac{1}{5} & \frac{1}{5} & 0 \\ 0 & 1 & -2 & 0 & 0 & 1 \end{pmatrix}$$
(1.5.3.9)

$$\stackrel{R_1=R_1+R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{8}{5} & \frac{2}{5} & \frac{1}{5} & 0\\ 0 & 1 & \frac{-2}{5} & \frac{3}{5} & \frac{1}{5} & 0\\ 0 & 1 & -2 & 0 & 0 & 1 \end{pmatrix} (1.5.3.10)$$

$$\stackrel{R_3=R_3-R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & \frac{8}{5} & \frac{2}{5} & \frac{1}{5} & 0 \\
0 & 1 & \frac{-2}{5} & \frac{-3}{5} & \frac{1}{5} & 0 \\
0 & 0 & \frac{-8}{5} & \frac{3}{5} & \frac{-1}{5} & 1
\end{pmatrix}$$
(1.5.3.11)

$$\stackrel{R_1=R_1+R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & | & 1 & 0 & 1 \\
0 & 1 & \frac{-2}{5} & | & \frac{-3}{5} & \frac{1}{5} & 0 \\
0 & 0 & \frac{-8}{5} & | & \frac{3}{5} & \frac{-1}{5} & 1
\end{pmatrix}$$
(1.5.3.12)

$$\stackrel{R_3 = \frac{-5}{8}R_3}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & 1 & 0 & 1 \\ 0 & 1 & \frac{-2}{5} & \frac{-3}{5} & \frac{1}{5} & 0 \\ 0 & 0 & 1 & \frac{-3}{8} & \frac{1}{8} & \frac{-5}{8} \end{pmatrix} (1.5.3.13)$$

$$\stackrel{R_3=R_2+\frac{2}{5}R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & 1 & 0 & 1 \\
0 & 1 & 0 & \frac{-3}{4} & \frac{1}{4} & \frac{-1}{4} \\
0 & 0 & 1 & \frac{-3}{8} & \frac{1}{8} & \frac{-5}{8}
\end{pmatrix}$$
(1.5.3.14)

For a matrix to be invertible, it has to be a matrix of full rank. Here, the matrix \mathbf{B} is of full

rank ($Rank(\mathbf{B}) = 3$). Therefore **B** is invertible and the inverse matrix \mathbf{B}^{-1} can be written from (1.5.3.14):

$$\mathbf{B}^{-1} = \begin{pmatrix} 1 & 0 & 1\\ \frac{-3}{4} & \frac{1}{4} & \frac{-1}{4}\\ \frac{-3}{8} & \frac{1}{8} & \frac{-5}{8} \end{pmatrix}$$
 (1.5.3.15)

$$\mathbf{A} = \begin{pmatrix} 5 & 0 & 0 \\ 1 & 5 & 0 \\ 0 & 1 & 5 \end{pmatrix} \tag{1.5.4.1}$$

For which **X** does there exist a scalar c such that AX = cX

Solution: Given

$$\mathbf{A} = \begin{pmatrix} 5 & 0 & 0 \\ 1 & 5 & 0 \\ 0 & 1 & 5 \end{pmatrix} \tag{1.5.4.2}$$

The given matrix has single eigenvalue as it is the lower triangular matrix and has equal diagonal elements. Hence $c_1 = c_2 = c_3 = 5$. To find the corresponding eigenvector, consider the following

$$(\mathbf{A} - c\mathbf{I})\mathbf{X} = 0 \tag{1.5.4.3}$$

$$\implies \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$
 (1.5.4.4)

Solving the homogeneous system of linear equations by performing rref, we get

$$\begin{pmatrix} 30 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_2 \longleftrightarrow R_1} \begin{pmatrix} 31 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(1.5.4.5)$$

Hence we get,

$$x_1 = 0, x_2 = 0, x_3 = t$$
 (1.5.4.6)

where, x_3 is arbitrary. Therefore,

$$\mathbf{X} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} t \tag{1.5.4.7}$$

Hence, the given matrix has single eigenvector and is not diagonalizable.

1.5.5. Discover whether

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 0 & 2 & 3 & 4 \\ 0 & 0 & 3 & 4 \\ 0 & 0 & 0 & 4 \end{pmatrix} \tag{1.5.5.1}$$

is invertible, and find A^{-1} if it exists.

Solution: The matrix A is in row reduced echolon form with four pivot elements. Therefore the rank(A) is 4. Hence the rows of matrix A constitute of 4 linearly independent vectors. Thus it can be concluded that matrix A is invertible. Using Gauss-Jordan Elimination, if there exists an elimentary matrix E such that $E[A\ I] = [I\ E]$ then E is the inverse of A i.e $E = A^{-1}$.

$$[\mathbf{A} \ \mathbf{I}] = [\mathbf{I} \ \mathbf{E}] \text{ then } \mathbf{E} \text{ is the inverse of } \mathbf{A} \text{ i.e}$$

$$= \mathbf{A}^{-1}.$$

$$[\mathbf{A} \ \mathbf{I}] = \begin{pmatrix} 1 & 2 & 3 & 4 & | & 1 & 0 & 0 & 0 \\ 0 & 2 & 3 & 4 & | & 0 & 1 & 0 & 0 \\ 0 & 0 & 3 & 4 & | & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 4 & | & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$(1.5.5.2)$$

$$\stackrel{R_1 \leftarrow R_1 - R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & 0 & | & 1 & -1 & 0 & 0 \\ 0 & 2 & 3 & 4 & | & 0 & 1 & 0 & 0 \\ 0 & 0 & 3 & 4 & | & 0 & 0 & 1 & 0 \\ 0 & 0 & 3 & 4 & | & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$(1.5.5.3)$$

$$\stackrel{R_2 \leftarrow R_2 - R_3}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & 0 & | & 1 & -1 & 0 & 0 \\ 0 & 2 & 0 & 0 & | & 0 & 1 & -1 & 0 \\ 0 & 0 & 3 & 4 & | & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$(1.5.5.4)$$

$$\stackrel{R_3 \leftarrow R_3 - R_4}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & 0 & | & 1 & -1 & 0 & 0 \\ 0 & 2 & 0 & 0 & | & 0 & 1 & -1 & 0 \\ 0 & 0 & 3 & 0 & | & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 4 & | & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$(1.5.5.5)$$

$$(1.5.5.5)$$

$$(R_{4} \leftarrow \frac{R_{4}}{4})$$

$$R_{2} \leftarrow \frac{R_{2}}{2} R_{3} \leftarrow \frac{R_{3}}{3}$$

$$(1 \quad 0 \quad 0 \quad 0 \quad | \quad 1 \quad -1 \quad 0 \quad 0)$$

$$(0 \quad 1 \quad 0 \quad 0 \quad | \quad 0 \quad \frac{1}{2} \quad -\frac{1}{2} \quad 0)$$

$$(0 \quad 0 \quad 1 \quad 0 \quad | \quad 0 \quad 0 \quad \frac{1}{3} \quad -\frac{1}{3}$$

$$(0 \quad 0 \quad 0 \quad 1 \quad | \quad 0 \quad 0 \quad 0 \quad \frac{1}{4})$$

$$= [\mathbf{I} \quad \mathbf{E}]$$

$$(1.5.5.6)$$

Therefore, for the given problem,

$$\mathbf{A}^{-1} = \begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & \frac{1}{2} & -\frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{3} & -\frac{1}{3} \\ 0 & 0 & 0 & \frac{1}{4} \end{pmatrix}$$
 (1.5.5.7)

Generalization of above result to a matrix of any arbitrary size: Let

$$\mathbf{A} = \begin{pmatrix} a_1 & a_2 & a_3 & \dots & a_N \\ 0 & a_2 & a_3 & \dots & a_N \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & a_N \end{pmatrix}$$
 (1.5.5.8)

Then

$$\mathbf{E}_{1}\mathbf{A} = \begin{pmatrix} 1 & -1 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & 1 \end{pmatrix} \begin{pmatrix} a_{1} & a_{2} & a_{3} & \dots & a_{N} \\ 0 & a_{2} & a_{3} & \dots & a_{N} \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & a_{N} \end{pmatrix}$$

$$(1.5.5.9)$$

$$= \begin{pmatrix} a_1 & 0 & 0 & \dots & 0 \\ 0 & a_2 & a_3 & \dots & a_N \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & a_N \end{pmatrix}$$
 (1.5.5.10)

$$\mathbf{E}_{2}\mathbf{E}_{1}\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & -1 & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & 1 \end{pmatrix} \begin{pmatrix} a_{1} & 0 & 0 & \dots & 0 \\ 0 & a_{2} & a_{3} & \dots & a_{N} \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & a_{N} \end{pmatrix}$$

$$(1.5.5.11)$$

$$= \begin{pmatrix} a_1 & 0 & 0 & \dots & 0 \\ 0 & a_2 & 0 & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & a_N \end{pmatrix}$$
 (1.5.5.12)

Proceeding in similar manner, we get

$$\mathbf{E}_{N}\mathbf{E}_{N-1}\dots\mathbf{E}_{2}\mathbf{E}_{1}\mathbf{A} = \mathbf{U} = \begin{pmatrix} a_{1} & 0 & 0 & \dots & 0 \\ 0 & a_{2} & 0 & \dots & 0 \\ 0 & 0 & a_{3} & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & a_{N} \end{pmatrix}$$

$$(1.5.5.13)$$

$$= \operatorname{diag} \begin{pmatrix} a_{1} & a_{2} & \dots & a_{N} \end{pmatrix}$$

$$(1.5.5.14)$$

$$\implies A = LU \qquad (1.5.5.15)$$

where
$$\mathbf{L} = \mathbf{E}_1^{-1} \mathbf{E}_2^{-1} \dots \mathbf{E}_N^{-1}$$
 invertible.
 $\Rightarrow \mathbf{A}^{-1} = \mathbf{U}^{-1} \mathbf{L}^{-1}$ invertible.
(1.5.5.16) 1.5.7. Let \mathbf{A} be an $n \times n$ (square) matrix, Prove the following two statements:

$$\Rightarrow \mathbf{A}^{-1} = \begin{pmatrix} \frac{1}{a_1} & 0 & 0 & \dots & 0 \\ 0 & \frac{1}{a_2} & 0 & \dots & 0 \\ 0 & 0 & \frac{1}{a_3} & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & \dots & \dots & \frac{1}{a_N} \end{pmatrix} \begin{pmatrix} 1 & -1 & 0 & \dots & \mathbf{B} \\ 0 & 1 & -1 & \dots & \mathbf{B} \\ 0 & 1 & -1 & \dots & \mathbf{B} \\ 0 & \dots & \dots & \mathbf{B} \\ 0 & \dots & \dots & \mathbf{B} \\ 0 & \dots & \dots & \mathbf{A} \\ 0 & \dots & \dots & \mathbf{B} \\ 0 & \dots &$$

Therefore

$$\mathbf{A}^{-1} = \begin{pmatrix} \frac{1}{a_1} & -\frac{1}{a_1} & 0 & 0 & \dots & 0\\ 0 & \frac{1}{a_2} & -\frac{1}{a_2} & 0 & \dots & 0\\ 0 & 0 & \frac{1}{a_3} & -\frac{1}{a_3} & \dots & 0\\ 0 & 0 & 0 & 0 & \dots & \frac{1}{a_N} \end{pmatrix}$$
(1.5.5.18)

From (1.5.5.18) for the above problem

$$\mathbf{A}^{-1} = \begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & \frac{1}{2} & -\frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{3} & -\frac{1}{3} \\ 0 & 0 & 0 & \frac{1}{4} \end{pmatrix}$$
 (1.5.5.19)

1.5.6. Suppose **A** is a 2×1 matrix and **B** is 1×2 matrix. Prove that **C=AB** is non invertible. **Solution:** Let's take **A** and **B** to be non zero

vectors. Now,we know that for C to be non invertible Cx = 0 should have a non trivial solution.So,

$$\mathbf{C}\mathbf{x} = 0$$
 (1.5.6.1)

$$\implies \mathbf{ABx} = 0 \tag{1.5.6.2}$$

Here, we know that **B** is 1×2 matrix and **x** is 2×1 matrix then **Bx** will result to a scalar constant k.

$$\implies \mathbf{A}k = 0 \tag{1.5.6.3}$$

For (1.5.6.3) to be true k should be zero. We also know that **B** is 1×2 matrix i.e. rows are less than column hence.

$$\mathbf{B}\mathbf{x} = 0$$
 (1.5.6.4)

will have a non trivial solution. Hence, using (1.5.6.3) and (1.5.6.4) we can say,

$$\mathbf{ABx} = 0 \tag{1.5.6.5}$$

will have a non trivial solution so, C is non

- - a) If **A** is invertible and AB = 0 for some $n \times n$

$$\mathbf{AB} = 0 \tag{1.5.7.1}$$

$$\Rightarrow \mathbf{A}^{-1}(\mathbf{AB}) = 0 \tag{1.5.7.2}$$

$$\implies (\mathbf{A}^{-1}\mathbf{A})\mathbf{B} = 0 \tag{1.5.7.3}$$

$$\Rightarrow \mathbf{A}^{-1}(\mathbf{A}\mathbf{B}) = 0 \qquad (1.5.7.2)$$

$$\Rightarrow (\mathbf{A}^{-1}\mathbf{A})\mathbf{B} = 0 \qquad (1.5.7.3)$$

$$\Rightarrow \mathbf{I}\mathbf{B} = 0 \quad [\because \mathbf{A}^{-1}\mathbf{A} = \mathbf{I}] \qquad (1.5.7.4)$$

$$\implies \mathbf{B} = 0 \tag{1.5.7.5}$$

b) If A is not invertible, then there exists an $n \times n$ matrix **B** such that AB = 0 but $B \neq 0$. Since A is not invertible, AX = 0 must have a non-trivial solution. Let the nontrivial solution be,

$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \tag{1.5.7.6}$$

Let **B** which is an $n \times n$ matrix have all its columns as v.

$$\mathbf{B} = \begin{pmatrix} \mathbf{y} & \mathbf{y} & \cdots & \mathbf{y} \end{pmatrix} \tag{1.5.7.7}$$

From equation (1.5.7.7), we can say that $\mathbf{B} \neq$ 0 but $\mathbf{AB} = 0$

1.5.8. Let

$$\mathbf{A} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \tag{1.5.8.1}$$

Prove, using elementary row operations that A is invertible if and only if $(ad - bc) \neq 0$

Solution:

The goal is to effect the transformation $(A|I) \rightarrow$ $(\mathbf{I}|\mathbf{A}^{-1})$. Augmenting **A** with the 2 × 2 identity matrix, we get:

$$\begin{pmatrix} a & b & 1 & 0 \\ c & d & 0 & 1 \end{pmatrix} \tag{1.5.8.2}$$

Now, if a = 0, switch the rows. If c is also

0, then the process of reducing **A** to **I** cannot even begin. So, one necessary condition for **A** to be invertible is that the entries a and c are not both 0.

a) Assume that $a \neq 0$, Then:

$$\begin{pmatrix} a & b & 1 & 0 \\ c & d & 0 & 1 \end{pmatrix} \stackrel{R_1=R_1/a}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{b}{a} & \frac{1}{a} & 0 \\ c & d & 0 & 1 \end{pmatrix}$$

$$(1.5.8.3)$$

$$\stackrel{R_2=R_2-cR_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{b}{a} & \frac{1}{a} & 0 \\ 0 & \frac{ad^a_{-bc}}{c} & \frac{-c}{c} & 1 \end{pmatrix}$$

Next, assuming that $ad - bc \neq 0$, we get:

$$\stackrel{R_1=R_1-\frac{b}{ad-bc}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{d}{ad-bc} & \frac{-b}{ad-bc} \\ 0 & \frac{ad-bc}{a} & \frac{-c}{a} & 1 \end{pmatrix}$$

$$\stackrel{R_2=R_2\frac{a}{ad-bc}}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{d}{ad-bc} & \frac{-b}{ad-bc} \\ 0 & 1 & \frac{-c}{ad-bc} & \frac{a}{ad-bc} \end{pmatrix}$$

Therefore, if $ad - bc \neq 0$, then the matrix is invertible and it's inverse is given by

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$
 (1.5.8.4)

b) In (1.5.8.3), we have assumed that $a \neq 0$. Now consider a = 0, then, as we have seen before, it is mandatory that $c \neq 0$:

$$\begin{pmatrix}
0 & b & 1 & 0 \\
c & d & 0 & 1
\end{pmatrix}
\stackrel{R_1 \leftrightarrow R_2}{\longleftrightarrow} \begin{pmatrix}
c & d & 0 & 1 \\
0 & b & 1 & 0
\end{pmatrix}$$

$$(1.5.8.5)$$

$$\stackrel{R_1 = R_1/c}{\longleftrightarrow} \begin{pmatrix}
1 & \frac{d}{c} & 0 & \frac{1}{c} \\
0 & b & 1 & 0
\end{pmatrix}$$

$$\stackrel{R_1 = R_1 - R_2 \times \frac{d}{bc}}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -\frac{d}{bc} & \frac{1}{c} \\
0 & b & 1 & 0
\end{pmatrix}$$

$$\stackrel{R_2 = R_2/b}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -\frac{d}{bc} & \frac{1}{c} \\
0 & 1 & \frac{1}{b} & 0
\end{pmatrix}$$

Therefore, When we consider a = 0 the matrix is invertible if $bc \neq 0$, which is included in the condition $ad - bc \neq 0$.

c) Similarly, consider c = 0, then, as we have

seen before, it is mandatory that $a \neq 0$:

$$\begin{pmatrix} a & b & 1 & 0 \\ 0 & d & 0 & 1 \end{pmatrix} \stackrel{R_1 = R_1/a}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{b}{a} & \frac{1}{a} & 0 \\ 0 & d & 0 & 1 \end{pmatrix}$$

$$(1.5.8.6)$$

$$\stackrel{R_1 = R_1 - R_2 \times \frac{b}{ad}}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{1}{a} & -\frac{b}{ad} \\ 0 & d & 0 & 1 \end{pmatrix}$$

$$\stackrel{R_2 = R_2/d}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{1}{a} & -\frac{b}{ad} \\ 0 & 1 & 0 & \frac{1}{d} \end{pmatrix}$$

Therefore, When we consider c = 0, the matrix is invertible if $ad \neq 0$, which is included in the condition $ad - bc \neq 0$.

Hence, it is proved from above three cases that the given matrix is invertible iff $ad - bc \neq 0$. 1.5.9. An $n \times n$ matrix **A** is called upper-triangular if $\mathbf{A} = 0$ for i > i that is, if every entry

An $n \times n$ matrix \mathbf{A} is called upper-triangular if $\mathbf{A}_{ij} = 0$ for i > j, that is, if every entry below the main diagonal is 0. Prove that an upper-triangular (square) matrix is invertible if and only if every entry on its main diagonal is different from 0. **Solution:** An $n \times n$ matrix \mathbf{A} is called upper-triangular if $\mathbf{A}_{ij} = 0$ for i > j, that is, if every entry below the main diagonal is 0. Prove that an upper-triangular (square) matrix is invertible if and only if every entry on its main diagonal is different from 0. Considering \mathbf{A} , an upper triangular matrix. Using the property that determinant of upper triangular matrix is the product of diagonal elements,

$$|\mathbf{A}| = \prod_{i=1}^{n} a_{i,i}$$
 (1.5.9.1)

If **A** be invertible then $|\mathbf{A}| \neq 0$. Hence from (1.5.9.1) we get,

$$\prod_{i=1}^{n} a_{i,i} \neq 0 \tag{1.5.9.2}$$

if any diagonal element is 0 then (1.5.9.2) won't be right hence no diagonal elements should be 0. Hence Proved.

1.5.10. Let A be a $m \times n$ matrix. Show that by a finite number of elementary row and/or column operations one can pass from A to a matrix R which is both row-reduced echelon and column-reduced echelon, i.e., $R_{ij} = 0$ if $i \neq j$, $R_{ii} = 1$, $1 \leq i \leq r$, $R_{ii} = 0$, if i > r. Show that R = PAQ, where P is an invertible $m \times m$ matrix and Q is an invertible $n \times n$ matrix.

Solution:

Lemma Every elementary matrix is invertible and the inverse is again an elementary matrix. If an elementary matrix E is obtained from I by using a certain row or column operation q, then E^{-1} is obtained from I by the "inverse" operation q^{-1} .

Solution Given **A** is a $m \times n$ matrix. Converting **A** into row reduced echelon form by performing a series of elementary row operations **P**. Let **R**' be the row reduced echelon matrix. Also, by using the lemma we can tell that **P** is invertible and order $m \times m$.

$$\mathbf{R}' = \mathbf{PA} \tag{1.5.10.1}$$

where,

$$R' = \begin{pmatrix} I & F \\ 0 & 0 \end{pmatrix}$$

I is an identity matrix, F is Free variables matrix and 0 represents a block of zeroes

 ${f R}'$ is in row-reduced echelon form. To perform column operations, elementary matrices should be multiplied on the right side in order to convert the ${f R}'$ into column-reduced echelon form

$$\mathbf{R} = \mathbf{R}'\mathbf{Q} \tag{1.5.10.2}$$

But performing column operations on a matrix is equivalent to performing row operations on the transposed matrix.

$$\mathbf{R}^{T} = (\mathbf{R}'\mathbf{Q})^{T}$$

$$\implies \mathbf{R}^{T} = \mathbf{Q}^{T}\mathbf{R}'^{T} \qquad (1.5.10.3)$$

Hence, by using lemma it can be observed that \mathbf{Q}^T is invertible and of the order $n \times n$. Converting \mathbf{R}^T to row-reduced echelon is equivalent to converting \mathbf{R} to column-reduced echelon.

$$\mathbf{R} = \mathbf{PAQ} \tag{1.5.10.4}$$

where,

$$\mathbf{R} = \begin{pmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{pmatrix} \tag{1.5.10.5}$$

I is an identity matrix and 0 represents a block

of zeroes. \mathbf{Q} is a upper triangular matrix. \mathbf{R} in (1.5.10.4) is in both row and column reduced echelon form. Hence proved. Example Let,

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 2 & 4 & 5 & 7 \\ 1 & 2 & 3 & 4 \end{pmatrix} \tag{1.5.10.6}$$

To convert (1.5.10.6) into row reduced echelon form, **A** has to be multiplied by **P**

$$\mathbf{P} = \begin{pmatrix} -5 & 3 & 0 \\ 2 & -1 & 0 \\ -1 & 0 & 1 \end{pmatrix} \tag{1.5.10.7}$$

$$\mathbf{R}' = \mathbf{P}\mathbf{A} = \begin{pmatrix} 1 & 2 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$
 (1.5.10.8)

 \mathbf{R}' is in row reduced echelon form. To convert (1.5.10.8) into column-reduced echelon form, elementary operations have to be performed on \mathbf{R}'^T . By multiplying all the elementary matrices,

$$\mathbf{Q}^T = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -2 & 1 & 0 & 0 \\ -1 & 0 & -1 & 1 \end{pmatrix} \tag{1.5.10.9}$$

$$\implies \mathbf{Q} = \begin{pmatrix} 1 & 0 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$
 (1.5.10.10)

So **PAQ** is in both row-reduced and column-reduced echelon form.

$$\mathbf{R} = \mathbf{PAQ} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \tag{1.5.10.11}$$

The inverses of **P** and **Q** are,

$$\mathbf{P}^{-1} = \begin{pmatrix} 1 & 3 & 0 \\ 2 & 5 & 0 \\ 1 & 3 & 1 \end{pmatrix}; \quad \mathbf{Q}^{-1} = \begin{pmatrix} 1 & 2 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$
(1.5.10.12)

1.5.11. Prove that the given matrix is invertible and

 A^{-1} has integer values.

$$\mathbf{A} = \begin{pmatrix} 1 & \frac{1}{2} & \cdot & \cdot & \frac{1}{n} \\ \frac{1}{2} & \frac{1}{3} & \cdot & \cdot & \frac{1}{n+1} \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \frac{1}{n} & \frac{1}{n+1} & \cdot & \cdot & \frac{1}{2n-1} \end{pmatrix}$$
 (1.5.11.1)

Solution: Proof that A is Invertible.

The elements of the given matrix is of the form:

$$A_{ij} = \frac{1}{i+j-1} = \int_0^1 t^{i+j-2} dt = \int_0^1 t^{i-1} t^{j-1} dt$$
(1.5.11.2)

We will prove that the matrix **A** is positive definite:

A is Positive definite, if $\mathbf{X}\mathbf{A}\mathbf{X}^T > 0$

Let
$$\mathbf{X} = (x_i)_{1 \le i \le n} \in \mathbb{R}^N$$
 (1.5.11.3)

$$\mathbf{XAX}^{T} = \sum_{1 \le i, j \le n} \frac{x_i x_j}{i + j - 1}$$
 (1.5.11.4)

From (1.5.11.2),

$$\mathbf{XAX}^{T} = \sum_{1 \le i, j \le n} x_{i} x_{j} \int_{0}^{1} t^{i+j-2} dt \qquad (1.5.11.5)$$
$$= \int_{0}^{1} \left(\sum_{i=1}^{n} x_{i} t^{i-1} \right) \left(\sum_{j=1}^{n} x_{j} t^{j-1} \right) dt \qquad (1.5.11.6)$$

$$\implies \mathbf{XAX}^{T} = \int_{0}^{1} \left(\sum_{i=1}^{n} x_{i} t^{i-1} \right)^{2} dt > 0$$
(1.5.11.7)

Thus, Matrix A is Positive definite.

Now, let's say λ is an eigen value of **A**. Then, for the corresponding eigen vector **X** = $(x_1, x_2, ..., x_n)$, we can write:

$$\mathbf{X}\mathbf{A}\mathbf{X}^{T} = \mathbf{X}\lambda\mathbf{X}^{T} \quad [\because \mathbf{A}\mathbf{X}^{T} = \lambda\mathbf{X}^{T}] \quad (1.5.11.8)$$

$$= ||\mathbf{X}||^2 \lambda \tag{1.5.11.9}$$

$$\implies \lambda = \frac{\mathbf{X}\mathbf{A}\mathbf{X}^T}{\|\mathbf{X}\|^2} > 0 \tag{1.5.11.10}$$

So, all of the eigenvalues belonging to **A** must be positive. The product of the eigenvalues of a matrix equals the determinant.

$$\left| \therefore \det(\mathbf{A}) > 0 \right| \tag{1.5.11.11}$$

Thus, the given matrix A is non-singular and

its inverse exist (Invertible).

Proof that A^{-1} has integer values.

Let us consider a set of shifted legendre polynomials as follow:

$$P_{i}(x) = \begin{cases} P_{1}(x) = P_{11} \\ P_{2}(x) = P_{21} + P_{22}x \\ P_{3}(x) = P_{31} + P_{32}x + p_{33}x^{2} \\ \vdots \\ P_{n}(x) = P_{n1} + P_{n2}x + P_{n3}x^{2} + \dots + P_{nn}x^{n-1} \end{cases}$$

$$(1.5.11.12)$$

Where, the coefficients P_{ij} are given as:

$$P_{ij} = (-1)^{i+j-1} {j-1 \choose i-1} {i+j-2 \choose i-1} \quad (1.5.11.13)$$

The shifted legendre polynomials are analogous to legendre polynomials, but defined on the interval [0,1] (whereas the interval is [-1,1] for legendre polynomial).

A set of shifted legendre polynomials obey the following orthogonal relationship:

$$\int_{0}^{1} P_{i}(x)P_{j}(x)dx = 0 \text{ for } i \neq j$$

$$(1.5.11.14)$$

$$\int_{0}^{1} P_{i}(x)P_{j}(x)dx = \frac{1}{2i+1} \text{ for } i = j$$

$$(1.5.11.15)$$

Forming a matrix \mathbf{P} whose elements are the coefficients of polynomials in (1.5.11.12)

Forming a matrix PAP^T , the elements of the matrix PAP^T can be written as:

$$\mathbf{PAP}_{ij}^{T} = \sum_{s=1}^{N} \sum_{r=1}^{N} P_{ir} P_{js} A_{rs}$$
 (1.5.11.17)

From (1.5.11.2) with suitable change in variable notations, we can write:

$$A_{rs} = \int_0^1 x^{r-1} x^{s-1} dx \qquad (1.5.11.18)$$

From (1.5.11.17) and (1.5.11.18),

$$\mathbf{PAP}_{ij}^{T} = \int_{0}^{1} \sum_{s=1}^{N} \sum_{r=1}^{N} P_{ir} P_{js} x^{r-1} x^{s-1} dx$$

$$= \int_{0}^{1} \sum_{s=1}^{N} P_{ir} x^{r-1} \sum_{r=1}^{N} P_{js} x^{s-1} dx$$

$$= \int_{0}^{1} P_{i}(x) P_{j}(x) dx \qquad (1.5.11.21)$$

From (1.5.11.15)

$$\mathbf{PAP}_{ij}^{T} = \begin{cases} 0 & i \neq j \\ \frac{1}{2i+1} & i = j \end{cases}$$
 (1.5.11.22)

Thus, Matrix PAP^T is diagonal matrix:

$$\mathbf{PAP}^{T} = \mathbf{D} = \begin{pmatrix} \frac{1}{3} & 0 & \dots & 0 \\ 0 & \frac{1}{5} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \frac{1}{2n+1} \end{pmatrix}$$
 (1.5.11.23)

From (1.5.11.23), the inverse of matrix **A** can be written as:

$$\mathbf{A} = \mathbf{P}^{-1} \mathbf{D} (\mathbf{P}^T)^{-1}$$
 (1.5.11.24)

$$\implies \mathbf{A}^{-1} = \mathbf{P}^T \mathbf{D}^{-1} \mathbf{P} \tag{1.5.11.25}$$

From (1.5.11.13) and (1.5.11.23), It can be clearly observed that the elements of matrix \mathbf{P} , \mathbf{P}^T and \mathbf{D}^{-1} are all integers given as:

$$\mathbf{P}_{ij} = (-1)^{i+j-1} \binom{j-1}{i-1} \binom{i+j-2}{i-1} \quad (1.5.11.26)$$

$$\mathbf{D}_{ij}^{-1} = \frac{1}{\mathbf{D}_{ij}} = \begin{cases} 0 & i \neq j \\ 2i+1 & i=j \end{cases}$$
 (1.5.11.27)

Since, matrix \mathbf{P} , \mathbf{P}^T and \mathbf{D}^{-1} are integer matrices, therefore \mathbf{A}^{-1} is also an integer matrix.

Hence proved.

Solution 2

https://github.com/Arko98/EE5609/blob/ master/Assignment 14

Let A_3 be 3×3 matrix i.e

$$\mathbf{A_3} = \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{3} \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} \end{pmatrix}$$
 (1.5.11.28)

Now we find the inverse of the matrix A_3 as

follows,

$$\begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{3} & 1 & 0 & 0 \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} & 0 & 1 & 0 \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} & 0 & 0 & 1 \end{pmatrix}$$

$$(1.5.11.29)$$

$$\stackrel{R_2=R_2-\frac{1}{2}R_1}{R_3=R_3-\frac{1}{3}R_1} \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{3} & 1 & 0 & 0 \\ 0 & \frac{1}{12} & \frac{1}{12} & -\frac{1}{2} & 1 & 0 \\ 0 & \frac{1}{12} & \frac{4}{45} & -\frac{1}{3} & 0 & 1 \end{pmatrix}$$

$$(1.5.11.30)$$

$$\stackrel{R_3=R_3-R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{3} & 1 & 0 & 0 \\ 0 & \frac{1}{12} & \frac{1}{12} & -\frac{1}{2} & 1 & 0 \\ 0 & 0 & \frac{1}{180} & \frac{1}{6} & -1 & 1 \end{pmatrix}$$

$$(1.5.11.31)$$

$$\stackrel{R_2=12R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{3} & 1 & 0 & 0 \\ 0 & 1 & 1 & -6 & 12 & 0 \\ 0 & 0 & 1 & 30 & -180 & 180 \end{pmatrix}$$

$$\stackrel{R_2=R_2-R_3}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{1}{2} & 0 & -9 & 60 & -60 \\ 0 & 1 & 0 & -36 & 192 & -180 \\ 0 & 0 & 1 & 30 & -180 & 180 \end{pmatrix}$$

$$\stackrel{R_1=R_1-\frac{1}{2}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & 9 & -36 & 30 \\ 0 & 1 & 0 & -36 & 192 & -180 \\ 0 & 0 & 1 & 30 & -180 & 180 \end{pmatrix}$$

Hence we see that A_3 is invertible and the inverse contains integer entries and A_3^{-1} is given by,

$$\mathbf{A_3^{-1}} = \begin{pmatrix} 9 & -36 & 30 \\ -36 & 192 & -180 \\ 30 & -180 & 180 \end{pmatrix} \tag{1.5.11.35}$$

Let, A_4 be 4×4 matrix as follows,

$$\mathbf{A_4} = \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \frac{1}{4} \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} & \frac{1}{5} \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} & \frac{1}{6} \\ \frac{1}{4} & \frac{1}{5} & \frac{1}{6} & \frac{1}{7} \end{pmatrix}$$
 (1.5.11.36)

Now, expressing A_4 using A_3 we get,

$$\mathbf{A_4} = \begin{pmatrix} \mathbf{A_3} & \mathbf{u} \\ \mathbf{u}^{\mathrm{T}} & d \end{pmatrix} \tag{1.5.11.37}$$

where,

$$\mathbf{u} = \begin{pmatrix} \frac{1}{4} \\ \frac{1}{5} \\ \frac{1}{6} \end{pmatrix}$$
 (1.5.11.38)
$$d = \frac{1}{2}$$
 (1.5.11.39)

Now assuming A_4 has an inverse, then from (1.5.11.37), the inverse of A_4 can be written using block matrix inversion,

Block matrix inversion

If a matrix is partitioned into four blocks, it can be inverted blockwise as follows:

$$\mathbf{M}^{-1} = \begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} \text{ then,}$$

$$\mathbf{M}^{-1} = \begin{pmatrix} \mathbf{A}^{-1} + \mathbf{A}^{-1} \mathbf{B} \left(\mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \right)^{-1} \mathbf{C} \mathbf{A}^{-1} & -\mathbf{A}^{-1} \mathbf{B} \left(\mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \right)^{-1} \\ - \left(\mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \right)^{-1} \mathbf{C} \mathbf{A}^{-1} & \left(\mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \right)^{-1} \end{pmatrix}$$

$$(1.5.11.40)$$

$$\therefore \mathbf{A}_{4}^{-1} = \begin{pmatrix} \mathbf{A}_{3}^{-1} + \mathbf{A}_{3}^{-1} \mathbf{u} x_{4}^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A}_{3}^{-1} & -\mathbf{A}_{3}^{-1} \mathbf{u} x_{4}^{-1} \\ -x_{4}^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A}_{3}^{-1} & x_{4}^{-1} \end{pmatrix}$$

$$= x_{4}^{-1} \begin{pmatrix} \mathbf{A}_{3}^{-1} x_{4} + \mathbf{A}_{3}^{-1} \mathbf{u} \mathbf{u}^{\mathsf{T}} \mathbf{A}_{3}^{-1} & -\mathbf{A}_{3}^{-1} \mathbf{u} \\ \mathbf{u}^{\mathsf{T}} \mathbf{A}_{3}^{-1} & 1 \end{pmatrix}$$

$$(1.5.11.42)$$

where,
$$x_4 = d - \mathbf{u}^{\mathrm{T}} \mathbf{A}_3^{-1} \mathbf{u}$$
 (1.5.11.43)

Now, the assumption of A_4 being invertible will hold if and only if A_3 is invertible, which has been proved in (1.5.11.35) and x_4 from (1.5.11.43) is invertible or x_4 is a nonzero scalar. We now prove that x_4 is invertible as follows,

$$x_{4} = \frac{1}{7} - \begin{pmatrix} \frac{1}{4} & \frac{1}{5} & \frac{1}{6} \end{pmatrix} \begin{pmatrix} 9 & -36 & 30 \\ -36 & 192 & -180 \\ 30 & -180 & 180 \end{pmatrix} \begin{pmatrix} \frac{1}{4} \\ \frac{1}{5} \\ \frac{1}{6} \end{pmatrix}$$

$$(1.5.11.44)$$

$$\implies x_{4} = \frac{1}{2800} \qquad (1.5.11.45)$$

Hence, x_4 is a scalar, hence x_4^{-1} exists and is given by,

$$x_4^{-1} = 2800 (1.5.11.46)$$

Hence, A_4 is invertible. Now putting the values

of $\mathbf{A_3^{-1}}$, x_4^{-1} and \mathbf{u} we get,

$$\mathbf{A_3^{-1}} + \mathbf{A_3^{-1}} \mathbf{u} x_4^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A_3^{-1}} = \begin{pmatrix} 16 & -120 & 240 \\ -120 & 1200 & -2700 \\ 240 & -2700 & 6480 \end{pmatrix}$$

$$(1.5.11.47)$$

$$-\mathbf{A_3^{-1}} \mathbf{u} x_4^{-1} = \begin{pmatrix} -140 \\ 1680 \\ -4200 \end{pmatrix} (1.5.11.48)$$

$$x_4^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A_3^{-1}} = \begin{pmatrix} -140 & 1680 & -4200 \\ (1.5.11.49) & (1.5.11.50) \end{pmatrix}$$

Putting values from (1.5.11.47), (1.5.11.48), (1.5.11.49) and (1.5.11.50) into (1.5.11.41) we get,

$$\mathbf{A_4^{-1}} = \begin{pmatrix} 16 & -120 & 240 & -140 \\ -120 & 1200 & -2700 & 1680 \\ 240 & -2700 & 6480 & -4200 \\ -140 & 1680 & -4200 & 2800 \end{pmatrix}$$
(1.5.11.51)

Hence, from (1.5.11.51) we proved that, A_4 is invertible and has integer entries.

By successively repeating this method, we can prove that A_5 , A_6 , A_7 ,.... and so on, are invertible and have integer values. Thus, we can say, A_{n-1} will be invertible with integer entries. Then we can represent A_n as follows,

$$\mathbf{A_n} = \begin{pmatrix} \mathbf{A_{n-1}} & \mathbf{u} \\ \mathbf{u}^{\mathrm{T}} & d \end{pmatrix} \tag{1.5.11.52}$$

where,

$$\mathbf{u} = \begin{pmatrix} \frac{1}{4} \\ \frac{1}{5} \\ \vdots \\ \frac{1}{2n-2} \end{pmatrix}$$
 (1.5.11.53)
$$d = \frac{1}{2n-1}$$
 (1.5.11.54)

Now assuming A_n has an inverse, then from (1.5.11.52), the inverse of A_n can be written

using block matrix inversion as follows,

$$\mathbf{A_{n}}^{-1} = \begin{pmatrix} \mathbf{A_{n-1}}^{-1} + \mathbf{A_{n-1}}^{-1} \mathbf{u} x_{n}^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A_{n-1}}^{-1} & -\mathbf{A_{n-1}}^{-1} \mathbf{u} x_{n}^{-1} \\ -x_{n}^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A_{n-1}}^{-1} & x_{n}^{-1} \end{pmatrix}$$

$$= x_{n}^{-1} \begin{pmatrix} x_{n} \mathbf{A_{n-1}}^{-1} + \mathbf{A_{n-1}}^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A_{n-1}}^{-1} & -\mathbf{A_{n-1}}^{-1} \mathbf{u} \\ -\mathbf{u}^{\mathsf{T}} \mathbf{A_{n-1}}^{-1} & 1 \end{pmatrix}$$

$$= (1.5.11.56)$$

where,

$$x_n = d - \mathbf{u}^{\mathrm{T}} \mathbf{A}_{n-1}^{-1} \mathbf{u}$$
 (1.5.11.57)

Now, the assumption of A_n being invertible will hold if and only if A_{n-1} is invertible, which is intuitively proved and x from (1.5.11.57) is invertible or x_n is a nonzero scalar. We now prove that x_n is invertible as follows,

$$x_{n} = \frac{1}{2n-1} - \begin{pmatrix} \frac{1}{4} & \frac{1}{5} & \dots & \frac{1}{2n-2} \end{pmatrix} \mathbf{A}_{\mathbf{n}-1}^{-1} \begin{pmatrix} \frac{\frac{1}{4}}{\frac{1}{5}} \\ \frac{1}{5} \\ \dots \\ \frac{1}{2n-2} \end{pmatrix}$$
(1.5.11.58)

In equation (1.5.11.58) \mathbf{u} contains no negative or zero entries, again $\mathbf{A}_{\mathbf{n-1}}^{-1}$ has non zero integer entries, hence $\mathbf{u}^{\mathsf{T}}\mathbf{A}_{\mathbf{n-1}}^{-1}\mathbf{u}$ is a non zero scalar. Moreover d is not equal to $\mathbf{u}^{\mathsf{T}}\mathbf{A}_{\mathbf{n-1}}^{-1}\mathbf{u}$ hence in (1.5.11.58) x is non-zero scalar and invertible and hence it has an inverse. Hence $\mathbf{A}_{\mathbf{n}}$ is invertible, proved.

Proof for A $_{n+1}$:

Expressing A_{n+1} using A_n we get:

$$\mathbf{A}_{\mathbf{n+1}} = \begin{pmatrix} \mathbf{A}_{\mathbf{n}} & \mathbf{u} \\ \mathbf{u}^{\mathsf{T}} & d \end{pmatrix} \tag{1.5.11.59}$$

where,

$$\mathbf{u} = \begin{pmatrix} \frac{1}{4} \\ \frac{1}{5} \\ \vdots \\ \frac{1}{2n-1} \end{pmatrix}$$
 (1.5.11.60)
$$d = \frac{1}{-}$$
 (1.5.11.61)

Now assuming A_{n+1} has an inverse, then from (1.5.11.59), the inverse of A_{n+1} can be written

using block matrix inversion as follows,

$$\mathbf{A_{n+1}}^{-1} = \begin{pmatrix} \mathbf{A_n^{-1}} + \mathbf{A_n^{-1}} \mathbf{u} x_{n+1}^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A_n^{-1}} & -\mathbf{A_n^{-1}} \mathbf{u} x_{n+1}^{-1} \\ -x_{n+1}^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A_n^{-1}} & x_{n+1}^{-1} \end{pmatrix}$$

$$= x_{n+1}^{-1} \begin{pmatrix} x_{n+1} \mathbf{A_n^{-1}} + \mathbf{A_n^{-1}} \mathbf{u}^{\mathsf{T}} \mathbf{A_n^{-1}} & -\mathbf{A_n^{-1}} \mathbf{u} \\ -\mathbf{u}^{\mathsf{T}} \mathbf{A_n^{-1}} & 1 \end{pmatrix}$$

$$= (1.5.11.63)$$

where,

$$x_{n+1} = d - \mathbf{u}^{\mathrm{T}} \mathbf{A}_{\mathbf{n}}^{-1} \mathbf{u}$$
 (1.5.11.64)

Now, the assumption of A_{n+1} being invertible will hold if and only if A_n is invertible, which is proved and x from (1.5.11.64) is invertible or x_{n+1} is a nonzero scalar. We now prove that x_{n+1} is invertible as follows,

$$x_{n+1} = \frac{1}{2n} - \begin{pmatrix} \frac{1}{4} & \frac{1}{5} & \dots & \frac{1}{2n-1} \end{pmatrix} \mathbf{A_n^{-1}} \begin{pmatrix} \frac{1}{4} \\ \frac{1}{5} \\ \dots \\ \frac{1}{2n-1} \end{pmatrix}$$
(1.5.11.65)

In equation (1.5.11.65) \mathbf{u} contains no negative or zero entries, again $\mathbf{A}_{\mathbf{n}}^{-1}$ has non zero integer entries, hence $\mathbf{u}^{\mathsf{T}}\mathbf{A}_{\mathbf{n}}^{-1}\mathbf{u}$ is a non zero scalar. Moreover d is not equal to $\mathbf{u}^{\mathsf{T}}\mathbf{A}_{\mathbf{n}}^{-1}\mathbf{u}$ hence in (1.5.11.65) x is non-zero scalar and invertible and hence it has an inverse. Hence $\mathbf{A}_{\mathbf{n}+1}$ is also invertible.

Problem statement: If A_{n-1}^{-1} is invertible and has integer values, Then A_n^{-1} also has integer values.

Proof:

The matrix A_n^{-1} can be expressed as:

$$\mathbf{A_n} = \begin{pmatrix} \mathbf{A_{n-1}} & \mathbf{u} \\ \mathbf{u}^T & d \end{pmatrix} \tag{1.5.11.66}$$

where,

$$\mathbf{u} = \begin{pmatrix} \frac{1}{n} \\ \frac{1}{n+1} \\ \vdots \\ \frac{1}{2n-2} \end{pmatrix}$$
 (1.5.11.67)

$$d = \frac{1}{2n-1} \tag{1.5.11.68}$$

The inverse of A_n can be written using block matrix inversion,

Block matrix inversion

If a matrix is partitioned into four blocks, it can be inverted blockwise as follows:

$$\mathbf{M}^{-1} = \begin{pmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{pmatrix} \text{ then,}$$

$$\mathbf{M}^{-1} = \begin{pmatrix} \mathbf{A}^{-1} + \mathbf{A}^{-1} \mathbf{B} \begin{pmatrix} \mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \end{pmatrix}^{-1} \mathbf{C} \mathbf{A}^{-1} & -\mathbf{A}^{-1} \mathbf{B} \begin{pmatrix} \mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \end{pmatrix}^{-1} \\ - \begin{pmatrix} \mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \end{pmatrix}^{-1} \mathbf{C} \mathbf{A}^{-1} & \begin{pmatrix} \mathbf{D} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} \end{pmatrix}^{-1} \end{pmatrix}$$

$$(1.5.11.69)$$

$$\mathbf{A_{n}}^{-1} = \begin{pmatrix} \mathbf{A_{n-1}^{-1}} + \mathbf{A_{n-1}^{-1}} \mathbf{u} x_{n}^{-1} \mathbf{u}^{\mathrm{T}} \mathbf{A_{n-1}^{-1}} & -\mathbf{A_{n-1}^{-1}} \mathbf{u} x_{n}^{-1} \\ -x_{n}^{-1} \mathbf{u}^{\mathrm{T}} \mathbf{A_{n-1}^{-1}} & x_{n}^{-1} \end{pmatrix}$$

$$(1.5.11.70)$$

where,

$$x_n = d - \mathbf{u}^{\mathrm{T}} \mathbf{A}_{\mathbf{n}-1}^{-1} \mathbf{u}$$
 (1.5.11.71)

For A_n^{-1} to have integer values, each of the four blocks in (1.5.11.70) should have integer values.

Let us first prove that x_n^{-1} have integer values as follow:

$$x_{n} = d - \mathbf{u}^{T} \mathbf{A}_{\mathbf{n}-1}^{-1} \mathbf{u}$$

$$= \frac{1}{2n-1} - \sum_{i}^{n-1} \sum_{j}^{n-1} u_{i} u_{j} (A_{n-1}^{-1})_{ij}$$

$$= \frac{1}{2n-1} - \sum_{i}^{n-1} \sum_{j}^{n-1} \frac{1}{n+i-1} \frac{1}{n+j-1} (A_{n-1}^{-1})_{ij}$$

$$(1.5.11.74)$$

Thus, x_n is a scalar of the form $\frac{p}{q}$. Also, by

calculation, $x_2 = \frac{1}{12}$, $x_3 = \frac{1}{180}$ and $x_3 = \frac{1}{2800}$. Thus, by induction, we can say p = 1 for any value of n. Thus, x_n^{-1} can be written as:

$$x_n^{-1} = \frac{1}{x_n} = \frac{q}{p} \tag{1.5.11.75}$$

Since, p = 1 for any value of n, Hence $x_n^{-1} = q$ is always integer.

Now, let us consider the block $-\mathbf{A}_{n-1}^{-1}\mathbf{u}x_n^{-1}$, we can write:

$$(-\mathbf{A}_{\mathbf{n}-1}^{-1}\mathbf{u}x_{n}^{-1})_{i,1} = -q\sum_{j}^{n-1}\frac{1}{n+j-1}(A_{n-1}^{-1})_{i,j}$$
(1.5.11.76)

Here, the considered block is a $(n-1) \times 1$ matrix, with each element of rational form $\frac{p_1}{a}$. For n = 2, 3, 4 the value of q_1 comes as $\frac{1}{1}$. Thus by induction, the value of q_1 is always 1 for any value of n. Hence, this matrix always has integer values.

Now, let us consider the block $-x_n^{-1}\mathbf{u}^T\mathbf{A}_{n-1}^{-1}$, we can write:

$$(-x_n^{-1}\mathbf{u}^T\mathbf{A}_{\mathbf{n-1}}^{-1})_{1,j} = -q\sum_{i}^{n-1}\frac{1}{n+i-1}(A_{n-1}^{-1})_{i,j}$$
(1.5.11.77)

Here, the considered block is a $1 \times (n-1)$ matrix, with each element of rational form $\frac{p_2}{a}$. For n = 2, 3, 4 the value of q_2 comes as $\frac{q_2}{1}$. Thus by induction, the value of q_2 is always 1 for any value of n. Hence, this matrix always has integer values.

Considering the block $\mathbf{A}_{\mathbf{n-1}}^{-1} + \mathbf{A}_{\mathbf{n-1}}^{-1} \mathbf{u} x_n^{-1} \mathbf{u}^{\mathrm{T}} \mathbf{A}_{\mathbf{n-1}}^{-1}$, let denote it as V_{n-1}^{-1} :

$$V_{n-1}^{-1} = \mathbf{A}_{n-1}^{-1} + \mathbf{A}_{n-1}^{-1} \mathbf{u} x_n^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A}_{n-1}^{-1}$$
 (1.5.11.78)
= $\mathbf{A}_{n-1}^{-1} + \mathbf{A}_{n-1}^{-1} \mathbf{u} \left(d - \mathbf{u}^{\mathsf{T}} \mathbf{A}_{n-1}^{-1} \mathbf{u} \right)^{-1} \mathbf{u}^{\mathsf{T}} \mathbf{A}_{n-1}^{-1}$ (1.5.11.79)

The Woodbury matrix identity is given by:

$$\frac{(A + UCV)^{-1} = A^{-1} - A^{-1}U(C^{-1} + VA^{-1}U)^{-1}VA^{2.1}}{(1.5.11.80) 2.1.1. \text{ If } \mathbf{F} \text{ is a field}}$$

Comparing (1.5.11.79) and (1.5.11.80), we can

write:

$$V_{n-1}^{-1} = \left(A_{n-1} - (2n-1)\mathbf{u}\mathbf{u}^{T}\right)^{-1} \qquad (1.5.11.81)$$

$$\therefore V_{n-1} = A_{n-1} - (2n-1)\mathbf{u}\mathbf{u}^{T} \qquad (1.5.11.82)$$

Here, V_{n-1} can be expanded as:

$$V_{n-1} = A_{n-1} - (2n-1)\mathbf{u}\mathbf{u}^{T} \qquad (1.5.11.83)$$

$$= \frac{1}{i+j-1} - \frac{(2n-1)}{(n+i-1)(n+j-1)} \qquad (1.5.11.84)$$

$$= \frac{n^{2} - ni - nj - ij}{(i+j-1)(n+i-1)(n+j-1)} \qquad (1.5.11.85)$$

$$= \frac{(n-i)(n-j)}{(i+j-1)(n+i-1)(n+j-1)} \qquad (1.5.11.86)$$

$$\therefore (V_{n-1})_{ij} = \left((A_{n-1})_{ij} \frac{(n-i)(n-j)}{(n+i-1)(n+j-1)} \right) \qquad (1.5.11.87)$$

Using (1.5.11.87), The inverse V_{n-1}^{-1} can be written as:

$$(V_{n-1}^{-1})_{ij} = \left((A_{n-1}^{-1})_{ij} \frac{(n+i-1)(n+j-1)}{(n-i)(n-j)} \right)$$
(1.5.11.88

Here, the considered block is a $(n-1)\times(n-1)$ matrix, with each element of rational form $\frac{p_3}{q_2}$. For n = 2, 3, 4 the value of q_3 comes as 1. Thus by induction, the value of q_3 is always 1 for any value of n. Hence, this matrix always has integer values.

Since, all the four blocks has integer values, the inverse A_n^{-1} has integer values. **Observations:**

- a) The given matrix is a $n \times n$ Hilbert matrix. Which is always invertible with its inverse having integer values.
- b) The Hilbert matrix is symmetric and positive definite.

2 Vector Spaces

(1.5.11.80) 2.1.1. If **F** is a field, verify that vector space of all ordered n-tuples \mathbf{F}^n is a vector space over the field **F**.

Solution: Let \mathbf{F}^n be a set of all ordered n-tuples

over F i.e

$$\mathbf{F}^{n} = \left\{ \begin{pmatrix} a_{1} \\ a_{2} \\ \vdots \\ a_{n} \end{pmatrix} : a_{1}, a_{2}, \dots, a_{n} \in \mathbf{F} \right\}$$
 (2.1.1.1)

For \mathbf{F}^n to be a vector space over \mathbf{F} it must satisfy the closure property of vector addition and scalar multiplication.

Vector Addition in \mathbf{F}^n :

Let $\alpha = (a_i)$ and $\beta = (b_i) \ \forall \ i = 1, 2, \dots, n \in \mathbf{F}^n$ then

$$\alpha + \beta = (a_i) + (b_i) \tag{2.1.1.2}$$

$$= \left(a_i + b_i\right) \tag{2.1.1.3}$$

Since

$$a_i + b_i \in \mathbf{F} \ \forall \ i = 1, 2, \cdots, n$$
 (2.1.1.4)

$$\implies \alpha + \beta \in \mathbf{F}^n$$
 (2.1.1.5)

Scalar multiplication in F^n over F:

Let $\alpha = (a_i) \ \forall \ i = 1, 2, \dots, n \in \mathbf{F}^n$ and $a \in \mathbf{F}$ then

$$a\alpha = (aa_i) \tag{2.1.1.6}$$

Since

$$aa_i \in \mathbf{F} \ \forall \ i = 1, 2 \cdots, n$$
 (2.1.1.7)

$$\implies a\alpha \in \mathbf{F}^n$$
 (2.1.1.8)

Associativity of addition in \mathbf{F}^n :

Let $\alpha = (a_i)$, $\beta = (b_i)$, $\gamma = (g_i) \forall i =$ $1, 2, \cdots, n \in \mathbf{F}^n$ then

$$\alpha + (\beta + \gamma) = (a_i) + (b_i + g_i)$$
 (2.1.1.9)
= $(a_i + b_i + g_i)$ (2.1.1.10)
= $(a_i + b_i) + (g_i)$ (2.1.1.11)
= $(\alpha + \beta) + \gamma$ (2.1.1.12)

Existence of additive identity in \mathbf{F}^n :

We have
$$\mathbf{0} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \in \mathbf{F}^n \text{ and } \alpha = (a_i) \ \forall \ i = a_i$$

 $1, 2, \cdots, n \in \mathbf{F}^n$ then

$$(a_i) + (0) = (a_i + 0)$$
 (2.1.1.13)
= (a_i) (2.1.1.14)

Therefore $\mathbf{0}$ is the additive identity in \mathbf{F}^n .

Existence of additive inverse of each element of \mathbf{F}^n :

If $\alpha = (a_i) \ \forall \ i = 1, 2, \dots, n \in \mathbb{F}^n$ then $(-a_i) \in \mathbf{F}^n$. Also we have

$$(-a_i) + (a_i) = \mathbf{0} \tag{2.1.1.15}$$

Therefore $-\alpha = (-a_i)$ is the additive inverse of α . Thus \mathbf{F}^n is an abelian group with respect to addition.

Futher we observe that

a) If $a \in \mathbf{F}$ and $\alpha = (a_i)$, $\beta = (b_i) \ \forall \ i = 1, 2, \dots, n \in \mathbf{F}^n$ then

$$a(\alpha + \beta) = a(a_i + b_i)$$
 (2.1.1.16)

$$= (a[a_i + b_i])$$
 (2.1.1.17)
= $(aa_i + ab_i)$ (2.1.1.18)
 $(aa_i) + (ab_i)$ (2.1.1.19)

$$= \left(aa_i + ab_i\right) \tag{2.1.1.18}$$

$$(aa_i) + (ab_i)$$
 (2.1.1.19)

$$= a(a_i) + a(b_i)$$
 (2.1.1.20)

$$= a\alpha + a\beta \tag{2.1.1.21}$$

b) If $a,b \in \mathbf{F}$ and $\alpha = (a_i) \ \forall \ i = 1, 2, \dots, n \in \mathbf{F}^n$ then

$$(a+b)\alpha = ([a+b]a_i)$$
 (2.1.1.22)

$$= \left(aa_i + ba_i\right) \tag{2.1.1.23}$$

$$= (aa_i) + (ba_i) \tag{2.1.1.24}$$

$$= a(a_i) + b(a_i)$$
 (2.1.1.25)

$$= a\alpha + b\alpha \tag{2.1.1.26}$$

c) If $a,b \in \mathbf{F}$ and $\alpha = (a_i) \ \forall \ i = 1, 2, \dots, n \in \mathbf{F}^n$

then

$$(ab)\alpha = ([ab]a_i)$$
 (2.1.1.27)
= $(a[ba_i])$ (2.1.1.28)
= $a(ba_i)$ (2.1.1.29)
= $a(b\alpha)$ (2.1.1.30)

d) If 1 is the unity element of **F** and α = $(a_i) \ \forall \ i = 1, 2, \cdots, n \in \mathbf{F}^n \text{ then}$

$$1\alpha = (1a_i)$$
 (2.1.1.31)
= (a_i) (2.1.1.32)
= α (2.1.1.33)

Hence \mathbf{F}^n is a vector space over \mathbf{F} .

2.1.2. If V is a vector space over field F, verify that:

$$(\alpha_1 + \alpha_2) + (\alpha_3 + \alpha_4) = [\alpha_2 + (\alpha_3 + \alpha_1)] + \alpha_4$$
(2.1.2.1)

Solution: Using property of commutativity of (+) in V

$$(\alpha_1 + \alpha_2) + (\alpha_3 + \alpha_4) = (\alpha_2 + \alpha_1) + (\alpha_3 + \alpha_4)$$
(2.1.2.2) 2.1.4.

Using property of associativity of (+) in V

$$(\alpha_2 + \alpha_1) + (\alpha_3 + \alpha_4) = \alpha_2 + [\alpha_1 + (\alpha_3 + \alpha_4)]$$
(2.1.2.3)

Using property of commutativity of (+) in V

$$\alpha_2 + [\alpha_1 + (\alpha_3 + \alpha_4)] = \alpha_2 + (\alpha_3 + \alpha_1) + \alpha_4$$
 (2.1.2.4)

Using property of associativity of (+) in V

$$\alpha_2 + (\alpha_3 + \alpha_1) + \alpha_4 = [\alpha_2 + (\alpha_3 + \alpha_1)] + \alpha_4$$
(2.1.2.5)

2.1.3. If \mathbb{C} is the field of complex numbers, which vectors in \mathbb{C}^3 are linear combinations of $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$,

$$\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$
 and $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$?

Solution: Expressing the given vectors as the 2.1.5. On \mathbb{R}^n define two operations columns of a matrix.

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ -1 & 1 & 1 \end{pmatrix} \tag{2.1.3.1}$$

The row reduced echelon form of the matrix on performing elementary row operations can be given as,

$$\mathbf{R} = \mathbf{C}\mathbf{A} \tag{2.1.3.2}$$

where C is the product of elementary matrices,

$$\mathbf{C} = \begin{pmatrix} 0 & 1 & -1 \\ -1 & 2 & -1 \\ 1 & -1 & 1 \end{pmatrix} \tag{2.1.3.3}$$

Thus we get,

$$\mathbf{R} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{2.1.3.4}$$

From (2.1.3.4), $rank(\mathbf{A}) = 3$. Thus \mathbf{A} is a full rank matrix. Hence the columns of A are linearly independent i.e., the given vectors are linearly independent and forms the basis for \mathbb{C}^3 .

Hence any vector $\mathbf{Y} \in \mathbf{C}^3$ can be written as the

linear combinations of
$$\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$
, $\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$ and $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$.

(2.1.2.2) 2.1.4. Let V be the set of all pairs (x,y) of real numbers and let F be the field of real numbers. Define

$$(x, y) + (x_1, y_1) = (x + x_1, y + y_1)$$
 (2.1.4.1)
 $c(x, y) = (cx, y)$ (2.1.4.2)

Is V with these operations, a vector space over the field of real numbers?

Solution: $V = \{(x,y) \mid x,y \in R\}$, consider u = $(x_1, y_1) \in V, a, b, c \in R$. Axioms with respect to addition and scalar multiplication.

a)

$$(a+b)u = (a+b)(x_1, y_1)$$
 (2.1.4.3)
= $((a+b)x_1, y_1) \neq au + bu$ (2.1.4.4)

Since V with the given operations the equation (2.1.4.4) contradicts the axioms of scalar multiplication. Hence it is not vector space over real number with these operations.

$$\alpha \oplus \beta = \alpha - \beta \tag{2.1.5.1}$$

$$c \cdot \alpha = -c\alpha \tag{2.1.5.2}$$

The operations on the right are usual ones.

Which of the axioms for a vector space are satisfied by $(\mathbb{R}^n, \oplus, \cdot)$?

Solution: Let $(\alpha, \beta, \gamma) \in \mathbb{R}^n$ and c, c_1, c_2 are scalars taken from the field \mathbb{R} where the vector space is defined on. Table 2.1.5 lists the axioms satisfied and not satisfied for $(\mathbb{R}^n, \oplus, \cdot)$.

2.1.6. Let **V** be the set of pairs (x, y) of real numbers and let **F** be the field of real numbers. Define

$$(x, y) + (x_1, y_1) = (x + x_1, 0)$$
 (2.1.6.1)

$$c(x, y) = (cx, 0)$$
 (2.1.6.2)

Is V, with these operations, a vector space?

Solution: V is a vector space if it satisfies all properties of the vector space. Let us consider the property of Existence of additive identity. According to Existence of additive identity, there is a unique vector $\mathbf{0}$ in \mathbf{V} called the zero vector, such that $\alpha + \mathbf{0} = \alpha$ for all α in \mathbf{V} .

Let $u = (x_1, y_1) \in \mathbf{V}$

$$u + \mathbf{0} = (x_1, y_1) + (0, 0)$$

$$= (x_1 + 0, 0)$$

$$= (x_1, 0)$$

$$\neq u$$
(2.1.6.3)

From (2.1.6.3), there does not exist an additive identity for V.

Hence V is not a vector space.

tions f on the real line such that

$$f(-t) = \overline{f(t)} \tag{2.1.7.1}$$

The bar denotes complex conjugation. Show that V, with the operations

$$(f+g)(t) = f(t) + g(t) (2.1.7.2)$$

$$(cf)(t) = cf(t)$$
 (2.1.7.3)

is a vector space over the field of real numbers. Give an example of a function in V which is not real valued.

Solution: To prove that V with the given operations is a vector space over the field of real numbers, we have to start by proving that additivity and homogeneity both hold true. So, we have to prove that (cf+g)(t) is equal to

cf(t)+g(t).

$$(cf+g)(t)$$
 (2.1.7.4)

$$= (cf)(t) + g(t)$$
 (2.1.7.5)

$$= cf(t) + g(t) (2.1.7.6)$$

Now, we know that $f(-t) = \overline{f(-t)}$ and so (cf+g)(t) should also satisfy the property,

$$(cf + g)(-t)$$
 (2.1.7.7)

$$= cf(-t) + g(-t)$$
 (2.1.7.8)

$$= c\overline{f(t)} + \overline{g(t)} \tag{2.1.7.9}$$

$$= \overline{cf(t) + g(t)} \tag{2.1.7.10}$$

$$= \overline{(cf+g)(t)} \tag{2.1.7.11}$$

Example Let's take f(x)=a+ix

$$f(1) = a + i \tag{2.1.7.12}$$

Hence, f(x) is not real valued. Now,

$$f(x) = a + ix (2.1.7.13)$$

$$f(-x) = a - ix (2.1.7.14)$$

$$f(-x) = \overline{f(x)}$$
 (2.1.7.15)

Since a and $x \in \mathbb{R}$, so $f \in \mathbb{V}$

2.2 Subspaces

2.1.7. Let \mathbb{V} be the set of all complex-valued func- 2.2.1. Which of the following set of vectors

$$\alpha = (a_1, a_2, \dots, a_n)$$

in \mathbb{R}^n are subspace of \mathbb{R}^n $(n \ge 3)$?

- a) All α such that $a_1 \ge 0$
- b) All α such that $a_1 + 3a_2 = a_3$
- c) All α such that $a_2 = a_1^2$
- d) All α such that $a_1a_2 = 0$
- e) All α such that a_2 is rational

Solution: Table 2.2.1 lists the summary of which set of vectors in \mathbb{R}^n are subspace of \mathbb{R}^n (n > 3).

- 2.2.2. Let **V** be the (real) vector space of all functions *f* from **R** into **R**.
 - a) Is f(0) = f(1) a subspace of **V**? **Solution:** A non-empty subset **W** of **V** is a

UNSATISTIFD	SATISFIED
Associativity of addition	Additive identity
$\alpha \oplus (\beta \oplus \gamma) = \alpha - \beta + \gamma$	$\alpha \oplus \beta = \alpha - \beta = \alpha$
$(\alpha \oplus \beta) \oplus \gamma = \alpha - \beta - \gamma$	Additive identity is β
$\alpha \oplus (\beta \oplus \gamma) \neq (\alpha \oplus \beta) \oplus \gamma$	unique $\beta = (0, 0,0)$
Commutativity of addition	Additive inverse
$\alpha \oplus \beta = \alpha - \beta$	$\alpha \oplus \alpha = \alpha - \alpha = 0$
$\beta \oplus \alpha = \beta - \alpha$	Additive inverse is α
$\alpha \oplus \beta \neq \beta \oplus \alpha$	
Scalar multiplication with field multiplication	
$(c_1c_2)\cdot\alpha=(-c_1c_2)\alpha$	
$c_1 \cdot (c_2 \cdot \alpha) = c_1 c_2 \alpha$	
$(c_1c_2)\cdot\alpha\neq c_1\cdot(c_2\cdot\alpha)$	
Identity element of scalar multiplication	
$1 \cdot \alpha = -\alpha = \alpha \text{ for } \alpha = (0, 0,, 0)$	
$1 \cdot \alpha = -\alpha \neq \alpha \forall \alpha \neq (0, 0,, 0)$	
Distributivity of scalar multiplication w.r.t vector addition	
$c \cdot (\alpha \oplus \beta) = -c(\alpha - \beta)$	
$c \cdot \alpha \oplus c \cdot \beta = -c\alpha - (-c\beta)$	
$c \cdot (\alpha \oplus \beta) \neq c \cdot \alpha \oplus c \cdot \beta$	
Distributivity of scalar multiplication w.r.t field addition	
$(c_1 + c_2) \cdot \alpha = -(c_1 + c_2)\alpha$	
$c_1 \cdot \alpha \oplus c_2 \cdot \beta = -c_1 \alpha - (-c_2 \beta)$	
$(c_1 + c_2) \cdot \alpha \neq c_1 \cdot \alpha \oplus c_2 \cdot \beta$	

TABLE 2.1.5: Axioms of vector space $(\mathbb{R}^n, \oplus, \cdot)$

$\alpha = (a_1, a_2, \dots, a_n)$		
Vector space	Subspace summary	
$\alpha = (a_1, a_2, a_3, a_4, \dots, a_n); a_1 \ge 0$	Not a subspace. Scalar multiplication is not satisfied. $-1(\alpha) \neq \alpha$	
$\alpha = (a_1, a_2, a_3, a_4, \dots, a_n); a_1 + 3a_2 =$	a ₃ It is a subspace	
$\alpha = (a_1, a_2, a_3, a_4, \dots, a_n); a_2 = a_1^2$	Not a subspace. Addition is not satisfied. $(a_1 + b_1)^2 \neq a_1^2 + b_1^2$	
$\alpha = (a_1, a_2, a_3, a_4, \dots, a_n); a_1 a_2 = 0$	Not a subspace. Addition is not satisfied. $a_1b_1 \neq 0$	
$\alpha = (a_1, a_2, a_3, a_4, \dots, a_n);$ a_2 is ration	al Not a subspace. Scalar multiplication is not satisfied. $a_2 \neq \sqrt{2}a_1$	

TABLE 2.2.1: Summary

subspace of **V** if and only if for each pair of vectors α β in **W** and each scalar c in **R** the vector $c\alpha + \beta$ is again in **W**. For each of the function to be a subspace, it must be closed with respect to addition and scalar multiplication in **V** defined as, for f g ϵ **W**

and $c \in R$

Then,

$$h = cf + g$$

$$h(0) = cf(0) + g(0)$$

$$= cf(1) + g(1)$$

$$= h(1)$$
(2.2.2.1)
(2.2.2.2)
(2.2.2.3)
(2.2.2.3)
(2.2.2.3)
(2.2.2.3)
(2.2.2.4)
(2.2.2.4)
(2.2.2.4)

Thus, h(0) = h(1). Therefore, **W** is a subset of **V** and also a vector space. Therefore **W** is a subspace of **V**.

Hence, f(0) = f(1) is a subspace of **V**.

b) Let **V** be the (real) vector space of all functions f from **R** into **R**. Is f(3) = 1 + f(-5) a subspace of **V**

Solution: For each of the function to be a subspace, it must be closed with respect to addition and scalar multiplication in V defined as, for f g ϵ W Then,

$$(f+g)(3) = f(3) + g(3) (2.2.2.5)$$

$$= 1 + f(-5) + 1 + g(-5) (2.2.2.6)$$

$$= 2 + f(-5) + g(-5) (2.2.2.7)$$

$$= 2 + (f+g)(-5) (2.2.2.8)$$

$$\neq 1 + (f+g)(-5) (2.2.2.9)$$

Since W is not closed with respect to addition \therefore It is not a subspace of V.

c) Let **V** be the (real) vector space of all functions f from **R** into **R**. Is f(-1) = 0 a subspace of **V**

Solution: A non-empty subset **W** of **V** is a subspace of **V** if and only if for each pair of vectors $\alpha \beta$ in **W** and each scalar c in \mathbb{R} the vector $c\alpha + \beta$ is again in **W**.

For each of the function to be a subspace, it must be closed with respect to addition and scalar multiplication in V defined as, for $f g \in W$ and $c \in \mathbb{R}$

Then,

$$(\mathbf{cf} + \mathbf{g})(-1) = \mathbf{cf}(-1) + \mathbf{g}(-1)$$
 (2.2.2.10)
= $\mathbf{c}(0) + 0$ (2.2.2.11)
= 0 (2.2.2.12)

Thus, $(\mathbf{cf} + \mathbf{g}(-1)) = 0$. Therefore, **W** is a subset of **V** and also a vector space. Therefore **W** is a subspace of **V**.

spanned by the vectors $\begin{pmatrix} 2 \\ -1 \\ 3 \\ 2 \end{pmatrix}$, $\begin{pmatrix} -1 \\ 1 \\ 1 \\ 3 \end{pmatrix}$ and $\begin{pmatrix} 1 \\ 1 \\ 9 \\ 5 \end{pmatrix}$

Hence, f(-1) = 0 is a subspace of **V**.

?

Solution: Expressing the given three vectors as columns of a matrix,

$$\mathbf{A} = \begin{pmatrix} 2 & -1 & 1 \\ -1 & 1 & 1 \\ 3 & 1 & 9 \\ 2 & -3 & -5 \end{pmatrix} \tag{2.2.3.1}$$

and

$$\mathbf{b} = \begin{pmatrix} 3 \\ -1 \\ 0 \\ -1 \end{pmatrix} \tag{2.2.3.2}$$

For the vector \mathbf{b} to be in the subspace of \mathbf{R}^4 spanned by the three vectors.

$$\mathbf{A}\mathbf{x} = \mathbf{b} \tag{2.2.3.3}$$

must have a solution.

$$\begin{pmatrix} 2 & -1 & 1 \\ -1 & 1 & 1 \\ 3 & 1 & 9 \\ 2 & -3 & -5 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 3 \\ -1 \\ 0 \\ -1 \end{pmatrix}$$
 (2.2.3.4)

Forming the augmented matrix and row reducing it by elementary row operations,

$$\begin{pmatrix}
2 & -1 & 1 & 3 \\
-1 & 1 & 1 & -1 \\
3 & 1 & 9 & 0 \\
2 & -3 & -5 & -1
\end{pmatrix}
\xrightarrow{R_2 \leftarrow 2R_2 + R_1, R_3 \leftarrow R_3 - \frac{3}{2}R_1}
\xrightarrow{R_4 \leftarrow R_4 - R_1}$$
(2.2.3.5)

$$\begin{pmatrix} 2 & -1 & 1 & 3 \\ 0 & 1 & 3 & 1 \\ 0 & \frac{5}{2} & \frac{15}{2} & \frac{-9}{2} \\ 0 & -2 & -6 & -4 \end{pmatrix} \xrightarrow{R_3 \leftarrow 2R_3 - 5R_2} \begin{pmatrix} 2 & -1 & 1 & 3 \\ 0 & 1 & 3 & 1 \\ 0 & 0 & 0 & -14 \\ 0 & 0 & 0 & -2 \end{pmatrix}$$

$$(2.2.3.6)$$

From (2.2.3.6), it is clear that the system does

not have a solution. Hence the vector $\begin{pmatrix} 3 \\ -1 \\ 0 \\ -1 \end{pmatrix}$ does

not lie in the subspace of \mathbf{R}^4 spanned by the given three vectors.

2.2.4. Let **W** be the set of all $(x_1, x_2, x_3, x_4, x_5)$ in \mathbb{R}^5 which satisfy

$$2x_1 - x_2 + \frac{4}{3}x_3 - x_4 = 0 (2.2.4.1)$$
$$x_1 + \frac{2}{3}x_3 - x_5 = 0 (2.2.4.2)$$

$$9x_1 - 3x_2 + 6x_3 - 3x_4 - 3x_5 = 0 (2.2.4.3)$$

Find a finite set of vectors which spans **W**. **Solution:** The given equations are,

$$2x_1 - x_2 + \frac{4}{3}x_3 - x_4 = 0 (2.2.4.4)$$
$$x_1 + \frac{2}{3}x_3 - x_5 = 0 (2.2.4.5)$$

$$9x_1 - 3x_2 + 6x_3 - 3x_4 - 3x_5 = 0 (2.2.4.6)$$

which can be written as,

$$\begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 \\ 1 & 0 & \frac{2}{3} & 0 & -1 \\ 9 & -3 & 6 & -3 & -3 \end{pmatrix} \mathbf{x} = 0 \tag{2.2.4.7}$$

Now, the augmented matrix,

$$\begin{pmatrix}
2 & -1 & \frac{4}{3} & -1 & 0 & 0 \\
1 & 0 & \frac{2}{3} & 0 & -1 & 0 \\
9 & -3 & 6 & -3 & -3 & 0
\end{pmatrix}$$
(2.2.4.8)

$$\stackrel{R_3=R_3-3R_1-3R_2}{\longleftrightarrow} \begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 & 0 \\ 1 & 0 & \frac{2}{3} & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$
(2.2.4.9)

$$\stackrel{R_2=R_2-\frac{1}{2}R_1}{\longleftrightarrow} \begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 & 0\\ 0 & \frac{1}{2} & 0 & \frac{1}{2} & -1 & 0\\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$
(2.2.4.10)

$$\stackrel{R_2=2R_2}{\longleftrightarrow} \begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 & 0\\ 0 & 1 & 0 & 1 & -2 & 0\\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(2.2.4.11)$$

$$\stackrel{R_1=R_1+R_2}{\longleftrightarrow} \begin{pmatrix} 2 & 0 & \frac{4}{3} & 0 & -2 & 0\\ 0 & 1 & 0 & 1 & -2 & 0\\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(2.2.4.12)$$

So,

$$2x_1 + \frac{4}{3}x_3 - 2x_5 = 0 (2.2.4.13)$$

$$x_2 + x_4 - 2x_5 = 0 (2.2.4.14)$$

Solving the equations we get,

$$x_1 = -\frac{2}{3}x_3 + x_5 \tag{2.2.4.15}$$

$$x_2 = -x_4 + 2x_5 \tag{2.2.4.16}$$

which can be written as,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} \tag{2.2.4.17}$$

$$= \begin{pmatrix} -\frac{2}{3}x_3 + x_5 \\ -x_4 + 2x_5 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 (2.2.4.18)

$$= x_3 \begin{pmatrix} -\frac{2}{3} \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} 0 \\ -1 \\ 0 \\ 1 \\ 0 \end{pmatrix} + x_5 \begin{pmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$
 (2.2.4.19)

where x_3, x_4 and $x_5 \in \mathbb{R}$. Hence, the vectors

$$\begin{pmatrix} -\frac{2}{3} \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 \\ 0 \\ 1 \\ 0 \end{pmatrix} \text{ and } \begin{pmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 1 \end{pmatrix} \text{ will span } \mathbf{W}$$

- 2.2.5. Let **F** be a field and let n be a positive integer (n≥2). Let **V** be the vector space of all n×n matrices over **F**. Which of the following set of matrices **A** in **V** are subspaces of **V**?
 - a) all invertible **A**;
 - b) all non-invertible A;
 - c) all A such that AB = BA, where B is some fixed matrix in V;
 - d) all **A** such that $A^2 = A$.

Solution:

a) Let the matrices A and $B \in V$, be set of invertible matrix. For them to be a subspace they need to be closed under addition. Let,

$$\mathbf{A} = \mathbf{I} \tag{2.2.5.1}$$

$$\mathbf{B} = -\mathbf{I} \tag{2.2.5.2}$$

It could be easily proven that both matrices **A** and **B** are invertible as,

$$rank(\mathbf{I}_{nxn}) = rank \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{pmatrix}_{nxn}$$

$$(2.2.5.3)$$

$$\implies rank(-\mathbf{I}_{nxn}) = rank(\mathbf{I}_{nxn}) = n$$

$$(2.2.5.4)$$

or it is a full rank matrix as there are n pivots.

$$\therefore \mathbf{A} + \mathbf{B} = \mathbf{0}. \tag{2.2.5.5}$$

But the zero matrix **0** is non-invertible as,

$$rank(\mathbf{0}_{nxn}) = rank \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}_{nxn}$$

$$(2.2.5.6)$$

$$\implies rank(\mathbf{0}_{nxn}) = 0$$

$$(2.2.5.7)$$

- \therefore the set of invertible matrices are not closed under addition. Hence not a subspace of V.
- b) Let the matrices $A_1, A_2, \dots, A_n \in V$, be set of non-invertible matrix. For them to be a subspace they need to be closed under addition. Let,

$$\mathbf{A_{1}} = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}_{nxn}$$

$$\mathbf{A_{2}} = \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}_{nxn}$$

$$(2.2.5.8)$$

$$(2.2.5.9)$$

$$(2.2.5.9)$$

$$\mathbf{A_n} = \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 \end{pmatrix}_{nxn}$$
 (2.2.5.10)

It could be proven that matrices A_1 ,

 A_2, \dots, A_n are non-invertible as,

$$rank(\mathbf{A_1}) = rank \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}$$

$$(2.2.5.12)$$

$$\implies rank(\mathbf{A_1}) = 1$$

$$(2.2.5.13)$$

or there is only one pivot hence rank is 1.

$$\implies \mathbf{A_1} + \mathbf{A_2} + \mathbf{A_3} + \cdots + \mathbf{A_n} = \mathbf{I}_{nxn}$$
(2.2.5.14)

Now the identity matrix **I** is invertible as shown in equation (2.2.5.4). ∴ the set of non-invertible matrices are not closed under addition. Hence not a subspace of V.

c) **Theorem 1:**. A non-empty subset W of V is a subspace of V if and only if for each pair of vectors α , β in W and each scalar $c \in F$, the vector $c\alpha + \beta \in W$.

Let the matrices A_1 and A_2 satisfy,

$$\mathbf{A_1B} = \mathbf{BA_1} \tag{2.2.5.15}$$

$$\mathbf{A_2B} = \mathbf{BA_2} \tag{2.2.5.16}$$

Let, $c \in \mathbf{F}$ be any constant.

$$(cA_1 + A_2)B = cA_1B + A_2B$$
 (2.2.5.17)

Substituting from equations (2.2.5.15) and (2.2.5.16) to (2.2.5.17),

$$\Rightarrow (c\mathbf{A}_1 + \mathbf{A}_2)\mathbf{B} = c\mathbf{B}\mathbf{A}_1 + \mathbf{B}\mathbf{A}_2$$

$$(2.2.5.18)$$

$$\Rightarrow \mathbf{B}c\mathbf{A}_1 + \mathbf{B}\mathbf{A}_2$$

$$(2.2.5.19)$$

$$\Rightarrow \mathbf{B}(c\mathbf{A}_1 + \mathbf{A}_2)$$

$$(2.2.5.20)$$

Thus, $(cA_1 + A_2)$ satisfy the criteria and from Theorem-1 it can be seen that the set is a subspace of V.

d) Let A and $B \in V$ be set of matrices such that,

$$\mathbf{A}^2 = \mathbf{A} \tag{2.2.5.21}$$

$$\mathbf{B}^2 = \mathbf{B} \tag{2.2.5.22}$$

Now for them to be closed under addition,

$$(\mathbf{A} + \mathbf{B})^2 = \mathbf{A} + \mathbf{B} \tag{2.2.5.23}$$

Which is not always same. Example let,

$$\mathbf{A} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \tag{2.2.5.24}$$

$$\mathbf{B} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \tag{2.2.5.25}$$

Clearly,

$$\mathbf{A}^2 = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} = \mathbf{A}$$
(2.2.5.26)

$$\mathbf{B}^2 = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \mathbf{B}$$
(2.2.5.27)

Now,

$$\mathbf{A} + \mathbf{B} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

$$(2.2.5.28)$$

$$\implies (\mathbf{A} + \mathbf{B})^2 = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix}$$

Hence, clearly from equations (2.2.5.28) and (2.2.5.29),

$$(\mathbf{A} + \mathbf{B})^2 \neq \mathbf{A} + \mathbf{B}$$
 (2.2.5.30)

 \therefore the set of all A such that $A^2 = A$ is not closed under addition. Hence, not a subspace of V.

- 2.2.6. a. Prove that only subspace of \mathbb{R}^1 are \mathbb{R}^1 and the zero subspace
 - b. Prove that a subspace of \mathbb{R}^2 is \mathbb{R}^2 , or the zero subspace, or consists of all scalar multiples of some fixed vector in \mathbb{R}^2 . (The last type of subspace is, intuitively, a straight line through the origin.)
 - c. Can you describe the subspaces of \mathbb{R}^3 ? Solution:
 - a. Let $W \neq 0$ be subspace of \mathbb{R}^1 . Then W is a nonempty subset of \mathbb{R}^1 and there exist $w \in W$ such that $w \neq 0$ which gives us that there exist w^{-1} .

Let $x \in \mathbb{R}^1$. Since W is in \mathbb{R}^1 we have that it is closed under scalar

multiplication which gives us that $(xw^{-1})w = x(w^{-1}w) = x.1 = x \in W$

Hence $\mathbb{R}^1 \subset W$ and therefore $W = \mathbb{R}^1$

Thus the only subspace of \mathbb{R}^1 distinct of 0 is \mathbb{R}^1 and therefore only subspaces of \mathbb{R}^1 are 0 and \mathbb{R}^1 .

b. Clearly, 0 and \mathbb{R}^2 itself are subspaces of \mathbb{R}^2 . If $u \neq 0$ and $u \in \mathbb{R}^2$ then span $\{\mathbf{u}\} = c\mathbf{u} : c \in \mathbb{R} = \text{set of all scalar multiples of } \mathbf{u}$ is a subspace of \mathbb{R}^2 .

To show that these are the only subspaces of \mathbb{R}^2 , assume that $W \subset \mathbb{R}^2$ is any subspace of \mathbb{R}^2 . Since $W \subset \mathbb{R}^2$ is a subspace of \mathbb{R}^2 , we have that $\mathbf{0} \in W$. If $W \neq \mathbf{0}$ then there is a vector $\mathbf{u} \neq 0$ and $\mathbf{u} \in W$, and hence W contains $c\mathbf{u}$ for every $c \in \mathbb{R}$. If $W \neq span\{\mathbf{u}\}$, then there is a vector $v \in W$ so that $\mathbf{v} \neq k\mathbf{u}$ for any $k \in \mathbb{R}$.

Then $\mathbf{z} = c\mathbf{u} + d\mathbf{v} \in span\{\mathbf{u}, \mathbf{v}\}$ for any $c, d \in \mathbb{R}$. Since W is a subspace $c\mathbf{u}$ and $d\mathbf{v} \in W$ for any $c, d \in \mathbb{R}$, and hence so does $\mathbf{z} = c\mathbf{u} + d\mathbf{v}$. Thus $\mathbf{z} \in span\{\mathbf{u}, \mathbf{v}\} \implies z \in W$, and so $span\{\mathbf{u}, \mathbf{v}\} \subset W \subset \mathbb{R}^2$.

Let $\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2$ be any vector in \mathbb{R}^2 , and let $\mathbf{u} = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} \neq \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ and let $\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} \neq \begin{pmatrix} 0 \\ 0 \end{pmatrix}$. We show that there are real numbers c and d so that $c\mathbf{u} + d\mathbf{v} = \mathbf{x}$

$$\begin{pmatrix} cu_1 \\ cu_2 \end{pmatrix} + \begin{pmatrix} dv_1 \\ dv_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$
 (2.2.6.1)

$$\begin{pmatrix} u_1 & v_1 \\ u_2 & v_2 \end{pmatrix} \begin{pmatrix} c \\ d \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$
 (2.2.6.2)

Since $\mathbf{v} \neq k\mathbf{u}$ for any $k \in \mathbb{R}$ and since $\mathbf{u} = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} \neq \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ assume that $u_1 \neq 0$, and since

 $k\mathbf{u} \neq \mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ assume that $v_2 \neq 0$.

$$A = \begin{pmatrix} u_1 & v_1 \\ u_2 & v_2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$
 (2.2.6.3)

Hence A is row equivalent to I_2 and so A is invertible and so (2.2.6.2) has unique solution for c and d. Thus for any $\mathbf{x} \in \mathbb{R}^2$ we can find real numbers c and d such that $\mathbf{x} = c\mathbf{u} + d\mathbf{v}$. Hence $\mathbf{x} \in \mathbb{R}^2 \implies x \in span\{\mathbf{u}, \mathbf{v}\}$. Thus $\mathbb{R}^2 \subset span\{\mathbf{u}, \mathbf{v}\} \subset W \subset \mathbb{R}^2$.

Hence $span\{\mathbf{u},\mathbf{v}\} = \mathbf{W} = \mathbb{R}^2$, and so the only subspace of \mathbb{R}^2 are $\mathbf{0}$, \mathbb{R}^2 , and $L = c\mathbf{u} : \mathbf{u} \neq 0, c \in \mathbb{R}$.

- c. The following are the subspaces of \mathbb{R}^3 :
 - 1. Origin is a trivial subspace of \mathbb{R}^3 .
 - 2. \mathbb{R}^3 itself is a trivial subspace of \mathbb{R}^3 .
 - 3. Every line through origin is subspace of \mathbb{R}^3 .
 - 4. Every plane in \mathbb{R}^3 passing through origin is a subspace \mathbb{R}^3 .

Proof: Let W be a plane passing through origin. We need $\mathbf{0} \in W$, but we have that since we're only considering planes that contain origin. Next, we need W is closed under vector addition. If $\mathbf{w_1}$ and $\mathbf{w_2}$ both belong to W, then so does $\mathbf{w_1} + \mathbf{w_2}$ because it's found by constructing a parallelogram, and the whole parallelogram lies in the plane W. Finally, we need W is closed under scalar products, but it is since scalar multiples lie in a straight line through the origin, and that line lies in W. Thus, each plane W passing through the origin is a subspace of \mathbb{R}^3 .

5. The intersection of any of the above subspaces will also be a subspace of \mathbb{R}^3 . Because intersection of subspaces of a vector space is also a subspace of vector space.

Proof: Let W be a collection of subspaces of V, and let $W = \cap W_i$ be their intersection. Since each W_i is a subspace, each of it contains the zero vector. Thus the zero vector is in the

intersection W, and W is non-empty. Let α and β be vectors in W and let c be a scalar. By definition of W, both α and β belong to each W_i , and because each W_i is a subspace, the vector $(c\alpha + \beta)$ is again in W. Hence by definition of subspace, W is a subspace of V.

These 5 are only subspaces of \mathbb{R}^3 possible. Because dimension of vector space \mathbb{R}^3 is 3. Any subspace of \mathbb{R}^3 should have dimension less than or equal to it's dimension. Hence possible dimensions of subspaces are 0,1,2,3. Only subspace with 0 dimension is origin. Subspaces of dimension 1 with zero vector are lines passing through origin. Subspaces of dimension 2 with zero vector are plane passing through origin. Subspace of dimension 3 are all of \mathbb{R}^3 itself.

2.2.7. Let \mathbf{W}_1 and \mathbf{W}_2 be subspaces of a vector space \mathbf{V} such that the set-theoretic union of \mathbf{W}_1 and \mathbf{W}_2 is also a subspace. Prove that one of the spaces \mathbf{W}_i is contained in the other. **Solution:** Given $\mathbf{W}_1 \cup \mathbf{W}_2$ is a subspace, we need to prove that

$$\mathbf{W}_1 \subseteq \mathbf{W}_2 \quad or \quad \mathbf{W}_2 \subseteq \mathbf{W}_1$$
 (2.2.7.1)

Let us assume that

$$\mathbf{W}_1 \not\subseteq \mathbf{W}_2 \tag{2.2.7.2}$$

We need to show that

$$\mathbf{W}_2 \subseteq \mathbf{W}_1 \tag{2.2.7.3}$$

i.e., the generators of W_2 are in W_1 . Consider a vector, $\mathbf{w}_1 \in \mathbf{W}_1 \backslash \mathbf{W}_2$ and a vector $\mathbf{w}_2 \in \mathbf{W}_2$. Since $\mathbf{W}_1 \cup \mathbf{W}_2$ is a subspace,

$$\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_1 \cup \mathbf{W}_2 \tag{2.2.7.4}$$

$$\implies$$
 $\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_1 \quad or$ (2.2.7.5)

$$\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_2 \tag{2.2.7.6}$$

But, $\mathbf{w}_1 + \mathbf{w}_2 \notin \mathbf{W}_2$ because for some vector $-\mathbf{w}_2 \in \mathbf{W}_2$,

$$(\mathbf{w}_1 + \mathbf{w}_2) - \mathbf{w}_2 = \mathbf{w}_1 \notin \mathbf{W}_2$$
 (2.2.7.7)

Hence it must be that, $\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_1$ because for some vector $-\mathbf{w}_1 \in \mathbf{W}_1$,

$$(\mathbf{w}_1 + \mathbf{w}_2) - \mathbf{w}_1 = w_2 \in \mathbf{W}_1$$
 (2.2.7.8)

Thus, we have shown that every vector \mathbf{w}_2 in \mathbf{W}_2 is also in \mathbf{W}_1 . Hence, $\mathbf{W}_2 \subseteq \mathbf{W}_1$

- 2.2.8. Let V be the vector space of all functions from \mathbf{R} into \mathbf{R} ; let $\mathbf{V_e}$ be the subset of even functions, f(-x) = f(x); let V_0 be the subset of odd functions, f(-x) = -f(x).
 - a) Prove that V_e and V_o are subspaces of V
 - b) Prove that $V_e + V_o = V$
 - c) Prove that $V_e \cap V_o = \{0\}$

Solution:

a) Prove that V_e and V_o are subspaces of V. A non-empty subset W of V is a subspace of **V** if and only if for each pair of vectors α , β in W and each scalar c in F the vector $c\alpha + \beta$ is again in W.

Let $\mathbf{u}, \mathbf{v} \in \mathbf{V_e}$ and $c \in \mathbf{R}$ and let $\mathbf{h} = c\mathbf{u} + \mathbf{v}$. 2.2.3. Let $\mathbf{W_1}$ and $\mathbf{W_2}$ be subspaces of a vector space Then,

$$\mathbf{h}(-x) = c\mathbf{u}(-x) + \mathbf{v}(-x)$$

$$= c\mathbf{u}(x) + \mathbf{v}(x) \qquad (2.2.8.1)$$

$$= \mathbf{h}(x)$$

From (2.2.8.1)

$$\implies \mathbf{h}(-x) = \mathbf{h}(x) \tag{2.2.8.2}$$

$$\implies h \in V_e \tag{2.2.8.3}$$

Let $\mathbf{u}, \mathbf{v} \in \mathbf{V_o}$ and $c \in \mathbf{R}$ and let $\mathbf{h} = c\mathbf{u} + \mathbf{v}$. Then,

$$\mathbf{h}(-x) = c\mathbf{u}(-x) + \mathbf{v}(-x)$$

$$= -c\mathbf{u}(x) - \mathbf{v}(x)$$

$$= -\mathbf{h}(x)$$
(2.2.8.4)

From (2.2.8.4)

$$\implies \mathbf{h}(-x) = -\mathbf{h}(x) \tag{2.2.8.5}$$

$$\implies$$
 h \in **V**₀ (2.2.8.6)

From (2.2.8.3) and (2.2.8.6), V_e and V_o are subspaces of V.

a) Prove that $V_e + V_o = V$.

Let $\mathbf{u} \in \mathbf{V}$

$$\mathbf{u_e}(x) = \frac{\mathbf{u}(x) + \mathbf{u}(-x)}{2}$$
 (2.2.1.7)

$$\mathbf{u_o}(x) = \frac{\mathbf{u}(x) - \mathbf{u}(-x)}{2}$$
 (2.2.1.8)

Equation equation (2.2.1.7) and (2.2.1.8), \mathbf{u}_{e} is

even and \mathbf{u}_0 is odd. Adding both the equations,

$$\mathbf{u} = \mathbf{u_e} + \mathbf{u_o} \tag{2.2.1.9}$$

a) Prove that $V_e \cap V_o = \{0\}$.

Let $\mathbf{u} \in \mathbf{V_e} \cap \mathbf{V_o}$

$$\mathbf{u} \in \mathbf{V_e} \implies \mathbf{u}(-x) = \mathbf{u}(x)$$
 (2.2.2.10)

$$\mathbf{u} \in \mathbf{V_0} \implies \mathbf{u}(-x) = -\mathbf{u}(x)$$
 (2.2.2.11)

Equating (2.2.2.10) and (2.2.2.11),

$$\mathbf{u}(x) = -\mathbf{u}(x) \tag{2.2.2.12}$$

$$\implies 2\mathbf{u}(x) = 0 \tag{2.2.2.13}$$

$$\implies \mathbf{u} = 0 \tag{2.2.2.14}$$

Equations (2.2.8.3), (2.2.8.6),(2.2.1.9),(2.2.2.14) proves 1, 2 and 3.

V such that

$$\mathbf{W_1} + \mathbf{W_2} = \mathbf{V} \tag{2.2.3.1}$$

and
$$W_1 \cap W_2 = 0$$
 (2.2.3.2)

Prove that for each vector α in **V** there are unique vectors α_1 in W_1 and α_2 in W_2 such that

$$\alpha = \alpha_1 + \alpha_2 \tag{2.2.3.3}$$

Solution: Suppose, vectors α_1 and α_2 are not unique.

Consider

$$\alpha_1' \in \mathbf{W_1},$$
 (2.2.3.4)

$$\alpha_2' \in \mathbf{W_2} \tag{2.2.3.5}$$

such that
$$\alpha = \alpha_1' + \alpha_2'$$
 (2.2.3.6)

(2.2.3.3) and (2.2.3.6) indicate

$$\alpha_1 + \alpha_2 = \alpha_1' + \alpha_2' \tag{2.2.3.7}$$

$$\implies \alpha_1 - \alpha_1' = \alpha_2' - \alpha_2 \qquad (2.2.3.8)$$

For α_1 and α'_1 lying in subspace W_1 , defined on field \mathbb{F} , the following holds

$$\alpha_1 + c\alpha_1' \in \mathbf{W}_1, c \in \mathbb{F} \tag{2.2.3.9}$$

$$c = -1 \implies \alpha_1 - \alpha_1' \in \mathbf{W_1} \qquad (2.2.3.10)$$

Similarly,
$$\alpha'_{2} - \alpha_{2} \in \mathbf{W}_{2}$$
 (2.2.3.11)

$$(2.2.3.8) \implies \alpha_1 - \alpha_1' \in \mathbf{W_2}$$
 (2.2.3.12)

(2.2.3.2),(2.2.3.10),(2.2.3.12) indicate

$$\alpha_{1} - \alpha'_{1} = \alpha'_{2} - \alpha_{2} = \mathbf{0}$$

$$\Longrightarrow \alpha_{1} = \alpha'_{1}$$
(2.2.3.13)
$$(2.2.3.14)$$

$$\alpha_2 = \alpha_2' \qquad (2.2.3.15)$$

So, there exists a unique $\alpha_1 \in W_1$ and $\alpha_2 \in W_2$ such that

$$\alpha = \alpha_1 + \alpha_2 \tag{2.2.3.16}$$

where $\alpha \in \mathbf{V}$

2.2.4. Let V be the set of all 2 X 2 matrices A with complex entries which satisfy $A_{11} + A_{22} = 0$ and Show that V is a vector space over the field of real numbers, with the usual operations of matrix addition and multiplication of a matrix by a scalar.

> Solution: A non-empty subset W of V is a subspace of V if and only if for each pair of vectors a, b in W and each scalar c in F the vector $c\mathbf{a} + \mathbf{b}$ is again in W

> Let M be the vector space of all 2 x 2 matrices over \mathbb{C} i.e set of complex numbers and let \mathbb{R} be set of real numbers.

Consider $A,B \in V$ and $c \in \mathbb{C}$

Let
$$\mathbf{A} = \begin{pmatrix} x & y \\ z & w \end{pmatrix}$$
 (2.2.4.1) vectors are not 2.3.2. Are the vectors and $\mathbf{B} = \begin{pmatrix} x' & y' \\ z' & w' \end{pmatrix}$ (2.2.4.2) $\alpha_1 = \begin{pmatrix} 1, 1, 2, 4 \\ \alpha_3 = \begin{pmatrix} 1, -1, -1 \\ \alpha_4 \end{pmatrix}$ (2.2.4.3) linearly independent

$$c\mathbf{A} + \mathbf{B} = \begin{pmatrix} cx + x' & cy + y' \\ cz + z' & cw + w' \end{pmatrix}$$
(2.2.4.4)

$$\therefore (cx + x') + (cw + w') = c(x + w) + (x' + w') = 0$$
(2.2.4.5)

Since V is a subset of M.

Therefore using above Theorem,

$$c\mathbf{A} + \mathbf{B} \, \epsilon \, \mathbf{V} \tag{2.2.4.6}$$

Hence V is a subspace of M as a vector space over \mathbb{C} . Hence **V** is a vector space over \mathbb{R} .

2.3 Bases and Dimension

2.3.1. Prove that if two vectors are linearly dependent, one of them is a scalar multiple of the other.

Solution: consider the row reduced matrix

$$\begin{pmatrix}
1 & 1 & 2 & 4 \\
2 & -1 & -5 & 2 \\
1 & -1 & -4 & 0 \\
2 & 1 & 1 & 6
\end{pmatrix}
\xrightarrow{R_2 \to R_2 - 2R_1}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & -3 & -9 & -6 \\
0 & -2 & -6 & -4 \\
2 & 1 & 1 & 6
\end{pmatrix}$$

$$(2.3.1.1)$$

$$\xrightarrow{R_4 \leftarrow R_4 - 2R_1}
\xrightarrow{R_2 \leftarrow R_4}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & -1 & -3 & -2 \\
0 & -2 & -6 & -4 \\
0 & -3 & -9 & -6
\end{pmatrix}$$

$$(2.3.1.2)$$

$$\xrightarrow{R_4 \leftarrow R_2}
\xrightarrow{R_2 \leftarrow -R_2}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & -1 & 3 & 2 \\
0 & -3 & -9 & -6 \\
0 & -2 & -6 & -4
\end{pmatrix}$$

$$(2.3.1.3)$$

$$\xrightarrow{R_3 \leftarrow R_3 + 3R_2}
\xrightarrow{R_4 \leftarrow R_4 + 2R_2}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & -3 & -9 & -6 \\
0 & -2 & -6 & -4
\end{pmatrix}$$

$$(2.3.1.3)$$

Therefore the rank = no. of pivot columns = 2 (less than no. of columns). Thus the fours vectors are not linearly independent.

$$\alpha_1 = (1, 1, 2, 4), \alpha_2 = (2, -1, -5, 2)$$
 (2.3.2.1)
 $\alpha_3 = (1, -1, -4, 0), \alpha_4 = (2, 1, 1, 6)$ (2.3.2.2)

linearly independent in R^4

Solution: consider the row reduced matrix

$$\begin{pmatrix}
1 & 1 & 2 & 4 \\
2 & -1 & -5 & 2 \\
1 & -1 & -4 & 0 \\
2 & 1 & 1 & 6
\end{pmatrix}
\xrightarrow{R_2 \to R_2 - 2R_1}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & -3 & -9 & -6 \\
0 & -2 & -6 & -4 \\
2 & 1 & 1 & 6
\end{pmatrix}$$

$$(2.3.2.3)$$

$$\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & -3 & -9 & -6 \\
0 & -2 & -6 & -4 \\
2 & 1 & 1 & 6
\end{pmatrix}$$

$$(2.3.2.3)$$

$$\begin{pmatrix}
R_4 \leftarrow R_4 - 2R_1 \\
R_2 \leftarrow R_4
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & -1 & -3 & -2 \\
0 & -2 & -6 & -4 \\
0 & -3 & -9 & -6
\end{pmatrix}$$

$$(2.3.2.4)$$

$$\begin{pmatrix}
R_4 \leftarrow R_4 - 2R_1 \\
R_2 \leftarrow R_4
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & -1 & -3 & -2 \\
0 & -2 & -6 & -4 \\
0 & -3 & -9 & -6
\end{pmatrix}$$

$$(2.3.3.4)$$

$$\begin{pmatrix}
R_4 \leftarrow R_2 \\
R_2 \leftarrow -R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & -3 & -9 & -6 \\
0 & 2 & -6 & -4
\end{pmatrix}$$

$$(2.3.3.5)$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 + 3R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & -2 & -6 & -4
\end{pmatrix}$$

$$(2.3.2.5)$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 + 3R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 + 3R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 + 3R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 + 3R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 + 3R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 + 3R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 - 9R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & 0 & 0 & 0 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 - 9R_2 \\
R_4 \leftarrow R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 - 9R_2 \\
R_4 \leftarrow R_4 - R_4 + 2R_2
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & 0 & 0 & 0 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$\begin{pmatrix}
R_3 \leftarrow R_3 - 9R_2 \\
R_4 \leftarrow R_4 - 2R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_4 - R_1 \\
R_4 \leftarrow R_4 - 2R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R_1
\end{pmatrix}
\begin{pmatrix}
R_3 \leftarrow R_3 - R_3 - R$$

Therefore the rank = no. of pivot columns = 2 (less than no. of columns). Thus the fours vectors are not linearly independent.

2.3.3. Find a basis for the subspace of \mathbb{R}^4 spanned by the four vectors

$$\alpha_1 = \begin{pmatrix} 1 & 1 & 2 & 4 \end{pmatrix}$$
 (2.3.3.1)

$$\alpha_1 = \begin{pmatrix} 1 & 1 & 2 & 1 \end{pmatrix}$$
 (2.3.3.1)
 $\alpha_2 = \begin{pmatrix} 2 & -1 & -5 & 2 \end{pmatrix}$ (2.3.3.2)
 $\alpha_3 = \begin{pmatrix} 1 & -1 & -4 & 0 \end{pmatrix}$ (2.3.3.3)

$$\alpha_3 = \begin{pmatrix} 1 & -1 & -4 & 0 \end{pmatrix} \tag{2.3.3.3}$$

$$\alpha_4 = \begin{pmatrix} 2 & 1 & 1 & 6 \end{pmatrix} \tag{2.3.3.4}$$

Solution: The basis of the given four vectors is equivalent to finding the basis of column-space $C(\mathbf{A})$ of a matrix **A** defined as follows,

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 1 & 2 \\ 1 & -1 & -1 & 1 \\ 2 & -5 & -4 & 1 \\ 4 & 2 & 0 & 6 \end{pmatrix} \tag{2.3.3.5}$$

Now we calculate the row echelon form of A

as follows,

$$\begin{pmatrix}
1 & 2 & 1 & 2 \\
1 & -1 & -1 & 1 \\
2 & -5 & -4 & 1 \\
4 & 2 & 0 & 6
\end{pmatrix}
\xrightarrow{R_2 = R_2 - R_1}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & -3 & -2 & -1 \\
0 & -9 & -6 & -3 \\
4 & 2 & 0 & 6
\end{pmatrix}$$

$$(2.3.3.6)$$

$$\xrightarrow{R_4 = R_4 - R_1}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & -3 & -2 & -1 \\
0 & -9 & -6 & -3 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$(2.3.3.7)$$

$$\xrightarrow{R_2 = -\frac{1}{3}R_2}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & -9 & -6 & -3 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$(2.3.3.8)$$

$$\xrightarrow{R_3 = R_3 - 9R_2}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & 0 & 0 & 0 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$(2.3.3.9)$$

$$\xrightarrow{R_4 = R_4 + 6R_2}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$

$$(2.3.3.10)$$

From (2.3.3.10) we can see that the first column and second column of A contains pivot values. Hence the column 1 and column 2 are the basis of the subspace of \mathbb{R}^4 spanned by the given vectors α_1 , α_2 , α_3 , α_4

Hence the required basis vectors are,

$$\mathbf{a_1} = \begin{pmatrix} 1 & 1 & 2 & 4 \end{pmatrix}$$
 (2.3.3.11)

$$\mathbf{a_2} = \begin{pmatrix} 2 & -1 & -5 & 2 \end{pmatrix}$$
 (2.3.3.12)

2.3.4. Show that the vectors

$$\alpha_1 = \begin{pmatrix} 1 & 0 & -1 \end{pmatrix}$$
 $\alpha_2 = \begin{pmatrix} 1 & 2 & 1 \end{pmatrix}$ (2.3.4.1)
 $\alpha_3 = \begin{pmatrix} 0 & -3 & 2 \end{pmatrix}$ (2.3.4.2)

form a basis for \mathbb{R}^3 . Express each of the standard basis vectors as linear combinations of $(\alpha_1 \ \alpha_2 \ \alpha_3)$

Solution:

Theorem 2.1. Let **V** be an n-dimensional vector space over the field \mathbf{F} , and let β and β' be two ordered basis of V. Then, there is a unique, necessarily invertible, $n \times n$ matrix **P** with entries in **F** such that

a)
$$\left[\alpha\right]_{\beta} = \mathbf{P} \left[\alpha\right]_{\beta'}$$

b) $\left[\alpha\right]_{\beta'} = \mathbf{P}^{-1} \left[\alpha\right]_{\beta}$

for every vector α in **V**. The columns of **P** are given by

$$\mathbf{P_j} = [\alpha_j]_{\beta}$$
 $j = 1, 2, ..., n$ (2.3.4.3)

In order to show that the set of vectors α_1 , α_2 , and α_3 are basis for \mathbb{R}^3 . We first show that α_1 , α_2 , and α_3 a are linearly independent in \mathbb{R}^3 and also they span \mathbb{R}^3 . Consider,

$$\mathbf{A} = \begin{pmatrix} \alpha_1^T & \alpha_2^T & \alpha_3^T \end{pmatrix} \tag{2.3.4.4}$$

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -3 \\ -1 & 1 & 2 \end{pmatrix} \tag{2.3.4.5}$$

Now,by row reduction

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -3 \\ -1 & 1 & 2 \end{pmatrix} \xrightarrow{R_3 = R_3 + R_1} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -3 \\ 0 & 2 & 2 \end{pmatrix} \quad (2.3.4.6)$$

$$\stackrel{R_3=R_3-R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -3 \\ 0 & 0 & 5 \end{pmatrix} \quad (2.3.4.7)$$

$$\stackrel{R_2 = \frac{R_2}{2}}{\longleftrightarrow} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & \frac{-3}{2} \\ 0 & 0 & 5 \end{pmatrix}$$
(2.3.4.8)

$$\stackrel{R_1=R_1-R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{3}{2} \\ 0 & 1 & \frac{-3}{2} \\ 0 & 0 & 5 \end{pmatrix} \quad (2.3.4.9)$$

$$\stackrel{R_3 = \frac{R_3}{5}}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{3}{2} \\ 0 & 1 & \frac{-3}{2} \\ 0 & 0 & 1 \end{pmatrix} \qquad (2.3.4.10)$$

$$\xrightarrow{R_1 = R_1 - \frac{3}{2}R_3} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & \frac{-3}{2} \\ 0 & 0 & 1 \end{pmatrix}$$
(2.3.4.11)

$$\stackrel{R_2 = R_2 + \frac{3}{2}R_3}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} (2.3.4.12)$$

(2.3.4.12) is the row reduced echelon form of **A** and since it is identity matrix of order 3, we say that vectors α_1 , α_2 , and α_3 are linearly independent and their column space is \mathbb{R}^3 which means vectors α_1 , α_2 , and α_3 span \mathbb{R}^3 . Hence, vectors α_1 , α_2 , and α_3 form a basis for \mathbb{R}^3 .

Now, use theorem (2.1), and calculate the inverse of (2.3.4.5) then the columns of A^{-1} will give the coefficients to write the standard basis vectors in terms of α_i 's. We try to find the inverse of A by row-reducing the augmented matrix.A|I

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -3 \\ -1 & 1 & 2 \end{pmatrix} \tag{2.3.4.13}$$

Now, by row reducing A|I as follows

$$\begin{pmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 2 & -3 & 0 & 1 & 0 \\ -1 & 1 & 2 & 0 & 0 & 1 \end{pmatrix} (2.3.4.14)$$

$$\stackrel{R_3=R_3+R_1}{\longleftrightarrow} \begin{pmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 2 & -3 & 0 & 1 & 0 \\ 0 & 2 & 2 & 1 & 0 & 1 \end{pmatrix} (2.3.4.15)$$

$$\stackrel{R_3=R_3-R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 2 & -3 & 0 & 1 & 0 \\ 0 & 0 & 5 & 1 & -1 & 1 \end{pmatrix} (2.3.4.16)$$

$$\stackrel{R_2 = \frac{R_2}{2}}{\longleftrightarrow} \begin{pmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & \frac{-3}{2} & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 5 & 1 & -1 & 1 \end{pmatrix} (2.3.4.17)$$

$$\stackrel{R_1=R_1-R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{3}{2} & 1 & \frac{-1}{2} & 0\\ 0 & 1 & \frac{-3}{2} & 0 & \frac{1}{2} & 0\\ 0 & 0 & 5 & 1 & -1 & 1 \end{pmatrix} (2.3.4.18)$$

$$\stackrel{R_3 = \frac{R_3}{5}}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{3}{2} & 1 & \frac{-1}{2} & 0\\ 0 & 1 & \frac{-3}{2} & 0 & \frac{1}{2} & 0\\ 0 & 0 & 1 & \frac{1}{5} & \frac{-1}{5} & \frac{1}{5} \end{pmatrix} (2.3.4.19)$$

$$\begin{array}{c}
R_{3} = \frac{R_{3}}{5} \\
\stackrel{R_{3}}{\longleftrightarrow} \\
\begin{pmatrix}
1 & 0 & \frac{3}{2} & 1 & \frac{-1}{2} & 0 \\
0 & 1 & \frac{-3}{2} & 0 & \frac{1}{2} & 0 \\
0 & 0 & 1 & \frac{1}{5} & \frac{-1}{5} & \frac{1}{5}
\end{pmatrix}$$

$$\xrightarrow{R_{1} = R_{1} - \frac{3R_{3}}{2}} \\
\begin{pmatrix}
1 & 0 & 0 & \frac{7}{10} & \frac{-1}{5} & \frac{-3}{10} \\
0 & 1 & \frac{-3}{2} & 0 & \frac{1}{2} & 0 \\
0 & 0 & 1 & \frac{1}{5} & \frac{-1}{5} & \frac{1}{5}
\end{pmatrix}$$

$$\xrightarrow{R_{2} = R_{2} + \frac{3R_{3}}{2}} \\
\begin{pmatrix}
1 & 0 & 0 & \frac{7}{10} & \frac{-1}{5} & \frac{-3}{10} \\
0 & 1 & 0 & \frac{3}{10} & \frac{1}{5} & \frac{3}{10} \\
0 & 0 & 1 & \frac{1}{5} & \frac{-1}{5} & \frac{1}{5}
\end{pmatrix}$$

$$(2.3.4.21)$$

$$\stackrel{R_2=R_2+\frac{3R_3}{2}}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & \frac{7}{10} & \frac{-1}{5} & \frac{-3}{10} \\ 0 & 1 & 0 & \frac{3}{10} & \frac{1}{5} & \frac{3}{10} \\ 0 & 0 & 1 & \frac{1}{5} & \frac{-1}{5} & \frac{1}{5} \end{pmatrix} (2.3.4.21)$$

Thus, by (2.3.4.21), we have

$$\mathbf{A}^{-1} = \begin{pmatrix} \frac{7}{10} & \frac{-1}{5} & \frac{-3}{10} \\ \frac{3}{10} & \frac{1}{5} & \frac{3}{10} \\ \frac{1}{5} & \frac{-1}{5} & \frac{1}{5} \end{pmatrix}$$
 (2.3.4.22)

Now, let $e_1 = (1 \ 0 \ 0)$, $e_2 = (0 \ 1 \ 0)$, and $\mathbf{e_3} = \begin{pmatrix} 0 & 0 & 1 \end{pmatrix}$ be the standard basis for \mathbb{R}^3 . Hence, each of the standard basis vectors as linear combinations of $\alpha_1, \alpha_2, \alpha_3$ is as under

$$\mathbf{e_1} = \frac{7}{10}\alpha_1 + \frac{3}{10}\alpha_2 + \frac{1}{5}\alpha_3 \qquad (2.3.4.23)$$

$$\mathbf{e_2} = -\frac{1}{5}\alpha_1 + \frac{1}{5}\alpha_2 - \frac{1}{5}\alpha_3 \tag{2.3.4.24}$$

$$\mathbf{e_3} = \frac{-3}{10}\alpha_1 + \frac{3}{10}\alpha_2 + \frac{1}{5}\alpha_3 \qquad (2.3.4.25)$$

2.3.5. Find three vectors in R^3 which are linearly dependent and are such that any two of them are linearly independent

Solution: Let

$$\mathbf{v_1} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \tag{2.3.5.1}$$

$$\mathbf{v_2} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \tag{2.3.5.2}$$

$$\mathbf{v_3} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \tag{2.3.5.3}$$

Then,

$$\mathbf{v_1} + \mathbf{v_2} - \mathbf{v_3} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \tag{2.3.5.4}$$

Thus, they are linearly dependent. As We know that $\mathbf{v_1}$ and $\mathbf{v_2}$ are linearly independent as they form the standard basis vectors.

Now, Suppose

$$\mathbf{av_1} + \mathbf{bv_2} = 0 \tag{2.3.5.5}$$

Then

$$\begin{pmatrix} a+b\\b\\0 \end{pmatrix} = \begin{pmatrix} 0\\0\\0 \end{pmatrix}$$
 (2.3.5.6)

Hence, the second co-ordinate implies $\mathbf{b} = 0$ and the first coordinate in turn implies $\mathbf{a} = 0$. Thus $\mathbf{v_1}$ and $\mathbf{v_2}$ are linearly independent. Similarly, $\mathbf{v_2}$ and $\mathbf{v_3}$ are linearly independent

2.3.6. Let **V** be the vector space of all 2×2 matrices over the field **F**. Prove that **V** has dimension 4 by exhibiting a basis for **V** which has four elements.

Solution:

Let

$$v_{11} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \quad v_{12} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$
 (2.3.6.1)

$$v_{21} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \quad v_{22} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
 (2.3.6.2)

Suppose $av_{11} + bv_{12} + cv_{21} + dv_{22} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$. Then

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$
 (2.3.6.3)

The only values of a, b, c, d which makes the equation (2.3.6.3) satisfied is, when a = b = c = d = 0. Thus v_1, v_2, v_3, v_4 are linearly independent.

Now, let $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$ be any 2×2 matrix. Then $\begin{pmatrix} a & b \\ c & d \end{pmatrix} = av_{11} + bv_{12} + cv_{21} + dv_{22}$. Thus $v_{11}, v_{12}, v_{21}, v_{22}$ span the space of 2×2 matrix.

Thus v_{11} , v_{12} , v_{21} , v_{22} are both linearly independent and they span the span of all 2×2 matrices. So, v_{11} , v_{12} , v_{21} , v_{22} constitute a basis for the space of all 2×2 matrices.

We know that, the dimension of a vector space V, denoted by dim(V), is the number of basis for V. Therefore, dim(V) = 4.

2.3.7. Let V be the vector space of all 2×2 matrices over the field \mathbb{F} . Let W_1 be the set of matrices of the form

$$\begin{pmatrix} x & -x \\ y & z \end{pmatrix} \tag{2.3.7.1}$$

and let W_2 be the set of matrices of the form

$$\begin{pmatrix} a & b \\ -a & c \end{pmatrix} \tag{2.3.7.2}$$

- a) Prove that W_1 and W_2 are subspaces of V.
- b) Find the dimension of W_1 , W_2 , $W_1 + W_2$ and $W_1 \cap W_2$.

Solution: A non-empty subset W of V is a subspace of V if and only if for each pair of vectors α , β in W and each scalar $c \in F$, the vector $c\alpha + \beta \in W$.

a) Let $A_1, A_2 \in W_1$ where,

$$A_1 = \begin{pmatrix} x_1 & -x_1 \\ y_1 & z_1 \end{pmatrix}, A_2 = \begin{pmatrix} x_2 & -x_2 \\ y_2 & z_2 \end{pmatrix}$$
 (2.3.7.3)

Let $c \in F$ then.

$$cA_1 + A_2 = \begin{pmatrix} cx_1 + x_2 & -cx_1 - x_2 \\ cy_1 + y_2 & cz_1 + z_2 \end{pmatrix} = \begin{pmatrix} u & -u \\ v & w \end{pmatrix}$$
(2.3.7.4)

Thus $cA_1 + A_2 \in W_1$. Hence W_1 is a **subspace.** Similarly, let $A_1, A_2 \in W_2$ where,

$$A_1 = \begin{pmatrix} a_1 & b_1 \\ -a_1 & c_1 \end{pmatrix}, A_2 = \begin{pmatrix} a_2 & b_2 \\ -a_2 & c_2 \end{pmatrix}$$
 (2.3.7.5)

Let $c \in F$ then,

$$cA_1 + A_2 = \begin{pmatrix} ca_1 + a_2 & cb_1 + b_2 \\ -ca_1 - a_2 & cc_1 + c_2 \end{pmatrix} = \begin{pmatrix} u & v \\ -u & w \end{pmatrix}$$
(2.3.7.6)

Thus $cA_1 + A_2 \in W_2$. Hence W_2 is a subspace.

b) The subspace W_1 can be given as,

$$\begin{pmatrix} x & -x \\ y & z \end{pmatrix} = x \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + z \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

$$= xA_1 + yA_2 + zA_2$$

$$(2.3.7.8)$$

Now,

$$x \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + z \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$(2.3.7.9)$$

$$\implies x = y = z = 0$$

$$(2.3.7.10)$$

 A_1, A_2, A_3 are linearly independent and spans W_1 . Thus $\{A_1, A_2, A_3\}$ forms basis for W_1 .

 \therefore dimension of W_1 is 3.

The subspace W_2 can be given as,

$$\begin{pmatrix} a & b \\ -a & c \end{pmatrix} = a \begin{pmatrix} 1 & 0 \\ -1 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
 2.3.8. Let **V** be the space of 2×2 matrices over **F**. Find a basis $\{\mathbf{A}_1, \mathbf{A}_2, \mathbf{A}_3, \mathbf{A}_4\}$ for **V** such that
$$= aA_1 + bA_2 + cA_2 \qquad (2.3.7.12)$$

$$A_j^2 = \mathbf{A}_j \text{ for each } j$$

Now,

$$a \begin{pmatrix} 1 & 0 \\ -1 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$(2.3.7.13)$$

$$\implies a = b = c = 0$$

$$(2.3.7.14)$$

 A_1, A_2, A_3 are linearly independent and spans W_2 . Thus $\{A_1, A_2, A_3\}$ forms basis for W_2 .

 \therefore dimension of W_2 is 3.

Subspace $W_1 + W_2$ is given by,

$$\begin{pmatrix} x+a & -x+b \\ y-a & z+c \end{pmatrix}$$
 (2.3.7.15)

For $x + a \neq -x + b \neq y - a \neq z + c$,

$$\begin{pmatrix} x+a & -x+b \\ y-a & z+c \end{pmatrix} = \begin{pmatrix} j & k \\ l & m \end{pmatrix}$$
 (2.3.7.16)
$$= j \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + k \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + l \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + m \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
 (2.3.7.17)
$$= jA_1 + kA_2 + lA_3 + mA_4$$
 (2.3.7.18)

Now,

$$jA_1 + kA_2 + lA_3 + mA_4 = 0$$
 (2.3.7.19)
 $\implies j = k = l = m = 0$ (2.3.7.20)

 A_1, A_2, A_3, A_4 are linearly independent and spans $W_1 + W_2$. Thus $\{A_1, A_2, A_3, A_4\}$ forms a basis.

 \therefore dimension of $W_1 + W_2$ is 4.

The subspace $W_1 \cap W_2$ is given as,

$$\begin{pmatrix} x & -x \\ -x & y \end{pmatrix} = x \begin{pmatrix} 1 & -1 \\ -1 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

$$= xA_1 + yA_2 \qquad (2.3.7.21)$$

Now,

$$x \begin{pmatrix} 1 & -1 \\ -1 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \quad (2.3.7.23)$$

$$\implies x = y = 0 \qquad (2.3.7.24)$$

 A_1, A_2 are linearly independent and spans $W_1 \cap W_2$. Thus, $\{A_1, A_2\}$ forms a basis.

 \therefore dimension of $W_1 \cap W_2$ is 2.

Find a basis $\{A_1, A_2, A_3, A_4\}$ for V such that $\mathbf{A}_{i}^{2} = \mathbf{A}_{j}$ for each j

Solution: Every 2×2 matrix may be written as

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
(2.3.8.1)

This shows that

$$\{\mathbf{A}_{1}, \mathbf{A}_{2}, \mathbf{A}_{3}, \mathbf{A}_{4}\} = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$
(2.3.8.2)

can be the basis for the space V of all 2×2 matrices. However A_2 and A_3 doesn't satisfy the property of $A^2 = A$. Consider b = 0 and c = 0, then the matrix

$$\begin{pmatrix} a & 0 \\ 0 & d \end{pmatrix} \tag{2.3.8.3}$$

can't be a basis as it is the linear combination of A_1 and A_4 . Hence either b or c or both must be non zero. Hence,

$$\mathbf{A}_2 = \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} \tag{2.3.8.4}$$

$$\mathbf{A}_3 = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} \tag{2.3.8.5}$$

Here, $\mathbf{A}_2^2 = \mathbf{A}_2$ and $\mathbf{A}_3^2 = \mathbf{A}_3$. Therefore the basis can be

$$\{\mathbf{A}_{1}, \mathbf{A}_{2}, \mathbf{A}_{3}, \mathbf{A}_{4}\} = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$
(2.3.8.6)

 $\{A_1, A_2, A_3, A_4\}$ forms the basis, iff they are linearly independent and the linear combination of them span the space **V**. To show that they are linearly independent, we show that the equation has a trivial solution.

$$a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$(2.3.8.7)$$

$$\implies a + b = 0$$

$$(2.3.8.8)$$

$$b = 0$$

$$(2.3.8.9)$$

$$c = 0$$

$$(2.3.8.10)$$

$$c + d = 0$$

$$(2.3.8.11)$$

The corresponding matrix form is Ax = 0

$$\begin{pmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$
 (2.3.8.12)

Row reducing the augmented matrix,

$$\begin{pmatrix}
1 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 1 & 0
\end{pmatrix}
\xrightarrow{R_2 \longleftrightarrow R_3}
\begin{pmatrix}
1 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & 0
\end{pmatrix}$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_2}
\begin{pmatrix}
1 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1 & 0
\end{pmatrix}$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_2}
\begin{pmatrix}
1 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & 0
\end{pmatrix}$$

$$\xrightarrow{(2.3.8.14)}$$

Therefore, a = b = c = d = 0. Hence the matrices are linearly independent. To show that the linear combination of $\{A_1, A_2, A_3, A_4\}$ span the space V, consider an arbitrary matrix,

$$\begin{pmatrix} w & x \\ y & z \end{pmatrix} \tag{2.3.8.15}$$

Compute a, b, c, d such that

$$\begin{pmatrix} w & x \\ y & z \end{pmatrix} = a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
(2.3.8.16)

$$= \begin{pmatrix} a+b & c \\ b & c+d \end{pmatrix} \tag{2.3.8.17}$$

Equating the entries, this produces system of linear equations,

$$a + b = w, y = b, x = c, z = c + d$$
 (2.3.8.18)

$$\implies a = w - y \tag{2.3.8.19}$$

$$b = y (2.3.8.20)$$

$$c = x (2.3.8.21)$$

$$d = z - x \tag{2.3.8.22}$$

In particular, there exists at least one solution regardless of the values of w, x, y, z. For example, consider the following matrix,

$$\begin{pmatrix} w & x \\ y & z \end{pmatrix} = \begin{pmatrix} 3 & 4 \\ -2 & 7 \end{pmatrix} \tag{2.3.8.23}$$

Here, a = 5, b = -2, c = 4, d = 3. Using (2.3.8.16), we get

$$5\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} - 2\begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} + 4\begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + 3\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 3 & 4 \\ -2 & 7 \end{pmatrix}$$
(2.3.8.24)

Hence
$$\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$
 forms

the basis for the given space V.

2.3.9. Let **V** be a vector space over a subfield **F** of complex numbers. Suppose α , β and γ are linearly independent vectors in **V**. Prove that $(\alpha+\beta),(\beta+\gamma)$ and $(\gamma+\alpha)$ are linearly independent.

Solution: Let α , β and γ be three n× 1 dimensional vectors. We need to prove that,

$$(\alpha + \beta \quad \beta + \gamma \quad \gamma + \alpha) \mathbf{x} = 0 \qquad (2.3.9.1)$$

will only have a trivial solution. The above equation can be written as

$$(\alpha \quad \beta \quad \gamma) \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \mathbf{x} = 0$$
 (2.3.9.2)

$$\mathbf{x}^T \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} \alpha^T \\ \boldsymbol{\beta}^T \\ \boldsymbol{\gamma}^T \end{pmatrix} = 0 \qquad (2.3.9.3)$$

Since, α , β and γ are independent.

$$\mathbf{x}^T \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} = 0 \tag{2.3.9.4}$$

In the above equation we can see that the 3×3 matrix has linearly independent rows and hence will have a trivial solution. So, **x** is a zero vector. Hence, $(\alpha+\beta)$, $(\beta+\gamma)$ and $(\gamma+\alpha)$ are linearly independent.

2.3.10. Prove that the space of all **m**x**n** matrices over the field **F** has dimension mn, by exhibiting a basis for this space.

Solution: Let M be the space of all $\mathbf{m} \times \mathbf{n}$ matrices. Let, $\mathbf{M}_{ij} \in \mathbf{M}$ be,

$$\mathbf{M}_{ij} = \begin{cases} 0 & m \neq i, n \neq j \\ 1 & m = i, n = j \end{cases}$$

$$(2.3.10.1) \text{ mm.}$$

$$(2.3.10.1) \text{ Suppose a G and were linearly independent.}$$

For example,

$$\mathbf{M}_{12} = \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 \end{pmatrix}_{mxn}$$
 (2.3.10.2)

Let $A \in M$ given as,

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \cdots & a_{mn} \end{pmatrix}_{\mathbf{m} \times \mathbf{m}}$$
(2.3.10.4)

Now clearly,

$$\mathbf{a}_{11} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}$$

$$(2.3.10.5)$$

$$\implies$$
 a₁₁ = **AM**₁₁ (2.3.10.6)

$$\therefore \mathbf{A} = \sum_{i=1}^{m} \sum_{j=1}^{n} a_{ij} M_{ij}$$
 (2.3.10.7)

 \implies **M**_{ij} span **M**. Also from the above equation **A**= 0 if and only if all elements are zero, that is,

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} = \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}$$
(2.3.10.8)

$$\implies a_{ij} = 0 \tag{2.3.10.9}$$

Hence, \mathbf{M}_{ij} are linearly independent as well. Hence, \mathbf{M}_{ij} constitutes a basis for \mathbf{M} . and number of elements in basis are mn. Hence dimension of space of all mxn matrices \mathbf{M} is mn.

Let V be a vector space over the field $F = \{0, 1\}$. Suppose α , β and γ are linearly independent vectors in V. Comment on $(\alpha + \beta)$, $(\beta + \gamma)$ and $(\gamma + \alpha)$

Solution: The addition of elements in the field **F** is defined as,

$$0 + 0 = 0$$

1 + 1 = 0 (2.3.11.1)

A set are vectors $\{v_1, v_2, v_3\}$ are linearly independent if

$$a\mathbf{v_1} + b\mathbf{v_2} + c\mathbf{v_3} = 0$$
 (2.3.11.2)

has only one trivial solution

$$a = b = c = 0$$
 (2.3.11.3)

Now,

$$a(\alpha + \beta) + b(\beta + \gamma) + c(\gamma + \alpha) = 0$$

$$(2.3.11.4)$$

$$\implies (a+c)\alpha + (a+b)\beta + (b+c)\gamma = 0$$

$$(2.3.11.5)$$

Writing (2.3.11.5) in matrix form,

$$(\alpha \quad \beta \quad \gamma) \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \mathbf{x} = 0$$
 (2.3.11.6)

where,

$$\mathbf{x} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$

$$\mathbf{x}^T \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = 0 \tag{2.3.11.7}$$

Since α , β and γ are linearly independent vectors,

$$\mathbf{x}^{T} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \end{pmatrix} \tag{2.3.11.8}$$

Transposing on both sides,

$$\begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \mathbf{x} = 0 \tag{2.3.11.9}$$

By using the properties from (2.3.11.1) and reducing (2.3.11.9) to row echelon form,

$$\begin{pmatrix}
1 & 0 & 1 \\
1 & 1 & 0 \\
0 & 1 & 1
\end{pmatrix}
\xrightarrow{R_2 \leftarrow R_1 + R_2}
\begin{pmatrix}
1 & 0 & 1 \\
0 & 1 & 1 \\
0 & 1 & 1
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_2 + R_3}
\begin{pmatrix}
1 & 0 & 1 \\
0 & 1 & 1 \\
0 & 0 & 0
\end{pmatrix}$$
(2.3.11.10)

Expressing (2.3.11.10) as a linear combination

of vectors,

$$a \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + b \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + c \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\implies \begin{pmatrix} a+c \\ b+c \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\implies a+c=0; \quad b+c=0 \qquad (2.3.11.11)$$

The solutions to (2.3.11.11) are,

$$a = b = c = 0;$$
 $a = b = c = 1$ (2.3.11.12)

Since there is no trivial solution, $(\alpha + \beta)$, $(\beta + \gamma)$ and $(\gamma + \alpha)$ are linearly dependent

2.3.12. Let **V** be the set of real numbers.Regard **V** as a vector space over the field of rational numbers, with usual operations.Prove that this vector space is not finite-dimensional.

Solution: Given V is a vector space over field \mathbb{Q} (rational numbers)

It is finite dimensional with dimensionality n if every vector \mathbf{v} in \mathbf{V} can be written as

$$\mathbf{v} = \sum_{i=0}^{n-1} c_i \alpha_i \qquad (2.3.12.1)$$

where
$$c_i \in \mathbb{O}$$
 (2.3.12.2)

and
$$\mathbf{B} = \{\alpha_0, \alpha_1, \dots, \alpha_{n-1}\}\$$
 (2.3.12.3)

is the basis with linearly independent α_i that is, basis is the largest set with linearly independent vectors from V

Consider the set of vectors $\{1, x\}$, where x is irrational.

Assume there exists non zero $\beta_0, \beta_1 \in \mathbb{Q}$ such that

$$\beta_0 + \beta_1 x = 0 \tag{2.3.12.4}$$

$$\implies x = -\frac{\beta_0}{\beta_1} \tag{2.3.12.5}$$

But x is irrational and $-\frac{\beta_0}{\beta_1}$ is rational so (2.3.12.5) can't be possible so $\beta_0, \beta_1 = 0$ Hence $\{1, x\}$ are independent. Similarly for the set $\{1, x, x^2\}$

for $\beta_0, \beta_1, \beta_2 \in \mathbb{Q}$

$$\beta_0 + \beta_1 x + \beta_2 x^2 = 0 (2.3.12.6)$$

 $\beta_1 x + \beta_2 x^2$ is irrational and β_0 is rational.

Therefore

$$\beta_0 = 0 \qquad (2.3.12.7)$$
and $\beta_1 x + \beta_2 x^2 = 0, (x \neq 0) \qquad (2.3.12.8)$

$$\implies \beta_1 + \beta_2 x = 0 \qquad (2.3.12.9)$$

$$\implies \beta_1, \beta_2 = 0 \qquad (2.3.12.10)$$

$$\therefore \beta_0 + \beta_1 x + \beta_2 x^2 = 0 \qquad (2.3.12.11)$$

$$\iff \beta_0, \beta_1, \beta_2 = 0 \qquad (2.3.12.12)$$

Hence $\{1, x, x^2\}$ are independent

for
$$\beta_0, \beta_1, \beta_2, \dots, \beta_n \in \mathbb{Q}$$
 (2.3.12.13)
 $\beta_0 + \beta_1 x + \beta_2 x^2 + \dots + \beta_n x^n = 0$ (2.3.12.14)
 $\iff \beta_0, \beta_1, \beta_2, \dots, \beta_n = 0$ (2.3.12.15)

To prove this for the set $A = \{1, x, x^2, \dots, x^{n+1}\}$

for
$$\beta_0, \beta_1, \beta_2, \dots, \beta_n, \beta_{n+1} \in \mathbb{Q}$$

(2.3.12.16)

$$\beta_0 + \beta_1 x + \dots + \beta_n x^n + \beta_{n+1} x^{n+1} = 0$$
(2.3.12.17)

Comparing to (2.3.12.7) and (2.3.12.8)

$$\beta_0 = 0$$
 (2.3.12.18)
 $\beta_1 + \beta_2 x + \dots + \beta_{n+1} x^n = 0$ (2.3.12.19)

Comparing with (2.3.12.14), we have $\beta_1,\beta_2,\ldots,\beta_{n+1}=0$

$$\therefore \beta_0 + \beta_1 x + \dots + \beta_n x^n + \beta_{n+1} x^{n+1} = 0$$
(2.3.12.20)
$$\iff \beta_0, \beta_1, \beta_2, \dots, \beta_n, \beta_{n+1} = 0$$
(2.3.12.21)

Hence A has linearly independent vectors Let the set $\mathbf{B} = \{1, x, x^2, \dots, x^m\}$ be the largest linearly independent set in V and hence can form the basis leading to dimensionality m + 1But from induction, we have proved that $\{1, x, x^2, \dots, x^m, x^{m+1}\}$ is also independent which is a contradiction to dimensionality being m+1

Hence we deduce that the vector space V is not finite dimensional over the field Q

form a basis for \Re^4 . Find the coordinates of each of the standard basis vectors in the ordered basis $(\alpha_1 \ \alpha_2 \ \alpha_3 \ \alpha_4)$

Solution:

Theorem 2.2. Let **V** be an n-dimensional vector space over the field \mathbf{F} , and let β and β' be two ordered basis of V. Then, there is a unique, necessarily invertible, $n \times n$ matrix **P** with entries in F such that

a)
$$\begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \mathbf{P} \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$$

b) $\begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \mathbf{P}^{-1} \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$

for every vector α in **V**. The columns of **P** are given by

$$\mathbf{P_j} = [\alpha_j]_{\beta}$$
 $j = 1, 2, ..., n$ (2.4.1.3)

Firt, we need to show that the set of vectors $\alpha_1, \alpha_2, \alpha_3$ and α_4 are basis for \Re^4 . For, this we first show that α_1 , α_2 , α_3 and α_4 are linearly independent in \Re^4 and also they span \Re^4 . Consider,

$$\mathbf{A} = \begin{pmatrix} \alpha_1^T & \alpha_2^T & \alpha_3^T & \alpha_4^T \end{pmatrix} \tag{2.4.1.4}$$

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 4 & 2 \end{pmatrix} \tag{2.4.1.5}$$

Now,

$$\begin{pmatrix}
1 & 0 & 1 & 0 \\
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 1 & 4 & 2
\end{pmatrix}
\xrightarrow{r_2=r_2-r_1}
\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 1 & 4 & 2
\end{pmatrix}
\qquad (2.4.1.6)$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 0 & -1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 1 & 4 & 2
\end{pmatrix}
\xrightarrow{r_2 \leftrightarrow r_3}
\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & -1 & 0 \\
0 & 1 & 4 & 2
\end{pmatrix}
\qquad (2.4.1.7)$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & -1 & 0 \\
0 & 0 & -1 & 0 \\
0 & 0 & 4 & 2
\end{pmatrix}
\xrightarrow{r_4=r_4-r_2}
\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 4 & 2
\end{pmatrix}
\qquad (2.4.1.8)$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & -1 & 0 \\
0 & 0 & 4 & 2
\end{pmatrix}
\xrightarrow{r_4=r_4-4r_3}
\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 2
\end{pmatrix}
\qquad (2.4.1.10)$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 2
\end{pmatrix}
\xrightarrow{r_4=r_4-4r_3}
\begin{pmatrix}
1 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 2
\end{pmatrix}
\qquad (2.4.1.11)$$

$$\begin{pmatrix}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 2
\end{pmatrix}
\xrightarrow{r_4=r_4-r_2}
\begin{pmatrix}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 2
\end{pmatrix}
\qquad (2.4.1.12)$$

(2.4.1.12) is the row reduced echelon form of **A** and since it is identity matrix of order 4, we say that vectors α_1 , α_2 , α_3 and α_4 are linearly independent and their column space is \Re^4 which means vectors α_1 , α_2 , α_3 and α_4 span \Re^4 . Hence, vectors α_1 , α_2 , α_3 and α_4 form a basis for \Re^4 .

Now, we use theorem (2.2), and if we calculate

the inverse of

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 4 & 2 \end{pmatrix} \tag{2.4.1.13}$$

then the columns of A^{-1} will give the coefficients to write the standard basis vectors in terms of $\alpha'_i s$. We try to find the inverse of A by row-reducing the augumented matrix.

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 4 & 2 & 0 & 0 & 0 & 1 \end{pmatrix}$$
 (2.4.1.14)

Now, we solve for A^{-1} as follows

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 1 & 4 & 2 & 0 & 0 & 0 & 1
\end{pmatrix}
\xrightarrow{r_2=r_2-r_1}$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 1 & 4 & 2 & 0 & 0 & 0 & 1
\end{pmatrix}$$
(2.4.1.15)

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 1 & 4 & 2 & 0 & 0 & 0 & 1
\end{pmatrix}
\xrightarrow{r_2 \leftrightarrow r_3}$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & 4 & 2 & 0 & 0 & 0 & 1
\end{pmatrix}$$
(2.4.1.16)

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & 4 & 2 & 0 & 0 & 0 & 1
\end{pmatrix}
\xrightarrow{r_4=r_4-r_2}$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\
0 & 0 & 4 & 2 & 0 & 0 & -1 & 1
\end{pmatrix}$$
(2.4.1.17)

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\
0 & 0 & 4 & 2 & 0 & 0 & -1 & 1
\end{pmatrix}
\xrightarrow{r_3 = -r_3}$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 4 & 2 & 0 & 0 & -1 & 1
\end{pmatrix}$$
(2.4.1.18)

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 4 & 2 & 0 & 0 & -1 & 1
\end{pmatrix}
\xrightarrow{r_4 = r_4 - 4r_3}$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 0 & 2 & -4 & 4 & -1 & 1
\end{pmatrix}$$
(2.4.1.19)

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 0 & 2 & -4 & 4 & -1 & 1
\end{pmatrix}
\xrightarrow{r_1 = r_1 - r_3}$$

$$\begin{pmatrix}
1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 0 & 2 & -4 & 4 & -1 & 1
\end{pmatrix}$$
(2.4.1.20)

$$\begin{pmatrix}
1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 0 & 2 & -4 & 4 & -1 & 1
\end{pmatrix}
\xrightarrow{r_4 = \frac{r_4}{2}}$$

$$\begin{pmatrix}
1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 0 & 1 & -2 & 2 & -\frac{1}{2} & \frac{1}{2}
\end{pmatrix}$$
(2.4.1.21)

Thus, by (2.4.1.21), we have

$$\mathbf{A}^{-1} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & -1 & 0 & 0 \\ -2 & 2 & \frac{-1}{2} & \frac{1}{2} \end{pmatrix}$$
 (2.4.1.22) 2.4.3. Let $\mathbf{B} = \begin{pmatrix} \alpha_1 & \alpha_2 & \alpha_3 \end{pmatrix}$ be the ordered basis for R^3 consisting of

Now, let $\mathbf{e_1} = (1 \ 0 \ 0 \ 0), \ \mathbf{e_2} = (0 \ 1 \ 0 \ 0),$ $\mathbf{e_3} = (0 \ 0 \ 1 \ 0) \text{ and } \mathbf{e_4} = (0 \ 0 \ 0 \ 1) \text{ be}$ the standard basis for \Re^4 . Hence,

$$\mathbf{e_1} = \alpha_3 - 2\alpha_4 \tag{2.4.1.23}$$

$$\mathbf{e_2} = \alpha_1 - \alpha_3 + 2\alpha_4$$
 (2.4.1.24)

$$\mathbf{e_3} = \alpha_2 - \frac{1}{2}\alpha_4 \tag{2.4.1.25}$$

$$\mathbf{e_4} = \frac{1}{2}\alpha_4 \tag{2.4.1.26}$$

2.4.2. Find the coordinate matrix of the vector $(1 \ 0 \ 1)$ in the basis of C^3 consisting of the vectors $\begin{pmatrix} 2i & 1 & 0 \end{pmatrix}$, $\begin{pmatrix} 2 & -1 & 1 \end{pmatrix}$, (0 1 + i 1 - i) in that order. **Solution:**

$$\begin{pmatrix} 1 & 0 & 1 \end{pmatrix} = \alpha_1 \begin{pmatrix} 2i & 1 & 0 \end{pmatrix} + \alpha_2 \begin{pmatrix} 2 & -1 & 1 \end{pmatrix} + \alpha_3 \begin{pmatrix} 0 & 1+i & 1-i \end{pmatrix}$$
(2.4.2.1)

$$\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 2i & 2 & 0 \\ 1 & -1 & 1+i \\ 0 & 1 & 1-i \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix}$$
 (2.4.2.2)

Now we find α_i by row reducing augmented

$$\begin{pmatrix} 2i & 2 & 0 & 1 \\ 1 & -1 & 1+i & 0 \\ 0 & 1 & 1-i & 1 \end{pmatrix} \xrightarrow{R_1 \to R_2} \begin{pmatrix} 1 & -1 & 1+i & 0 \\ 0 & 2+2i & 2-2i & 1 \\ 0 & 1 & 1-i & 1 \end{pmatrix}$$
 (2.4.2.3)

$$\stackrel{R_2 \leftarrow R_2/2}{\underset{R_3 \leftarrow R_3 - R_2}{\longleftarrow}} \begin{pmatrix}
1 & -1 & 1 + i & 0 \\
0 & 1 + i & 1 - i & \frac{1}{2} \\
0 & -i & 0 & \frac{1}{2}
\end{pmatrix} (2.4.2.4)$$

Therefore the coordinate matrix of the vector

$$\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = \begin{pmatrix} \frac{-1-i}{2} \\ \frac{1}{2} \\ \frac{3+i}{4} \end{pmatrix}$$
 (2.4.2.5)

$$\alpha_1 = \begin{pmatrix} 1 & 0 & -1 \end{pmatrix}, \alpha_2 = \begin{pmatrix} 1 & 1 & 1 \end{pmatrix}, \alpha_3 = \begin{pmatrix} 1 & 0 & 0 \end{pmatrix}.$$

What are the coordinates of vector $(a \ b \ c)$ in the ordered basis **B**?

Solution: Given

$$\mathbf{B} = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ -1 & 1 & 0 \end{pmatrix} \tag{2.4.3.1}$$

be the ordered basis for \mathbb{R}^3 , then the coordinates 2.4.4. Let **W** be the subspace of \mathbb{C}^3 spanned by $\alpha_1 = 0$ of vector,

$$\alpha = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{2.4.3.2}$$

in the ordered basis R^3 is the vector,

$$[\alpha]_{\mathbf{B}} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \tag{2.4.3.3}$$

hence

$$x_1\alpha_1 + x_2\alpha_2 + x_3\alpha_3 = \alpha \tag{2.4.3.4}$$

substituting (2.4.3.1) and (2.4.3.2) in (2.4.3.4)

$$\begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ -1 & 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$
 (2.4.3.5)

augmented matrix form

$$\begin{pmatrix} 1 & 1 & 1 & a \\ 0 & 1 & 0 & b \\ -1 & 1 & 0 & c \end{pmatrix}$$
 (2.4.3.6)

converting above matrix into row reduced echelon form

$$\begin{pmatrix}
1 & 1 & 1 & a \\
0 & 1 & 0 & b \\
-1 & 1 & 0 & c
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 + R_1}
\begin{pmatrix}
1 & 1 & 1 & a \\
0 & 1 & 0 & b \\
0 & 2 & 1 & c + a
\end{pmatrix}$$

$$(2.4.3.7)$$

$$\xrightarrow{R_3 \leftarrow R_3 - 2R_2}
\begin{pmatrix}
1 & 1 & 1 & a \\
0 & 1 & 0 & b \\
0 & 0 & 1 & a - 2b + c
\end{pmatrix}$$

$$(2.4.3.8)$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_2}
\begin{pmatrix}
1 & 0 & 1 & a - b \\
0 & 1 & 0 & b \\
0 & 0 & 1 & a - 2b + c
\end{pmatrix}$$

$$(2.4.3.9)$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_3}
\begin{pmatrix}
1 & 0 & 0 & b - c \\
0 & 1 & 0 & b \\
0 & 0 & 1 & a - 2b + c
\end{pmatrix}$$

 \therefore The coordinates of α w.r.t **B** is

$$[\alpha]_{\mathbf{B}} = \begin{pmatrix} b - c \\ b \\ a - 2b + c \end{pmatrix} \tag{2.4.3.11}$$

$$\begin{pmatrix} 1 \\ 0 \\ i \end{pmatrix} \text{ and } \alpha_2 = \begin{pmatrix} 1 \\ i \\ 1+i \end{pmatrix}.$$

a) Show that α_1 and α_2 form a basis for **W**.

b) Show that the vectors $\beta_1 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ and $\beta_2 = \begin{pmatrix} 1 \\ i \\ 1+i \end{pmatrix}$ are in **W** and form another basis for **W**.

c) What are the coordinates of α_1 and α_2 in the ordered basis $\{\beta_1, \beta_2\}$ for **W**.

Solution:

a) It is given that α_1 and α_2 span W. For α_1 and α_2 to be the basis for W they must be linearly independent. Let

$$S_1 = {\alpha_1, \alpha_2} = \left\{ \begin{pmatrix} 1\\0\\i \end{pmatrix}, \begin{pmatrix} 1+i\\1\\-1 \end{pmatrix} \right\} \quad (2.4.4.1)$$

Using row reduction on matrix $\mathbf{A} = (\alpha_1 \ \alpha_2)$

$$\begin{pmatrix}
1 & 1+i \\
0 & 1 \\
i & -1
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 - iR_1}
\begin{pmatrix}
1 & 1+i \\
0 & 1 \\
0 & -i
\end{pmatrix}
(2.4.4.2)$$

$$\xrightarrow{R_3 \leftarrow R_3 + iR_2}
\begin{pmatrix}
1 & 1+i \\
0 & 1 \\
0 & 0
\end{pmatrix}
(2.4.4.3)$$

$$\xrightarrow{R_1 \leftarrow R_1 - (i+1)R_2}
\begin{pmatrix}
1 & 0 \\
0 & 1 \\
0 & 0
\end{pmatrix}
(2.4.4.4)$$

Since A is a full-rank matrix the column vectors are linearly independent. Therefore $S_1 = \{\alpha_1, \alpha_2\}$ is a basis set for **W**.

b)

$$\beta_1 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \tag{2.4.4.5}$$

$$\beta_2 = \begin{pmatrix} 1\\i\\1+i \end{pmatrix} \tag{2.4.4.6}$$

Since column vectors of $\mathbf{A} = \begin{pmatrix} \alpha_1 & \alpha_2 \end{pmatrix}$ are basis for **W** and if β_1 and $\beta_2 \in \mathbf{W}$ there exist a unique solution x such that

$$(\alpha_1 \quad \alpha_2) \mathbf{x} = (\beta_1 \quad \beta_2)$$
 (2.4.4.7)

Using row reduction on augmented matrix

$$\begin{pmatrix}
1 & 1+i & | & 1 & 1 \\
0 & 1 & | & 1 & i \\
i & -1 & | & 0 & 1+i
\end{pmatrix} (2.4.4.8)$$

$$\stackrel{R3 \leftarrow R_3 - iR - 1}{\longleftrightarrow} \begin{pmatrix}
1 & 1+i & | & 1 & 1 \\
0 & 1 & | & 1 & i \\
0 & -i & | & -i & 1
\end{pmatrix} (2.4.4.9)$$

$$\stackrel{R_3 \leftarrow R_3 + iR_2}{\longleftrightarrow} \begin{pmatrix}
1 & 1+i & | & 1 & 1 \\
0 & 1 & | & 1 & i \\
0 & 0 & | & 0 & 0
\end{pmatrix}$$

$$\stackrel{R_1 \leftarrow R_1 - (i+1)R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & | & -i & 2-i \\
0 & 1 & | & 1 & i \\
0 & 0 & | & 0 & 0
\end{pmatrix}$$

$$\stackrel{(2.4.4.11)}{\Longrightarrow} \mathbf{x} = \begin{pmatrix}
-i & 2-i \\
1 & i
\end{pmatrix}$$

Therefore β_1 and $\beta_2 \in \mathbf{W}$. Consider

$$S_2 = \{\beta_1, \beta_2\} = \left\{ \begin{pmatrix} 1\\1\\0 \end{pmatrix}, \begin{pmatrix} 1\\i\\1+i \end{pmatrix} \right\} \quad (2.4.4.13)$$

and also let

$$\mathbf{B} = \begin{pmatrix} 1 & 1 \\ 1 & i \\ 0 & 1+i \end{pmatrix} \tag{2.4.4.14}$$

(2.4.4.12)

Using row reduction on matrix **B**

$$\begin{pmatrix} 1 & 1 \\ 1 & i \\ 0 & 1+i \end{pmatrix} \xrightarrow{R_2 \leftarrow R_2 - R_1} \begin{pmatrix} 1 & 1 \\ 0 & i-1 \\ 0 & 1+i \end{pmatrix} \quad (2.4.4.15)$$

$$\xrightarrow{R_2 \leftarrow \frac{R_2}{i-1}} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 0 & 1+i \end{pmatrix} \quad (2.4.4.16)$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_2} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \quad (2.4.4.17)$$

Since **B** is a full rank matrix the column vectors are linearly independent.

Let α be any vector in the subspace W, then it can be expressed as span $\{\alpha_1, \alpha_2\}$ i.e

$$\alpha = (\alpha_1 \quad \alpha_2) \mathbf{x_1} = \mathbf{A} \mathbf{x_1} \tag{2.4.4.18}$$

 $S_2 = \{\beta_1, \beta_2\}$ spans **W** if any vector $\alpha \in \mathbf{W}$ can be expressed as

$$\alpha = (\beta_1, \beta_2) \mathbf{x_2} = \mathbf{B} \mathbf{x_2} \tag{2.4.4.19}$$

From (2.4.4.18) and (2.4.4.19) we conclude

$$\mathbf{B}\mathbf{x}_2 = \mathbf{A}\mathbf{x}_1$$
 (2.4.4.20)

$$\implies \mathbf{x}_2 = \mathbf{B}^{-1} \mathbf{A} \mathbf{x}_1 \tag{2.4.4.21}$$

Therefore from (2.4.4.21) $\mathbf{x_2}$ exists if **B** is invertible. From (2.4.4.17) we conclude x_2 exists and hence any vector $\alpha \in \mathbf{W}$ can be expressed as span $\{\beta_1, \beta_2\}$. Therefore $\{\beta_1, \beta_2\}$ is basis for **W**.

c) Since $\alpha_1, \alpha_2 \in \mathbf{W}$ and $\{\beta_1, \beta_2\}$ are ordered basis for W there must exist unique value of x such that

$$(\beta_1 \quad \beta_2) \mathbf{x} = (\alpha_1 \quad \alpha_2)$$
 (2.4.4.22)

Using row reduction on (2.4.4.22) we get,

$$\begin{pmatrix} 1 & 1 & | & 1 & 1+i \\ 1 & i & | & 0 & 1 \\ 0 & 1+i & | & i & -1 \end{pmatrix}$$
(2.4.4.23)

$$\xrightarrow{R_2 \leftarrow R_2 - R_1} \begin{pmatrix} 1 & 1 & | & 1 & 1 + i \\ 0 & i - 1 & | & -1 & -i \\ 0 & 1 + i & | & i & -1 \end{pmatrix}$$
(2.4.4.24)

$$\xrightarrow{R_2 \leftarrow \frac{R_2}{i-1}} \begin{pmatrix} 1 & 1 & | & 1 & 1+i \\ 0 & 1 & | & \frac{1+i}{2} & \frac{-1+i}{2} \\ 0 & 1+i & | & i & -1 \end{pmatrix}$$

$$(2.4.4.25)$$

$$\stackrel{R_3 \leftarrow R_3 - (i+1)R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 1 & | & 1 & 1+i \\
0 & 1 & | & \frac{1+i}{2} & \frac{-1+i}{2} \\
0 & 0 & | & 0 & 0
\end{pmatrix}$$
(2.4.4.26)

$$\stackrel{R_1 \leftarrow R_2 - R_1}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & | & \frac{1-i}{2} & \frac{3+i}{2} \\ 0 & 1 & | & \frac{1+i}{2} & \frac{-1+i}{2} \\ 0 & 0 & | & 0 & 0 \end{pmatrix}$$
(2.4.4.27)

$$\implies \mathbf{x} = \frac{1}{2} \begin{pmatrix} 1 - i & 3 + i \\ 1 + i & -1 + i \end{pmatrix}$$
 (2.4.4.28)

Thus the column vectors of (2.4.4.28) are corresponding coordinates of α_1 and α_2 in ordered basis $\{\beta_1, \beta_2\}$.

2.4.5. let
$$\alpha = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$
 and $\beta = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ be vectors in \mathbb{R}^2 such that

 $x_1y_1 + x_2y_2 = 0$; $x_1^2 + x_2^2 = y_1^2 + y_2^2 = 1$. Proove that $\beta = \{\alpha, \beta\}$ is a basis of \mathbb{R}^2 . Find the coordinates of the vector (a, b) in the ordered basis $\beta = \{\alpha, \beta\}$. (The conditions on α and β say, geometrically, that α and β are perpendicular and each has length 1).

Solution: we need to show that α and β are linearly independent in order to proove that $\beta = \{\alpha, \beta\}$ is a basis of \mathbb{R}^2 .

Given in the question are:

$$\alpha^T \beta = 0 \tag{2.4.5.1}$$

$$\|\alpha\|^2 = \|\beta\|^2 = 1$$
 (2.4.5.2)

Let,

$$\mathbf{A} = \begin{pmatrix} \alpha & \beta \end{pmatrix} = \begin{pmatrix} x_1 & y_1 \\ x_2 & y_2 \end{pmatrix} \tag{2.4.5.3}$$

then,

$$\mathbf{A}^{T}\mathbf{A} = \begin{pmatrix} ||\alpha||^{2} & \alpha^{T}\beta \\ \alpha^{T}\beta & ||\beta||^{2} \end{pmatrix}$$
 (2.4.5.4)

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{2.4.5.5}$$

$$\therefore \mathbf{A}^T \mathbf{A} = \mathbf{I} \tag{2.4.5.6}$$

Inverse of **A** exist. A^T is the inverse of **A**. Thus, the columns of **A** are linearly independent i.e, α and β are linearly independent.

Hence, $\beta = \{\alpha, \beta\}$ is a basis of \mathbb{R}^2 .

To, find the coordinates of the vector (a, b) in the ordered basis $\beta = \{\alpha, \beta\}$.

$$(\alpha \quad \beta) \begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix}$$
 (2.4.5.7)

$$\mathbf{A}^T \mathbf{A} \begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = \mathbf{A}^T \begin{pmatrix} a \\ b \end{pmatrix} \tag{2.4.5.8}$$

$$\begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = \mathbf{A}^T \begin{pmatrix} a \\ b \end{pmatrix}$$
 (2.4.5.9)

$$\begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = \begin{pmatrix} x_1 & x_2 \\ y_1 & y_2 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} \tag{2.4.5.10}$$

- 2.4.6. Let **V** be the vector space over the complex numbers of all functions from **R** into **C**, i.e. the space of all complex-valued functions on the real line. Let $f_1(x) = 1$, $f_2(x) = e^{ix}$, $f_3(x) = e^{-ix}$.
 - (a) Prove that f_1 , f_2 , and f_3 are linearly dependent.
 - (b) Let $g_1(x) = 1$, $g_2(x) = \cos(x)$, $g_3(x) = \sin(x)$. Find an invertible 3×3 matrix P such that $g_i = \sum_{i=1}^3 P_{ii} f_i$

Solution:

(a) To check for independence, lets represent the function in a polynomial format as

$$\alpha f_1 + \beta f_2 + \gamma f_3 = 0 \tag{2.4.6.1}$$

$$\alpha + \beta e^{ix} + \gamma e^{-ix} = 0 (2.4.6.2)$$

Multiply the whole equation with e^{ix}

$$\beta(e^{ix})^2 + \alpha e^{ix} + \gamma = 0$$
 (2.4.6.3)

Let $y = e^{ix}$

$$\beta y^2 + \alpha y + \gamma = 0 {(2.4.6.4)}$$

The above quadratic polynomial in y can be zero for atmost two values of y. But,

$$y = e^{ix} \quad x \in \mathbf{R} \tag{2.4.6.5}$$

So (2.4.6.4) cannot be zero for all $y = e^{ix}$. Which implies there is a contradiction.

Then, the only case when (2.4.6.4) gets satisfied is

$$\alpha = \beta = \gamma = 0 \tag{2.4.6.6}$$

Therefore, f_1 , f_2 , f_3 are linearly independent.

(b) We need to find the coordinates of vectors g_i where i = 1, 2, 3 in ordered basis

$$B = \begin{pmatrix} f_1 & f_2 & f_3 \end{pmatrix}$$
 (2.4.6.7)

It is given that $g_1 = 1$, which can be written as

$$g_1 = f_1 \qquad (2.4.6.8)$$

$$\implies g_1 = \begin{pmatrix} f_1 & f_2 & f_3 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \qquad (2.4.6.9)$$

We can use the following identities:-

$$\cos(x) = \frac{1}{2}e^{ix} + \frac{1}{2}e^{-ix}$$
 (2.4.6.10)

$$\sin(x) = \frac{1}{2i}e^{ix} - \frac{1}{2i}e^{-ix} \qquad (2.4.6.11)$$

Comparing equations (2.4.6.10) and (2.4.6.11) with f_2 , f_3 , we can write g_2 and g_3 as

$$g_2 = \frac{1}{2}f_2 + \frac{1}{2}f_3$$
 (2.4.6.12)

$$\implies g_2 = \begin{pmatrix} f_1 & f_2 & f_3 \end{pmatrix} \begin{pmatrix} 0 \\ \frac{1}{2} \\ \frac{1}{2} \end{pmatrix}$$
 (2.4.6.13)

$$g_3 = \frac{1}{2i}f_2 - \frac{1}{2i}f_3 \qquad (2.4.6.14)$$

$$\implies g_3 = \begin{pmatrix} f_1 & f_2 & f_3 \end{pmatrix} \begin{pmatrix} 0 \\ \frac{1}{2i} \\ \frac{-1}{2i} \end{pmatrix} \quad (2.4.6.15)$$

Therefore, the required matrix P is

$$P = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2i} \\ 0 & \frac{1}{2} & \frac{-1}{2i} \end{pmatrix}$$
 (2.4.6.16)

2.4.7. Let **V** be the real vector space of all polynomial functions from \mathbb{R} to \mathbb{R} of degree 2 or less, i.e, the space of all functions f of the form,

$$f(x) = c_0 + c_1 x + c_2 x^2$$

Let t be a fixed real number and define

$$g_1(x) = 1, g_2(x) = x + t, g_3(x) = (x + t)^2$$

Prove that $\beta = \{g1, g2, g3\}$ is a basis for V. If

$$f(x) = c_0 + c_1 x + c_2 x^2$$

what are the coordinates of f in the ordered basis β

Solution: We start by taking,

$$\mathbf{f} = \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} \tag{2.4.7.1}$$

Let's start by proving that \mathbf{g} is linearly independent.

$$\mathbf{g} = \mathbf{Bf} \tag{2.4.7.2}$$

where,

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 \\ t & 1 & 0 \\ t^2 & 2t & 1 \end{pmatrix} \tag{2.4.7.3}$$

Now,

$$\mathbf{v}^T \mathbf{g} = 0 \tag{2.4.7.4}$$

$$\implies \mathbf{v}^T \mathbf{B} \mathbf{f} = 0 \tag{2.4.7.5}$$

Since **f** is linearly independent,

$$\mathbf{v}^T \mathbf{B} = 0 \tag{2.4.7.6}$$

$$\mathbf{B}^T \mathbf{v} = 0 \tag{2.4.7.7}$$

Since \mathbf{B}^T is an upper triangular matrix with

non zero values in principal diagonal, it is 2.5 Summary of Row Equivalence invertible matrix and hence v will be zero vector. Now, Finding the inverse of \mathbf{B}^T

$$\begin{pmatrix}
1 & t & t^{2} & 1 & 0 & 0 \\
0 & 1 & 2t & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 0 & 1
\end{pmatrix}$$
(2.4.7.8) 2.5.1. Consider the vectors in \mathbb{R}^{4} defined by:
$$\alpha_{1} = \begin{pmatrix}
-1 \\ 0 \\ 1 \\ 2
\end{pmatrix}, \quad \alpha_{2} = \begin{pmatrix}
3 \\ 4 \\ -2 \\ 5
\end{pmatrix} \text{ and } \alpha_{3} = \begin{pmatrix}
1 \\ 4 \\ 0 \\ 9
\end{pmatrix}.$$
Find a system of homogeneous linear equation for which the space of solutions is example subspace of \mathbb{R}^{4} spanned by the given vectors.

$$\alpha_{1} = \begin{pmatrix}
-1 \\ 0 \\ 1 \\ 2
\end{pmatrix}, \quad \alpha_{2} = \begin{pmatrix}
3 \\ 4 \\ -2 \\ 5
\end{pmatrix} \text{ and } \alpha_{3} = \begin{pmatrix}
1 \\ 4 \\ 0 \\ 9
\end{pmatrix}.$$
Find a system of homogeneous linear equation for which the space of solutions is example subspace of \mathbb{R}^{4} spanned by the given vectors.

$$\alpha_{1} = \begin{pmatrix}
1 & 0 & 0 & 1 & -t & t^{2} \\
0 & 1 & 2t & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 0 & 1
\end{pmatrix}$$
Solution: A system of linear equation is homogeneous if all of the constant is homogeneous.

$$\xrightarrow{R_2 = R_2 - 2tR_3} \begin{pmatrix} 1 & 0 & 0 & 1 & -t & t^2 \\ 0 & 1 & 0 & 0 & 1 & -2t \\ 0 & 0 & 1 & 0 & 0 & 1 \end{pmatrix}$$

$$(2.4.7.11)$$

So,

$$(\mathbf{B}^T)^{-1} = \begin{pmatrix} 1 & -t & t^2 \\ 0 & 1 & -2t \\ 0 & 0 & 1 \end{pmatrix}$$
 (2.4.7.12)

Now, to find the coordinates,

$$f(x) = \mathbf{w}^T \mathbf{g} \tag{2.4.7.13}$$

So,

$$\mathbf{c}^T \mathbf{f} = \mathbf{w}^T \mathbf{g} \tag{2.4.7.14}$$

$$\mathbf{c}^T \mathbf{f} = \mathbf{w}^T \mathbf{B} \mathbf{f} \tag{2.4.7.15}$$

$$(\mathbf{c}^T - \mathbf{w}^T \mathbf{B})\mathbf{f} = 0 (2.4.7.16)$$

Since, **f** is linearly independent,

$$\mathbf{c}^T - \mathbf{w}^T \mathbf{B} = 0 \tag{2.4.7.17}$$

$$\mathbf{c}^T = \mathbf{w}^T \mathbf{B} \tag{2.4.7.18}$$

$$\mathbf{c}^T \mathbf{B}^{-1} = \mathbf{w}^T \tag{2.4.7.19}$$

$$(\mathbf{B}^{-1})^T \mathbf{c} = \mathbf{w} \tag{2.4.7.20}$$

Using (2.4.7.12) in (2.4.7.20)

$$\mathbf{w} = \begin{pmatrix} 1 & -t & t^2 \\ 0 & 1 & -2t \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} c_0 \\ c_1 \\ c_2 \end{pmatrix}$$
 (2.4.7.21)

$$\alpha_1 = \begin{pmatrix} -1\\0\\1\\2 \end{pmatrix}, \ \alpha_2 = \begin{pmatrix} 3\\4\\-2\\5 \end{pmatrix} \text{ and } \alpha_3 = \begin{pmatrix} 1\\4\\0\\9 \end{pmatrix}.$$

Find a system of homogeneous linear equations for which the space of solutions is exactly the subspace of \mathbb{R}^4 spanned by the given three vectors.

Solution: A system of linear equations is homogeneous if all of the constant terms are zero. It can be represented as,

$$\mathbf{AX} = 0 \tag{2.5.1.1}$$

Let **R** be a echelon matrix which is reduced to A. Then the systems AX = 0 and RX = 0 have the same solutions. Here,

$$\mathbf{A} = \begin{pmatrix} -1 & 0 & 1 & 2 \\ 3 & 4 & -2 & 5 \\ 1 & 4 & 0 & 9 \end{pmatrix} \tag{2.5.1.2}$$

By row reducing on A, we get:

$$\mathbf{A} = \begin{pmatrix} -1 & 0 & 1 & 2 \\ 3 & 4 & -2 & 5 \\ 1 & 4 & 0 & 9 \end{pmatrix} \xrightarrow{R_3 = R_3 - 2R_1 - R_2} \begin{pmatrix} -1 & 0 & 1 & 2 \\ 3 & 4 & -2 & 5 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(2.5.1.3)$$

$$\xrightarrow{R_2 = R_2 + 3R_1} \begin{pmatrix} -1 & 0 & 1 & 2 \\ 0 & 4 & 1 & 11 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(2.5.1.4)$$

$$\xrightarrow{R_1 = -R_1} \begin{pmatrix} 1 & 0 & -1 & -2 \\ 0 & 4 & 1 & 11 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

The bais vector is non zero vector which are given from 2.5.1.5,

$$\rho_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \\ -2 \end{pmatrix}, \rho_2 = \begin{pmatrix} 0 \\ 4 \\ 1 \\ 11 \end{pmatrix}$$
 (2.5.1.6)

 ρ_1 , ρ_2 forms the basis of the solution space. The subspace spanned by b_1 and b_2 is given as:

$$\left(\rho_1 \quad \rho_2\right) \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = \mathbf{X} \tag{2.5.1.7}$$

Using 2.5.1.7, we can write the augmented matrix as:

$$\begin{pmatrix} 1 & 0 & x_1 \\ 0 & 4 & x_2 \\ -1 & 1 & x_3 \\ -2 & 11 & x_4 \end{pmatrix} \xleftarrow{R_3 = 4R_3 + 4R_1 - R_2}$$

$$\begin{pmatrix} 1 & 0 & x_1 \\ 0 & 4 & x_2 \\ 0 & 0 & 4x_1 - x_2 + 4x_3 \\ -2 & 11 & x_4 \end{pmatrix} \longleftrightarrow_{R_4 = 4R_4 + 8R_1 - 11R_2}$$

(2.5.1.9)

$$\begin{pmatrix}
1 & 0 & x_1 \\
0 & 4 & x_2 \\
0 & 0 & 4x_1 - x_2 + 4x_3 \\
0 & 0 & 8x_1 - 11x_2 + 4x_4
\end{pmatrix}$$
(2.5.1.10

Using 2.5.1.10, The required homogeneous equation is given as:

$$\begin{pmatrix} 4 & -1 & 4 & 0 \\ 8 & -11 & 0 & 4 \end{pmatrix} \mathbf{X} = 0 \tag{2.5.1.11}$$

2.5.2. Let s < n and A an $s \times n$ matrix with entries in the field \mathbb{F} . Use Theorem 4 to show that there is a non-zero \mathbf{x} in $\mathbb{F}^{n \times 1}$ such that $\mathbf{A}\mathbf{x} = \mathbf{0}$.

Solution: Theorem 4:Let \mathbb{V} be a vector space which is spanned by a finite set of vectors $\beta_1, \beta_2, ..., \beta_m$. Then any independent set of vectors in \mathbb{V} is finite and contains no more than m elements. Let \mathbb{V} be a vector space spanned by $a_1, a_2, ..., a_n$, where $a_i, i=1,2,...,n$ are columns of matrix $\mathbf{A}_{s \times n}$.

$$\mathbf{A} = \begin{pmatrix} a_1 & a_2 & \dots & a_n \end{pmatrix} \tag{2.5.2.1}$$

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & & \vdots \\ a_{s1} & a_{s2} & \dots & a_{sn} \end{pmatrix}$$
 (2.5.2.2)

Let us take a_i , i=1,2,...,n as standard $s \times 1$ bases.

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & \dots & 0 & a_{1,s+1} & \dots & a_{1n} \\ 0 & 1 & \dots & 0 & a_{2,s+1} & \dots & a_{2n} \\ \vdots & & \vdots & & \vdots & & \vdots \\ 0 & 0 & \dots & 1 & a_{s,s+1} & \dots & a_{sn} \end{pmatrix}$$
(2.5.2.3)

From (2.5.2.3), it is clear that

$$dim(col(A)) \le s \tag{2.5.2.4}$$

$$\implies rank(A) \le s$$
 (2.5.2.5)

Now, from rank-nullity theorem,

$$rank(A) + nullity(A) = n$$
 (2.5.2.6)

$$nullity(A) = n - rank(A)$$
 (2.5.2.7)

$$\implies nullity(A) > 0$$
 (2.5.2.8)

From equation (2.5.2.8) it is clear that there will be a non zero \mathbf{x} such that $\mathbf{A}\mathbf{x} = \mathbf{0}$

2.5.3. Let

$$\alpha_1 = \begin{pmatrix} 1 & 1 & -2 & 1 \end{pmatrix}^T$$
 (2.5.3.1)

$$\alpha_2 = \begin{pmatrix} 3 & 0 & 4 & -1 \end{pmatrix}^T$$
 (2.5.3.2)

$$\alpha_3 = \begin{pmatrix} -1 & 2 & 5 & 2 \end{pmatrix}^T$$
 (2.5.3.3)

Let

$$\alpha = \begin{pmatrix} 4 & -5 & 9 & -7 \end{pmatrix}^T \tag{2.5.3.4}$$

$$\beta = \begin{pmatrix} 3 & 1 & -4 & 4 \end{pmatrix}^T \tag{2.5.3.5}$$

$$\gamma = \begin{pmatrix} -1 & 1 & 0 & 1 \end{pmatrix}^T \tag{2.5.3.6}$$

- a) Which of the vectors α , β , γ are in the subspace of \mathbb{R}^4 spanned by α_i ?
- b) Which of the vectors α , β , γ are in the subspace of \mathbb{C}^4 spanned by α_i ?
- c) Does this suggest a theorem?

Solution:

a) The linear combination of α_i for i = 1, 2, 3 spans subspace S. We can write,

$$c_{1} \begin{pmatrix} 1\\1\\-2\\1 \end{pmatrix} + c_{2} \begin{pmatrix} 3\\0\\4\\-1 \end{pmatrix} + c_{3} \begin{pmatrix} -1\\2\\5\\2 \end{pmatrix} = \text{span(S)}$$
(2.5.3.7)

where c_1, c_2, c_3 are scalars. Vectors in matrix

form is given by

$$\mathbf{A} = \begin{pmatrix} 1 & 3 & -1 \\ 1 & 0 & 2 \\ -2 & 4 & 5 \\ 1 & -1 & 2 \end{pmatrix} \tag{2.5.3.8}$$

We can observe that the columns of matrix **A** formed by vectors α_i are independent as the rank of matrix is 3. Hence α_i forms basis for subspace S.

i) Checking for α : To check if a solution exists for $AX = \alpha$. The corresponding agumented matrix can be written as,

$$(\mathbf{A} \quad \alpha) = \begin{pmatrix} 1 & 3 & -1 & 4 \\ 1 & 0 & 2 & -5 \\ -2 & 4 & 5 & 9 \\ 1 & -1 & 2 & -7 \end{pmatrix}$$
 (2.5.3.9)

On performing row-reduction on (2.5.3.9),

As Rank($(\mathbf{A} \ \alpha)$)=Rank((\mathbf{A}))=3, the vector α can be represented as linear combination of α_i . From equation (2.5.3.10), we can write

$$-3\begin{pmatrix} 1\\1\\-2\\1 \end{pmatrix} + 2\begin{pmatrix} 3\\0\\4\\-1 \end{pmatrix} - 1\begin{pmatrix} -1\\2\\5\\2 \end{pmatrix} = \begin{pmatrix} 4\\-5\\9\\-7 \end{pmatrix}$$
(2.5.3.11)

Hence α is in the subspace S.

ii) Checking for β : To check if a solution exists for $\mathbf{AX} = \beta$. The corresponding agumented matrix can be written as,

$$(\mathbf{A} \quad \beta) = \begin{pmatrix} 1 & 3 & -1 & 3 \\ 1 & 0 & 2 & 1 \\ -2 & 4 & 5 & -4 \\ 1 & -1 & 2 & 4 \end{pmatrix}$$
 (2.5.3.12)

On performing row-reduction on

(2.5.3.12),

$$(\mathbf{A} \quad \beta) = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$
 (2.5.3.13)

As Rank($(\mathbf{A} \ \beta)$)=4 and Rank(\mathbf{A})=3, Solution doesn't exist for $AX = \beta$ and hence β is not in the subspace S.

iii) Checking for γ : To check if a solution exists for $\mathbf{AX} = \gamma$. The corresponding agumented matrix can be written as,

$$(\mathbf{A} \quad \gamma) = \begin{pmatrix} 1 & 3 & -1 & -1 \\ 1 & 0 & 2 & 1 \\ -2 & 4 & 5 & 0 \\ 1 & -1 & 2 & 1 \end{pmatrix}$$
 (2.5.3.14)

On performing row-reduction on (2.5.3.14),

$$(\mathbf{A} \quad \gamma) = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$
 (2.5.3.15)

As Rank($(\mathbf{A} \quad \gamma)$)=4 and Rank((\mathbf{A}))=3, Solution doesn't exist for $AX = \gamma$ and hence γ is not in the subspace S.

- b) In part 1, we haven't considered the field to be either \mathbb{R} or \mathbb{C} . The above equations solved holds for field \mathbb{C} and that implies, they hold for field \mathbb{R} also. Hence α is in the subspace and β and γ are not in the subspace.
- c) **Theorem suggested:** Let \mathbb{F}_1 and \mathbb{F}_2 are two fields where \mathbb{F}_2 is subfield of \mathbb{F}_1 . Let α_i , i=1,2,3...,n forms basis for subspace of \mathbb{F}_2^n and a vector $\alpha \in \mathbb{F}_2^n$. Then α is in the subspace of \mathbb{F}_2^n spanned by α_i , i=1,2,3...,n if only if α is in the subspace of \mathbb{F}_1^n spanned by α_i , i=1,2,3...,n.
- 2.5.4. In C^3 , let $\alpha_1 = (1, 0, -i)$, $\alpha_2 = (1 + i, 1 i, 1)$, $\alpha_3 = (i, i, i)$. Prove that these vectors form a basis for C^3 . What are the coordinates of the vector (a,b,c) in the basis?

Solution: Now,

$$C_{1}\alpha_{1} + C_{2}\alpha_{2} + C_{3}\alpha_{3} = \mathbf{0}$$

$$(2.5.4.1)$$

$$\Longrightarrow C_{1}\begin{pmatrix} 1\\0\\-i \end{pmatrix} + C_{2}\begin{pmatrix} 1+i\\1-i\\1 \end{pmatrix} + C_{3}\begin{pmatrix} i\\i\\i \end{pmatrix} = \mathbf{0}$$

So,

$$\begin{pmatrix} 1 & 1+i & i \\ 0 & 1-i & i \\ -i & 1 & i \end{pmatrix} \begin{pmatrix} C_1 \\ C_2 \\ C_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$
 (2.5.4.3)

(2.5.4.2)

Considering the co-efficient matrix *A*:

$$\begin{pmatrix}
1 & 1+i & i \\
0 & 1-i & i \\
-i & 1 & i
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 + iR_1}
\begin{pmatrix}
1 & 1+i & i \\
0 & 1-i & i \\
0 & i & i-1
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3/i}
\begin{pmatrix}
1 & 1+i & i \\
0 & 1-i & i \\
0 & 1 & 1+i
\end{pmatrix}
\xrightarrow{R_3 \leftarrow (1-i)R_3}$$

$$\begin{pmatrix}
1 & 1+i & i \\
0 & 1-i & i \\
0 & 1-i & i
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 1+i & i \\
0 & 1-i & i \\
0 & 0 & 2-i
\end{pmatrix}$$

$$\xrightarrow{R_2 \leftarrow \frac{1+i}{1-i}R_2}
\begin{pmatrix}
1 & 1+i & i \\
0 & 1+i & -1 \\
0 & 0 & 2-i
\end{pmatrix}
\xrightarrow{R_1 \leftarrow R_1 - R_2}$$

$$\begin{pmatrix}
1 & 0 & i+1 \\
0 & 1+i & -1 \\
0 & 0 & 2-i
\end{pmatrix}$$

$$\begin{pmatrix}
1 & 0 & i+1 \\
0 & 1+i & -1 \\
0 & 0 & 2-i
\end{pmatrix}$$
(2.5.4.4)

Now let

$$R = \begin{pmatrix} 1 & 0 & i+1 \\ 0 & 1+i & -1 \\ 0 & 0 & 2-i \end{pmatrix}$$
 (2.5.4.5)

Where R is the row reduced form of matrix A. So α_1,α_2 and α_3 are linearly independent which implies that these 3 vectors form a basis of vector space C^3 .

Now, consider a vector $\beta = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$ and let the

coordinates are
$$x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$
 such that

$$Ax = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{2.5.4.6}$$

$$\implies x = A^{-1} \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{2.5.4.7}$$

Let us consider a matrix (A|I) where I is a 3x3 identity matrix. Now, applying the Gauss-

Jordon theorem we can get A^{-1}

$$\begin{pmatrix} 1 & 1+i & i & 1 & 0 & 0 \\ 0 & 1-i & i & 0 & 1 & 0 \\ -i & 1 & i & 0 & 0 & 1 \end{pmatrix}$$

$$\begin{matrix} \begin{matrix} & & & & & & & & & & & & & & & \\ & & & & & & & & & & \\ & & & & & & & & & \\ & & & & & & & & \\ & & & & & & & & \\ & & & & & & & & \\ & & & & & & & & \\ &$$

$$\begin{pmatrix} 1 & 0 & 0 & \frac{1-4i}{5} & \frac{1-2i}{5} & \frac{-2+4i}{5} \\ 0 & 1 & 0 & \frac{1-2i}{5} & \frac{1+3i}{5} & -\frac{2+i}{5} \\ 0 & 0 & 1 & \frac{3-i}{5} & -\frac{2+i}{5} & -\frac{3i+1}{5} \end{pmatrix}$$

$$(2.5.4.9)$$

$$\implies (I|A^{-1}) = \begin{pmatrix} 1 & 0 & 0 & \frac{1-4i}{5} & \frac{1-2i}{5} & \frac{-2+4i}{5} \\ 0 & 1 & 0 & \frac{1-2i}{5} & \frac{1+3i}{5} & -\frac{2+i}{5} \\ 0 & 0 & 1 & \frac{3-i}{5} & -\frac{2+i}{5} & -\frac{3i+1}{5} \end{pmatrix}$$

$$(2.5.4.10)$$

$$\implies A^{-1} = \begin{pmatrix} \frac{1-4i}{5} & \frac{1-2i}{5} & \frac{-2+4i}{5} \\ \frac{3-i}{5} & \frac{1+3i}{5} & -\frac{2+i}{5} \\ \frac{3-i}{5} & -\frac{2+i}{5} & -\frac{3i+1}{5} \end{pmatrix}$$

$$(2.5.4.11)$$

So,

$$x = A^{-1} \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$

$$(2.5.4.12)$$

$$\implies x = \begin{pmatrix} \frac{1-4i}{5} & \frac{1-2i}{5} & \frac{-2+4i}{5} \\ \frac{1-2i}{5} & \frac{1+3i}{5} & -\frac{2+i}{5} \\ \frac{3-i}{5} & -\frac{2+i}{5} & -\frac{3i+1}{5} \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$

$$(2.5.4.13)$$

$$\implies \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} \frac{1-4i}{5} & \frac{1-2i}{5} & \frac{-2+4i}{5} \\ \frac{1-2i}{5} & \frac{1+3i}{5} & -\frac{2+i}{5} \\ \frac{3-i}{5} & -\frac{2+i}{5} & -\frac{3i+1}{5} \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$

$$(2.5.4.14)$$

2.5.5. Give an explicit description of the type $b_j = \sum_{i=1}^{r} b_{ki} R_{ij}$ for the vectors

$$\beta = (b_1, b_2, b_3, b_4, b_5)$$

in R^5 which are linear combinations of the vectors

$$\alpha_1 = (1, 0, 2, 1, -1),$$
 (2.5.5.1)

$$\alpha_2 = (-1, 2, -4, 2, 0),$$
 (2.5.5.2)

$$\alpha_3 = (2, -1, 5, 2, 1),$$
 (2.5.5.3)

$$\alpha_4 = (2, 1, 3, 5, 2)$$
 (2.5.5.4)

Solution: Above matrix represented as: $Ax = \beta$

$$\begin{pmatrix} 1 & -1 & 2 & 2 \\ 0 & 2 & -1 & 1 \\ 2 & -4 & 5 & 3 \\ 1 & 2 & 2 & 5 \\ -1 & 0 & 1 & 2 \end{pmatrix} x = \beta$$
 (2.5.5.5)

$$\begin{pmatrix} 1 & -1 & 2 & 2 \\ 0 & 2 & -1 & 1 \\ 2 & -4 & 5 & 3 \\ 1 & 2 & 2 & 5 \\ -1 & 0 & 1 & 2 \end{pmatrix} x = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \\ b_5 \end{pmatrix}$$
 (2.5.5.6)

 $AX^T = Y^T$, where A, X and Y matrices have shown below:

$$A = \begin{pmatrix} 1 & -1 & 2 & 2 \\ 0 & 2 & -1 & 1 \\ 2 & -4 & 5 & 3 \\ 1 & 2 & 2 & 5 \\ -1 & 0 & 1 & 2 \end{pmatrix}$$
 (2.5.5.7)

$$X = \begin{pmatrix} \frac{67}{100} & -\frac{167}{100} & -2 & \frac{133}{100} \\ \frac{1}{2} & -\frac{3}{2} & -\frac{5}{2} & \frac{3}{2} \\ -\frac{4}{25} & \frac{117}{100} & \frac{3}{2} & -\frac{83}{100} \\ -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{pmatrix}$$
(2.5.5.8)

$$Y = \begin{pmatrix} 1 & 0 & 2 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$
 (2.5.5.9)

Now since the Y^T matrix is full rank thus we can say that columns of matrix A are linearly independent and b is described by (2.5.5.6)

2.5.6. Let $\mathbb V$ be a vector space which is spanned by the rows of matrix

$$\mathbf{A} = \begin{pmatrix} 3 & 21 & 0 & 9 & 0 \\ 1 & 7 & -1 & -2 & -1 \\ 2 & 14 & 0 & 6 & 1 \\ 6 & 42 & -1 & 13 & 0 \end{pmatrix} \tag{2.5.6.1}$$

a. Find a basis for \mathbb{V}

b. Tell which vectors $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$ are elements of $\mathbb V$

c. If $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$ is in \mathbb{V} , what are its coordinates in

the basis chosen in part(a)?

Solution: Row reducing (2.5.6.1)

$$\begin{pmatrix}
3 & 21 & 0 & 9 & 0 \\
1 & 7 & -1 & -2 & -1 \\
2 & 14 & 0 & 6 & 1 \\
6 & 42 & -1 & 13 & 0
\end{pmatrix}$$

$$\stackrel{R_1 \leftarrow \frac{R_1}{3}}{\longleftrightarrow} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
1 & 7 & -1 & -2 & -1 \\
2 & 14 & 0 & 6 & 1 \\
6 & 42 & -1 & 13 & 0
\end{pmatrix}$$

$$\begin{array}{c}
R_{3} \leftarrow R_{3} - 2R_{1} \\
R_{2} \leftarrow R_{2} - R_{1}
\end{array}
\begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & -1 & -5 & -1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & -1 & -5 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{2}}
\begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & -1 & -5 & -1 \\
0 & 0 & 0 & 1 \\
0 & 0 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_{2} \leftarrow -R_{2}}
\begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}}
\xrightarrow{R_{2} \leftarrow R_{2} - R_{3}}
\begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 0 \\
0 & 0 & 1 & 5 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}}
\xrightarrow{R_{2} \leftarrow R_{2} - R_{3}}
\begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}}
\xrightarrow{R_{2} \leftarrow R_{2} - R_{3}}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}}
\xrightarrow{R_{2} \leftarrow R_{2} - R_{3}}$$

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\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}}
\xrightarrow{R_{4} \leftarrow R_{4} - R_{4}}
\xrightarrow{R_{4} \leftarrow R_{4} - R_{4$$

a. For the basis of \mathbb{V} , we can take the non zero rows of (2.5.6.2)

$$\rho_1 = \begin{pmatrix} 1 \\ 7 \\ 0 \\ 3 \\ 0 \end{pmatrix} \rho_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \\ 5 \\ 0 \end{pmatrix} \rho_3 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$
 (2.5.6.3)

b. Vectors which are elements of \mathbb{V} are of the form:

$$c_{1}\rho_{1} + c_{2}\rho_{2} + c_{3}\rho_{3}$$

$$= \begin{pmatrix} c_{1} \\ 7c_{1} \\ c_{2} \\ 3c_{1} + 5c_{2} \\ c_{3} \end{pmatrix}$$

$$(2.5.6.4)$$

where c_1, c_2, c_3 are scalars.

c. Expressing (2.5.6.4) in matrix form, if
$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 is

in V,it must be of the form

$$\begin{pmatrix} 1 & 0 & 0 \\ 7 & 0 & 0 \\ 0 & 1 & 0 \\ 3 & 5 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 (2.5.6.5)

The augmented matrix form

$$\begin{pmatrix}
1 & 0 & 0 & x_1 \\
7 & 0 & 0 & x_2 \\
0 & 1 & 0 & x_3 \\
3 & 5 & 0 & x_4 \\
0 & 0 & 1 & x_5
\end{pmatrix}$$
(2.5.6.6)

Converting the above matrix into row reduced echelon form

$$\begin{pmatrix}
1 & 0 & 0 & x_1 \\
7 & 0 & 0 & x_2 \\
0 & 1 & 0 & x_3 \\
3 & 5 & 0 & x_4 \\
0 & 0 & 1 & x_5
\end{pmatrix}
\xrightarrow{R_4 \leftarrow R_4 - 3R_1}
\begin{pmatrix}
1 & 0 & 0 & x_1 \\
0 & 0 & 0 & x_2 - 7x_1 \\
0 & 1 & 0 & x_3 \\
0 & 5 & 0 & x_4 - 3x_1 \\
0 & 0 & 1 & x_5
\end{pmatrix}$$

$$\stackrel{R_2 \leftarrow R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & x_1 \\
0 & 1 & 0 & x_3 \\
0 & 0 & 0 & x_2 - 7x_1 \\
0 & 5 & 0 & x_4 - 3x_1 \\
0 & 0 & 1 & x_5
\end{pmatrix}$$

$$\stackrel{R_4 \leftarrow R_4 - 5R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & x_1 \\
0 & 1 & 0 & x_3 \\
0 & 0 & 0 & x_2 - 7x_1 \\
0 & 0 & 0 & x_4 - 3x_1 - 5x_3 \\
0 & 0 & 1 & x_5
\end{pmatrix}$$

$$\stackrel{R_5 \leftarrow R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & x_1 \\
0 & 1 & 0 & x_3 \\
0 & 0 & 1 & x_5 \\
0 & 0 & 0 & x_4 - 3x_1 - 5x_3 \\
0 & 0 & 0 & x_2 - 7x_1 \\
0 & 0 & 0 & x_2 - 7x_1
\end{pmatrix}$$

$$(2.5.6.7)$$

From (2.5.6.7),the coordinates of $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$ in the

basis is

$$\begin{pmatrix} x_1 \\ x_3 \\ x_5 \end{pmatrix}$$
 (2.5.6.8)

2.5.7. Let A be an m x n matrix over the field F, and consider the system of equations AX = Y. Prove that this system of equations has a solutions if and only if the row rank of A is equal to the row rank of augmented matrix of the system. Solution: Consider A as,

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \cdots & a_{mn} \end{pmatrix}_{m \times n}$$
(2.5.7.1)

Now to solve the equation AX = Y for X, we first write its augmented matrix A_Y and then convert it into row reduced echelon form given as R_Y . Here R is the row reduced form of matrix A. Also we know that, for AX = Y to have a solution, Y should be in column space of A.

$$\mathbf{A}_{\mathbf{Y}} = \begin{pmatrix} \mathbf{A} & | & \mathbf{Y} \end{pmatrix} \tag{2.5.7.2}$$

Assume that the last \mathbf{k} rows of \mathbf{R} are zero rows. This implies that we have \mathbf{k} number of linear dependent rows in matrix \mathbf{A} . Hence the row rank of matrix \mathbf{A} is $\mathbf{r} = \text{m-k}$. As there are m-k number of non-zero vectors in the row of \mathbf{R} . Now,

$$\mathbf{R}_{Y} = \mathbf{E}\mathbf{A}_{Y} \quad (2.5.7.3)$$

$$\implies \mathbf{C} \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} & y_1 \\ a_{21} & a_{22} & \cdots & a_{2n} & y_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} & y_m \end{pmatrix} (2.5.7.4)$$

Here, **E** is product of matrix responsible for performing operation on matrix such as scaling, Row exchange etc, given as,

$$\mathbf{E} = E_i E_{i-1} \cdots E_1 \tag{2.5.7.5}$$

Hence row reduced echelon form, $\mathbf{R}_{\mathbf{Y}}$ is given as,

$$\mathbf{R}_{\mathbf{Y}} = E\mathbf{A}_{\mathbf{Y}} = \begin{pmatrix} E\mathbf{A} & | & E\mathbf{Y} \end{pmatrix} \tag{2.5.7.6}$$

Now RREF can be represented in Block matrix

as,

$$\implies \mathbf{R}_{\mathbf{Y}} = \begin{pmatrix} I & F & | & Y \\ 0 & 0 & | & \end{pmatrix} \tag{2.5.7.7}$$

Also from equation (2.5.7.7) it can be observed that for AX = Y to have a solution,

$$y'_{m-k} = y'_{m-k-1} = \dots = y'_{m-1} = y'_m = 0$$
 (2.5.7.8)

Hence, the rank of $\mathbf{R}'_{\mathbf{v}}$ is, also \mathbf{r} . This implies rank of augmented matrix A_Y is also \mathbf{r} .

Example: Let A be a 3×4 matrix given as,

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 5 & 7 \\ 2 & 9 & 3 & 6 \\ 1 & 7 & 2 & 9 \end{pmatrix} \tag{2.5.7.9}$$

$$\mathbf{Y} = \begin{pmatrix} 4\\8\\14 \end{pmatrix} \tag{2.5.7.10}$$

For the time being it could be concluded that the following set of linear equation have a 3.1.1. Find weather given functions T from \mathbb{R}^2 into solution as $Y \in \text{Column-space}(A)$ Hence to solve AX = Y first the augmented matrix A_Y is given as,

$$\mathbf{A}_{\mathbf{Y}} = \begin{pmatrix} 1 & 2 & 5 & 7 & | & 4 \\ 2 & 9 & 3 & 6 & | & 8 \\ 1 & 7 & 2 & 9 & | & 14 \end{pmatrix} \tag{2.5.7.11}$$

Now reducing in RREF we get,

$$\mathbf{R}_{\mathbf{Y}} = E\mathbf{A}_{\mathbf{Y}} \tag{2.5.7.12}$$

Here E is elementary matrix given as,

$$E = \begin{pmatrix} -3/20 & 31/20 & -39/20 \\ -1/20 & -3/20 & 7/20 \\ 1/4 & -1/4 & 1/4 \end{pmatrix}$$
 (2.5.7.13)

$$\implies \mathbf{R}_{\mathbf{Y}} = E\mathbf{A}_{\mathbf{Y}} = \begin{pmatrix} E\mathbf{A} & E\mathbf{Y} \end{pmatrix} \quad (2.5.7.14)$$
$$\implies \mathbf{R}_{\mathbf{Y}} = \begin{pmatrix} I & F & Y \end{pmatrix} \quad (2.5.7.15)$$

$$\mathbf{I}_{3x3} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{2.5.7.16}$$

$$\mathbf{F}_{3x1} = \begin{pmatrix} -93/10\\19/10\\5/2 \end{pmatrix} \tag{2.5.7.17}$$

$$\mathbf{Y}_{3x1} = \begin{pmatrix} -31/2 \\ 7/2 \\ 5/2 \end{pmatrix} \tag{2.5.7.18}$$

Hence, rank = 3 as there are 0 non zero rows in row reduced echelon form. Also it could be observed that both \mathbf{R} and $\mathbf{R}_{\mathbf{Y}}$ have non zero rows hence we have a solution and same rank.

3 Linear Transformations

3.1 Linear Transformations

 \mathbb{R}^2 are linear transformations or not a)

$$T\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 + x_1 \\ x_2 \end{pmatrix} \tag{3.1.1.1}$$

Solution: Counter example can be given as follows:-

$$x_1 = x_2 = 0 (3.1.1.2)$$

Substituting (3.1.1.2) in (3.1.1.1) we get,

$$T\begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{3.1.1.3}$$

(3.1.1.3) is clearly false because linear transformation on $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$ will always be equal to $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$

$$\mathbf{T} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_2 \\ x_1 \end{pmatrix} \tag{3.1.1.4}$$

Does function **T** from \mathbb{R}^2 into \mathbb{R}^2 is Linear Transformation.

Solution: Let,

b)

$$\mathbf{x}, \mathbf{y} \in \mathbb{R}^2 \tag{3.1.1.5}$$

Using transformation on T,

$$\mathbf{T}(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.1.1.6}$$

From (3.1.1.4) we get,

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \tag{3.1.1.7}$$

With c being a scalar,

$$\mathbf{T}(c\mathbf{x} + \mathbf{y}) = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} (c\mathbf{x} + \mathbf{y})$$
 (3.1.1.8)

$$= \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} c \mathbf{x} + \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \mathbf{y} \qquad (3.1.1.9)$$

$$= \begin{pmatrix} 0 & c \\ c & 0 \end{pmatrix} \mathbf{x} + \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \mathbf{y} \qquad (3.1.1.10)$$

$$= c \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \mathbf{x} + \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \mathbf{y} \qquad (3.1.1.11)$$

From (3.1.1.11) we can say,

c)

$$\mathbf{T}(c\mathbf{x} + \mathbf{y}) = c\mathbf{T}(\mathbf{x}) + \mathbf{T}(\mathbf{y}) \qquad (3.1.1.12)$$

Hence from (3.1.1.12) we can say T is a Linear Transformation from \mathbb{R}^2 to \mathbb{R}^2

$$\mathbf{T}(x_1, x_2) = (x_1^2, x_2) \tag{3.1.1.13}$$

Solution: If **T** were a linear transformation then we would have

$$\mathbf{T} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \qquad (3.1.1.14)$$

$$\implies \mathbf{T} \begin{pmatrix} -1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} \end{pmatrix} = -1.\mathbf{T} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \qquad (3.1.1.15)$$

$$\implies -1. \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \qquad (3.1.1.16)$$

which is a contradiction, since

$$\mathbf{T} \begin{pmatrix} -1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \qquad (3.1.1.17)$$
$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} \neq \begin{pmatrix} -1 \\ 0 \end{pmatrix}. \qquad (3.1.1.18)$$

Hence non-linear transformation.

d) Is the following function **T** from \mathbb{R}^2 into \mathbb{R}^2 is linear transformation?

$$\mathbf{T} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \sin(x_1) \\ x_2 \end{pmatrix}$$

Solution: Let,

$$\mathbf{x} = \begin{pmatrix} \pi \\ 0 \end{pmatrix}; \quad \mathbf{y} = \begin{pmatrix} \frac{\pi}{2} \\ 0 \end{pmatrix}$$

$$\mathbf{T}(\mathbf{x} + \mathbf{y}) = \mathbf{T} \begin{pmatrix} \frac{3\pi}{2} \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \end{pmatrix}$$
 (3.1.1.19)

$$\mathbf{T}(\mathbf{x}) + \mathbf{T}(\mathbf{y}) = \mathbf{T} \begin{pmatrix} \pi \\ 0 \end{pmatrix} + \mathbf{T} \begin{pmatrix} \frac{\pi}{2} \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$
(3.1.1.20)

From (3.1.1.19) and (3.1.1.20) additive transformation property is not satisfied. Hence not a linear transformation.

e) Verify whether $T(x_1, x_2) = (x_1 - x_2, 0)$ is a linear transformation or not. **Solution:** Let V and W be the vector spaces. The function $T: V \to W$ is called a linear transformation of V into W if for all U and V in V and for any scalar V in field V,

$$T(k\mathbf{u} + \mathbf{v}) = kT(\mathbf{u}) + T(\mathbf{v}) \qquad (3.1.1.21)$$

Given,

$$T\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \tag{3.1.1.22}$$

Consider,

$$T(k\mathbf{x} + \mathbf{y}) = \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} (k\mathbf{x} + \mathbf{y}) \quad (3.1.1.23)$$

$$= \begin{pmatrix} k & -k \\ 0 & 0 \end{pmatrix} \mathbf{x} + \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \mathbf{y}$$

$$= k \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \mathbf{x} + \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \mathbf{y}$$

$$(3.1.1.24)$$

$$T(\mathbf{x} + \mathbf{y}) = kT(\mathbf{x}) + T(\mathbf{y}) \quad (3.1.1.26)$$

Therefore, the given function T is a linear transformation.

3.1.2. Find the range, rank, null space, and nullity for the zero transformation and the identity transformation on a finite-dimensional space V. **Solution:**

Suppose vector space V has dim(V) = n. Table 3.1.2 provides the properties of range, rank, null space and nullity of zero and identity transformation on a vector space V

3.1.3. a) Let **F** be a field and let **V** be the space of polynomial functions f from **F** into **F**, given by

$$f(x) = c_0 + c_1 x + \dots + c_n x^n$$

Let **D** be a linear differentiation transforma-

Properties	Zero Transformation	Identity Transformation
Transformation	$T_0(\mathbf{v}) = 0$	$T_I(\mathbf{v}) = \mathbf{v}$
Range	Zero subspace {0}	V
Rank	$\dim(0) = 0$	$dim(\mathbf{V}) = n$
Null space	V	Zero subspace {0}
Nullity	$\dim(\mathbf{V}) = \mathbf{n}$	$\dim(0) = 0$

TABLE 3.1.2: Properties of Zero and Identity transformation

tion defined as

$$(\mathbf{D}f)(x) = \frac{df(x)}{dx}$$

Then find the range and null space of **D**.

b) Let R be the field of real numbers and let
 V be the space of all functions from R into R which are continuous. Let T be linear transformation defined by

$$(\mathbf{T}f)(x) = \int_0^x f(t) \, dt$$

Find the range and null space of **T**. **Solution:** Let the vector space of n-dimension be deined as

$$\mathbf{V} = \left\{ f : \mathbf{F} \to \mathbf{F} : f(x) = \sum_{k=0}^{n} c_k x^k, \ c_k \in \mathbf{F} \right\}$$
(3.1.3.1)

The corresponding standard basis for V is

$$\left\{ \begin{pmatrix} 1\\0\\\vdots\\0 \end{pmatrix}, \begin{pmatrix} 0\\x\\\vdots\\0 \end{pmatrix}, \cdots, \begin{pmatrix} 0\\0\\\vdots\\x^{n-1} \end{pmatrix} \right\}$$
(3.1.3.2)

a) Let f and $g \in \mathbf{V}$ and let α and $\beta \in \mathbf{F}$ then

$$\mathbf{D}(\alpha f + \beta g) = \frac{d(\alpha f(x) + \beta g(x))}{dx}$$
 (3.1.3.3)
$$= \alpha \frac{df(x)}{dx} + \beta \frac{dg(x)}{dx}$$
 (3.1.3.4)
$$= \alpha (\mathbf{D}f) + \beta (\mathbf{D}g)$$
 (3.1.3.5)

Therefore **D** is a linear transformation. The **D** transformation maps the k^{th} basis vector as follows

$$\mathbf{D} \begin{pmatrix} 0 \\ \vdots \\ 0 \\ x^k \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ \vdots \\ kx^{k-1} \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$
 (3.1.3.6)

Since the transformed vector

$$\begin{pmatrix} 0 \\ \vdots \\ kx^{k-1} \\ 0 \\ \vdots \\ 0 \end{pmatrix} \in \mathbf{V} \tag{3.1.3.7}$$

the range of **D** is the vector space **V**. Thus the transformation is defined as $\mathbf{D}: \mathbf{V} \to \mathbf{V}$. Therefore the **D** Transformation on the basis vector set is

$$\mathbf{D} \begin{pmatrix} 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 1 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 1 & 0 \\ 0 & 0 & 0 & \cdots & 0 & 1 \end{pmatrix}$$
(3.1.3.8)

$$= \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 2 & \cdots & 0 & 0 \\ 0 & 0 & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 & n-2 \\ 0 & 0 & 0 & \cdots & 0 & 0 \end{pmatrix}$$
(3.1.3.9)

Thus the **D** transformation coefficient matrix

is

$$D = \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 2 & \cdots & 0 & 0 \\ 0 & 0 & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 & n-2 \\ 0 & 0 & 0 & \cdots & 0 & 0 \end{pmatrix}$$
(3.1.3.10)

Since D contains a zero row hence |D| = 0. Therefore **D** transformation matrix is singular. The nullspace for differentiation transformation is defined as

$$\mathbf{N} = \{ f \in \mathbf{V} : \mathbf{D}f = 0 \} \tag{3.1.3.11}$$

Let the coefficient matrix of $f \in \mathbf{V}$ be

$$\mathbf{f} = \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_{n-1} \end{pmatrix} \tag{3.1.3.12}$$

then

$$\mathbf{D}f = 0 \qquad (3.1.3.13)$$

$$\Rightarrow \begin{pmatrix}
0 & 1 & 0 & \cdots & 0 & 0 \\
0 & 0 & 2 & \cdots & 0 & 0 \\
0 & 0 & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
0 & 0 & 0 & \cdots & 0 & n-2 \\
0 & 0 & 0 & \cdots & 0 & 0
\end{pmatrix}
\begin{pmatrix}
c_0 \\
c_1 \\
\vdots \\
c_{n-1}
\end{pmatrix} = \mathbf{0}$$

$$(3.1.3.14)$$

Since *D* is in row reduced echolon form and rank(D) = n - 1 the solution of (3.1.3.14) is

$$\mathbf{f} = \begin{pmatrix} k \\ 0 \\ \vdots \\ 0 \end{pmatrix} \tag{3.1.3.15}$$

where $k \in \mathbf{R}$. Therefore the nullspace for $\mathbf{D} : \mathbf{V} \to \mathbf{V}$ is

$$\mathbf{N} = \left\{ \begin{pmatrix} k \\ 0 \\ \vdots \\ 0 \end{pmatrix} : k \in \mathbf{R} \right\} \tag{3.1.3.16}$$

b) Let f and $g \in \mathbf{V}$ and let α and $\beta \in \mathbf{F}$ then

$$\mathbf{T}(\alpha f + \beta g) = \int_0^x (\alpha f(t) + \beta g(t)) dt$$

$$= \alpha \int_0^x f(t) dt + \beta \int_0^x g(t) dt$$

$$= \alpha (\mathbf{T}f) + \beta (\mathbf{T}g) \qquad (3.1.3.19)$$

Therefore **T** is a linear transformation. The **T** transformation maps the k^{th} basis vector as follows

$$\mathbf{T} \begin{pmatrix} 0 \\ \vdots \\ x^{k} \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \frac{x^{k+1}}{k+1} \\ \vdots \\ 0 \end{pmatrix}$$
 (3.1.3.20)

Since the transformed vector

$$\begin{pmatrix} 0 \\ \vdots \\ 0 \\ \frac{x^{k+1}}{k+1} \\ \vdots \\ 0 \end{pmatrix} \in \mathbf{V}$$
 (3.1.3.21)

the range of T is the vector space V. Thus the transformation is defined as $T:V \to V$. Therefore the T Transformation on the basis vector set is

$$\mathbf{T} \begin{pmatrix} 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 1 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 1 & 0 \\ 0 & 0 & 0 & \cdots & 0 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 0 & 0 & 0 & \cdots & 0 & 0 \\ 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & \frac{1}{2} & 0 & \cdots & 0 & 0 \\ 0 & \frac{1}{2} & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 0 & \cdots & \frac{1}{n-1} & 0 \\ 0 & 0 & 0 & \cdots & \frac{1}{n-1} & 1 \\ 0 & 0 & 0 & \cdots & \frac{1}{n-1} & 1 \end{pmatrix}$$

$$(3.1.3.22)$$

Thus the **T** transformation coefficient matrix

is

$$T = \begin{pmatrix} 0 & 0 & 0 & \cdots & 0 & 0 \\ 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & \frac{1}{2} & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 0 & \cdots & \frac{1}{n-1} & 0 \\ 0 & 0 & 0 & \cdots & 0 & \frac{1}{n} \end{pmatrix}$$
(3.1.3.24)

Since T contains a zero row hence |T| = 0. Therefore T transformation matrix is singular. The nullspace for integration transformation is defined as

$$\mathbf{N} = \{ f \in \mathbf{V} : \mathbf{T}f = 0 \} \tag{3.1.3.25}$$

Let the coefficient matrix of $f \in \mathbf{V}$ be

$$\mathbf{f} = \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_{n-1} \end{pmatrix} \tag{3.1.3.26}$$

then

$$\begin{array}{c}
\mathbf{T}f = 0 & (3.1.3.27) \\
\begin{pmatrix}
0 & 0 & 0 & \cdots & 0 & 0 \\
1 & 0 & 0 & \cdots & 0 & 0 \\
0 & \frac{1}{2} & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
0 & 0 & 0 & \cdots & 0 & 0 \\
0 & 0 & 0 & \cdots & \frac{1}{n-1} & 0 \\
0 & 0 & 0 & \cdots & 0 & \frac{1}{n}
\end{pmatrix} \begin{pmatrix}
c_0 \\
c_1 \\
\vdots \\
c_{n-1}
\end{pmatrix} = \mathbf{0}$$

$$(3.1.3.28)$$

Since T is in row reduced echolon form and rank(T) = n the solution of (3.1.3.28) is

$$\mathbf{f} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \tag{3.1.3.29}$$

where $k \in \mathbf{R}$. Therefore the nullspace for $\mathbf{T} : \mathbf{V} \to \mathbf{V}$ is

$$\mathbf{N} = \left\{ \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} : k \in \mathbf{R} \right\} \tag{3.1.3.30}$$

3.1.4. Is there a linear transformation **T** from \mathbb{R}^3 into

 \mathbb{R}^2 such that,

$$\mathbf{T} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{3.1.4.1}$$

$$\mathbf{T} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \tag{3.1.4.2}$$

Solution: Linear Transformation A linear transformation is a function $\mathbf{T}: \mathbb{R}^n \to \mathbb{R}^m$ which satisfies:

1. $\forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n$,

$$\mathbf{T}(\mathbf{x} + \mathbf{y}) = \mathbf{T}(\mathbf{x}) + \mathbf{T}(\mathbf{y}) \tag{3.1.4.3}$$

2. $\forall \mathbf{x} \in \mathbb{R}^n$ and $\mathbf{c} \in \mathbb{R}$,

$$\mathbf{T}(c\mathbf{x}) = c\mathbf{T}(\mathbf{x}) \tag{3.1.4.4}$$

Matrix of the Linear Transformation Let, **T**: $\mathbb{R}^n \to \mathbb{R}^m$ is a linear transformation and $\mathbf{x} \in \mathbb{R}^n$ is given by ,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \tag{3.1.4.5}$$

$$= x_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} + \dots + x_n \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$$
 (3.1.4.6)

Let $e_1, e_2, ..., e_n$ be the standard basis of \mathbb{R}^n and the equation (3.1.4.6) can be rewritten as,

$$\mathbf{x} = \sum_{i=1}^{n} x_i \mathbf{e_i} \tag{3.1.4.7}$$

$$\mathbf{T}\left(\mathbf{x}\right) = \sum_{i=1}^{n} x_i \mathbf{T}\left(\mathbf{e_i}\right)$$
 (3.1.4.8)

$$= (\mathbf{T}(\mathbf{e}_1) \quad \mathbf{T}(\mathbf{e}_2) \quad \dots \quad \mathbf{T}(\mathbf{e}_n)) \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$
(3.1.4.9)

$$\mathbf{T}(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.1.4.10}$$

Where,

$$\mathbf{A} = (\mathbf{T}(\mathbf{e}_1) \ \mathbf{T}(\mathbf{e}_n) \ \dots \ \mathbf{T}(\mathbf{e}_n)) \ (3.1.4.11)$$

If **T** is any linear transformation which maps $\mathbb{R}^n \to \mathbb{R}^m$ there is always an $m \times n$ matrix **A** with the property that

$$\mathbf{T}(\mathbf{x}) = \mathbf{A}\mathbf{x}$$

where , $\mathbf{x} \in \mathbb{R}^n$ *Solution* Let,

$$\mathbf{v} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \tag{3.1.4.12}$$

$$\mathbf{u} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \tag{3.1.4.13}$$

Given,

$$\mathbf{T}\left(\mathbf{v}\right) = \begin{pmatrix} 1\\0 \end{pmatrix} \tag{3.1.4.14}$$

$$\mathbf{T}\left(\mathbf{u}\right) = \begin{pmatrix} 0\\1 \end{pmatrix} \tag{3.1.4.15}$$

Let the standard basis vectors is denoted as,

$$\mathbf{e_1} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \tag{3.1.4.16}$$

$$\mathbf{e_2} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \tag{3.1.4.17}$$

$$\mathbf{e_3} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \tag{3.1.4.18}$$

Let, $\mathbf{T}: \mathbb{R}^3 \to \mathbb{R}^2$ be a linear transformation. Then the function \mathbf{T} is just matrix-vector multiplication $\mathbf{T}(\mathbf{x}) = \mathbf{A}\mathbf{x}$ for some matrix \mathbf{A} as shown in equation (3.1.4.10)

Matrix A of order 2×3 is given by,

$$\mathbf{A} = (\mathbf{T}(\mathbf{e}_1) \ \mathbf{T}(\mathbf{e}_2) \ \mathbf{T}(\mathbf{e}_3)) \qquad (3.1.4.19)$$

Consider the vector $\mathbf{b} \in \mathbb{R}^3$ which is the linear combinations of the vectors \mathbf{v} and \mathbf{u} . For $x_1, x_2 \in \mathbb{R}$,

 $\mathbf{b} = \begin{pmatrix} b_1 \\ b_2 \\ b_2 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} + x_2 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ (3.1.4.20)

$$\mathbf{T} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} = x_1 \mathbf{T} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} + x_2 \mathbf{T} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \qquad (3.1.4.21)$$

To find x_1, x_2 , we solve the linear system, $\mathbf{M}\mathbf{x} = \mathbf{b}$ where \mathbf{M} is the 3 × 2 matrix obtained by stacking the given vectors \mathbf{v} and \mathbf{u} as columns

$$\mathbf{M} = \begin{pmatrix} 1 & 1 \\ -1 & 1 \\ 1 & 1 \end{pmatrix} \tag{3.1.4.22}$$

$$\begin{pmatrix} 1 & 1 \\ -1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$
 (3.1.4.23)

The augumented matrix of the equation (3.1.4.23) is given by ,

$$\begin{pmatrix} 1 & 1 & b_1 \\ -1 & 1 & b_2 \\ 1 & 1 & b_3 \end{pmatrix} \tag{3.1.4.24}$$

By row reducing the above equation (3.1.4.24),

$$\begin{pmatrix} 1 & 1 & b_1 \\ -1 & 1 & b_2 \\ 1 & 1 & b_3 \end{pmatrix} \xrightarrow{R_2 = R_2 + R_1} \begin{pmatrix} 1 & 1 & b_1 \\ 0 & 2 & b_2 + b_1 \\ 1 & 1 & b_3 \end{pmatrix}$$

$$(3.1.4.25)$$

$$\begin{pmatrix} 1 & 1 & b_1 \\ 0 & 2 & b_2 + b_1 \\ 1 & 1 & b_3 \end{pmatrix} \xrightarrow{R_3 = R_3 - R_1} \begin{pmatrix} 1 & 1 & b_1 \\ 0 & 2 & b_2 + b_1 \\ 0 & 0 & b_3 - b_1 \end{pmatrix}$$

$$(3.1.4.26)$$

$$\begin{pmatrix} 1 & 1 & b_1 \\ 0 & 2 & b_2 + b_1 \\ 0 & 0 & b_3 - b_1 \end{pmatrix} \xrightarrow{R_2 = \frac{R_2}{2}} \begin{pmatrix} 1 & 1 & b_1 \\ 0 & 1 & \frac{b_2 + b_1}{2} \\ 0 & 0 & b_3 - b_1 \end{pmatrix}$$

$$(3.1.4.27)$$

$$\begin{pmatrix} 1 & 1 & b_1 \\ 0 & 1 & \frac{b_2 + b_1}{2} \\ 0 & 0 & b_3 - b_1 \end{pmatrix} \xrightarrow{R_1 = R_1 - R_2} \begin{pmatrix} 1 & 0 & \frac{b_1 - b_2}{2} \\ 0 & 1 & \frac{b_2 + b_1}{2} \\ 0 & 0 & b_3 - b_1 \end{pmatrix}$$

$$(3.1.4.28)$$

Now equation (3.1.4.23) can be written as,

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \frac{b_1 - b_2}{2} \\ \frac{b_2 + b_1}{2} \\ b_3 - b_1 \end{pmatrix}$$
(3.1.4.29)

Solving the above equation we get,

$$x_1 = \frac{b_1 - b_2}{2} \tag{3.1.4.30}$$

$$x_2 = \frac{b_1 + b_2}{2} \tag{3.1.4.31}$$

Substituting the above equations (3.1.4.30), (3.1.4.31) in equation (3.1.4.21), we

get,

$$\mathbf{T} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} = \left(\frac{b_1 - b_2}{2}\right) \mathbf{T} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} + \left(\frac{b_1 + b_2}{2}\right) \mathbf{T} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$
(3.1.4.32)

Substituting the equations (3.1.4.1) and (3.1.4.2) in equation (3.1.4.32) we get,

$$\mathbf{T} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} = \left(\frac{b_1 - b_2}{2} \right) \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \left(\frac{b_1 + b_2}{2} \right) \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$
(3.1.4.33)

$$\mathbf{T} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} = \begin{pmatrix} \frac{b_1 - b_2}{2} \\ \frac{b_1 + b_2}{2} \end{pmatrix}$$
 (3.1.4.34)

Using the above equation (3.1.4.34) we compute,

$$\mathbf{T}\left(\mathbf{e_1}\right) = \mathbf{T} \begin{pmatrix} 1\\0\\0 \end{pmatrix} = \begin{pmatrix} \frac{1}{2}\\\frac{1}{2} \end{pmatrix} \tag{3.1.4.35}$$

$$\mathbf{T}\left(\mathbf{e_2}\right) = \mathbf{T} \begin{pmatrix} 0\\1\\0 \end{pmatrix} = \begin{pmatrix} \frac{-1}{2}\\\frac{1}{2} \end{pmatrix} \tag{3.1.4.36}$$

$$\mathbf{T}\left(\mathbf{e}_{3}\right) = \mathbf{T} \begin{pmatrix} 0\\0\\1 \end{pmatrix} = \begin{pmatrix} 0\\0 \end{pmatrix} \tag{3.1.4.37}$$

Substituting the equations (3.1.4.35),(3.1.4.36) and (3.1.4.37) in equation (3.1.4.19) we get,

$$\mathbf{A} = \begin{pmatrix} \frac{1}{2} & \frac{-1}{2} & 0\\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix} \tag{3.1.4.38}$$

$$\implies \mathbf{A} = \frac{1}{2} \begin{pmatrix} 1 & -1 & 0 \\ 1 & 1 & 0 \end{pmatrix}$$
 (3.1.4.39)

Therefore from the above matrix **A** we can say that there is a linear transformation **T** from \mathbb{R}^3 into \mathbb{R}^2 which satisfies the given conditions (3.1.4.1) and (3.1.4.2).

3.1.5. Describe explicitly the linear transformation T from \mathbb{F}^2 into \mathbb{F}^2 such that $T(\epsilon_1) = (a, b), T(\epsilon_2) = (c, d)$. **Solution:** We are given a linear transformation,

$$T: \mathbb{F}^2 \to \mathbb{F}^2 \tag{3.1.5.1}$$

The transformation for \in_1 and \in_2 can be written

as,

$$T(\epsilon_1) = \begin{pmatrix} a \\ b \end{pmatrix} \tag{3.1.5.2}$$

$$T(\epsilon_2) = \begin{pmatrix} c \\ d \end{pmatrix} \tag{3.1.5.3}$$

Now,let's assume \in_1 and \in_2 as linearly independent. So the linear transformation T for any vector \mathbf{v} in two dimensional space will be,

$$T(\mathbf{v}) = (T(\epsilon_1) \quad T(\epsilon_2))\mathbf{v} \tag{3.1.5.4}$$

$$= \begin{pmatrix} a & c \\ b & d \end{pmatrix} \mathbf{v} \tag{3.1.5.5}$$

Now, there can be two cases here, transformation of linearly independent vector can be independent or it can be dependent. Considering the first case and (3.1.5.5) we can say that,

$$Range(T) = \text{columnspace of} \begin{pmatrix} a & c \\ b & d \end{pmatrix}$$
 (3.1.5.6)

Now, considering the case when linear transformation will be linearly dependent,

$$Range(T) = \text{columnspace of} \begin{pmatrix} a \\ b \end{pmatrix}$$
 (3.1.5.7)

Now, considering that vectors ϵ_1 and ϵ_2 itself are linearly dependent.Let $\mathbf{v} = \epsilon_1 + \epsilon_2$

$$T(\mathbf{v}) = T(\epsilon_1) + T(\epsilon_2)$$
 (3.1.5.8)

$$= T(\epsilon_1) + T(k \epsilon_1)$$
 (3.1.5.9)

$$= (k+1)T(\in_1) \tag{3.1.5.10}$$

$$= (k+1)\binom{a}{b}$$
 (3.1.5.11)

We can see from above equation that when \in_1 and \in_2 as linearly dependent then the transformation T will be along the line only.

Vectors Independent	Vectors Dependent
$T(\mathbf{v}) = \begin{pmatrix} a & c \\ b & d \end{pmatrix} \mathbf{v}$	$T(\mathbf{v}) = (\mathbf{k} + 1) \begin{pmatrix} a \\ b \end{pmatrix}$
Output:	Output:
On the plane	On the line

TABLE 3.1.5

3.1.6. Let \mathbb{F} be a subfield of the complex numbers and let \mathbb{T} be the function from \mathbb{F}^3 into \mathbb{F}^3 defined

by

$$\mathbb{T}(x_1, x_2, x_3) =$$

$$(3.1.6.1)$$

$$(x_1 - x_2 + 2x_3, 2x_1 + x_2, -x_1 - 2x_2 + 2x_3)$$

$$(3.1.6.2)$$

- a) Verify that \mathbb{T} is a linear transformation.
- b) If (a, b, c) is a vector in \mathbb{F}^3 , what are the conditions on a, b, c that the vector be in the range of \mathbb{T} ? What is the rank of \mathbb{T} ?
- c) What are the conditions on a, b, c that (a, b, c) be in the null space of \mathbb{T} ? What is the nullity of \mathbb{T} ?

Solution: Representing the transformation in matrix form

$$\mathbb{T}(x_1, x_2, x_3) = \mathbf{Tx} \tag{3.1.6.3}$$

$$\mathbf{T} = \begin{pmatrix} 1 & -1 & 2 \\ 2 & 1 & 0 \\ -1 & -2 & 2 \end{pmatrix} \tag{3.1.6.4}$$

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \tag{3.1.6.5}$$

Part (a) Consider the matrices $\mathbf{x}, \mathbf{y} \in \mathbb{F}^3$ and the scalar $c \in \mathbb{F}$

By the associativity of matrix multiplications, we can write

$$\mathbf{T}(c\mathbf{x} + \mathbf{y}) = \mathbf{T}(c\mathbf{x}) + \mathbf{T}\mathbf{y}$$
 (3.1.6.6)

$$= c\mathbf{T}\mathbf{x} + \mathbf{T}\mathbf{y} \tag{3.1.6.7}$$

So, T is a linear transformation. Part (b)

$$range(\mathbf{T}) = \{ \mathbf{y} : \mathbf{T}\mathbf{x} = \mathbf{y} \text{ where } \mathbf{x}, \mathbf{y} \in \mathbb{F}^3 \}$$
(3.1.6.8)

$$\mathbf{y} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad (3.1.6.9)$$

$$Tx = y (3.1.6.10)$$

$$\implies$$
 BT $\mathbf{x} = \mathbf{B}\mathbf{y}$ (3.1.6.11)

$$\implies \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0\\ \frac{-2}{3} & \frac{1}{3} & 0\\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 & 2\\ 2 & 1 & 0\\ -1 & -2 & 2 \end{pmatrix} \mathbf{x} = (3.1.6.12)$$

$$\begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0\\ \frac{-2}{3} & \frac{1}{3} & 0\\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} a\\ b\\ c \end{pmatrix} (3.1.6.13)$$

$$\begin{pmatrix} 1 & 0 & \frac{2}{3} \\ 0 & 1 & \frac{-4}{3} \\ 0 & 0 & 0 \end{pmatrix} \mathbf{x} = \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0 \\ \frac{-2}{3} & \frac{1}{3} & 0 \\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} (3.1.6.14)$$

So, rank(T)=2 and comparing the third row element in LHS and RHS of (3.1.6.14)

$$-a + b + c = 0 (3.1.6.15)$$

All vectors $\begin{pmatrix} a \\ b \\ c \end{pmatrix} \in \mathbb{F}^3$ that satisfy (3.1.6.15) lie in the range of **T** Part (c)

 $nullspace(\mathbf{T}) = \{\mathbf{x} : \mathbf{T}\mathbf{x} = \mathbf{0} \text{ where } \mathbf{x} \in \mathbb{F}^3\}$ (3.1.6.16)

$$\mathbf{x} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{3.1.6.17}$$

$$Tx = 0$$
 (3.1.6.18)

$$BTx = 0$$
 (3.1.6.19)

where **BT** is in reduced row echelon form

$$\begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0\\ \frac{-2}{3} & \frac{1}{3} & 0\\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 & 2\\ 2 & 1 & 0\\ -1 & -2 & 2 \end{pmatrix} \mathbf{x} = \mathbf{0}$$
 (3.1.6.20)

$$\implies \begin{pmatrix} 1 & 0 & \frac{2}{3} \\ 0 & 1 & \frac{-4}{3} \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad (3.1.6.21)$$

$$\implies a + \frac{2}{3}c = 0$$
 (3.1.6.22)

$$b - \frac{4}{3}c = 0 \qquad (3.1.6.23)$$

The number of free variables in the reduced row echelon form of **T** is 1 hence nullity(**T**) =1

So, the null space of T is set of all vectors

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix} \in \mathbb{F}^3 \text{ that satisfy } (3.1.6.22) \text{ and } (3.1.6.23)$$

Note

 $rank(\mathbf{T})+nullity(\mathbf{T})=2+1=dim(\mathbb{F}^3)$

3.1.7. Describe explicitly a linear transformation from R^3 into R^3 which has as its range the subspace spanned by $\begin{pmatrix} 1 & 0 & -1 \end{pmatrix}$ and $\begin{pmatrix} 1 & 2 & 2 \end{pmatrix}$. **Solution:** Transformation T from R^3 to R^3 range gives the column space. Hence,

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.1.7.1}$$

$$T(\mathbf{x}) = \begin{pmatrix} 1 & 1 \\ 0 & 2 \\ -1 & 2 \end{pmatrix} \mathbf{x}$$
 (3.1.7.2)

3.1.8. Let **V** be the vector space of all $n \times n$ matrices over the field \mathbb{F} , and let **B** be a fixed $n \times n$ matrix. If a transformation T defined as follows,

$$T(\mathbf{A}) = \mathbf{A}\mathbf{B} - \mathbf{B}\mathbf{A}$$

Prove that T is a linear transformation from V into V Solution: Let,

$$\mathbf{A_1} \in \mathbf{V} \tag{3.1.8.1}$$

$$\mathbf{A_2} \in \mathbf{V} \tag{3.1.8.2}$$

If c be any scalar of the field \mathbb{F} we get,

$$c\mathbf{A_1} + \mathbf{A_2} \in \mathbf{V} \tag{3.1.8.3}$$

Applying transformation T on $(c\mathbf{A_1} + \mathbf{A_2})$ we get,

$$T(c\mathbf{A}_{1} + \mathbf{A}_{2}) = (c\mathbf{A}_{1} + \mathbf{A}_{2})\mathbf{B} - \mathbf{B}(c\mathbf{A}_{1} + \mathbf{A}_{2})$$

$$(3.1.8.4)$$

$$= c\mathbf{A}_{1}\mathbf{B} + \mathbf{A}_{2}\mathbf{B} - c\mathbf{B}\mathbf{A}_{1} - \mathbf{B}\mathbf{A}_{2}$$

$$(3.1.8.5)$$

$$= c(\mathbf{A}_{1}\mathbf{B} - \mathbf{B}\mathbf{A}_{1}) + (\mathbf{A}_{2}\mathbf{B} - \mathbf{B}\mathbf{A}_{2})$$

$$(3.1.8.6)$$

$$= cT(\mathbf{A}_{1}) + T(\mathbf{A}_{2})$$

$$(3.1.8.7)$$

From (3.1.8.7) we conclude that T is a linear transformation from vector space V to V.

3.1.9. Let V be the set of all complex numbers regarded as a vector space over the field of real numbers(usual operations). Find a function from V into V which is a linear transformation on the above vector space, but which is not a linear transformation on $\mathbb C$ i.e., which is not complex linear.

Solution: Let

$$T: V \to V \tag{3.1.9.1}$$

be a function such that,

$$T(x + iy) = Re(x + iy) = x$$
 (3.1.9.2)

$$\implies T: x + iy \rightarrow x$$
 (3.1.9.3)

where $x, y \in \mathbb{R}$.

Let, $\alpha = a + ib$, $\beta = c + id$.

$$T (kα + β) = T (ka + ikb + c + id)$$

$$(3.1.9.4)$$

$$= T (ka + c + i(kb + d))$$

$$(3.1.9.5)$$

$$= ka + c$$

$$= kT (α) + T (β)$$

$$(3.1.9.7)$$

Now, let $z \in V$ such that,

$$z = i$$
 (3.1.9.8)

$$\implies T(z) = T(i) = 0$$
 (3.1.9.9)

We can also write,

$$T(i) = T(i(1)) = iT(1) = i \neq 0$$
 (3.1.9.10)

Thus from (3.1.9.7), T is real linear transformation and from (3.1.9.10), T is not complex linear.

3.1.10. Let **V** be the space of $n \times 1$ matrices over F and let **W** be the space of $m \times 1$ matrices over F. Let **A** be a fixed $m \times n$ matrix over F and let T be the linear transformation from **V** into **W** defined by $T(\mathbf{X}) = \mathbf{A}\mathbf{X}$. Prove that T is the zero transformation if and only if **A** is the zero matrix. **Solution:** If $\mathbf{A}_{m \times n}$ is a zero transformation and $\mathbf{X}_{n \times 1}$ is a vector, then

$$\mathbf{AX} = \mathbf{0}_{m \times 1} \tag{3.1.10.1}$$

Let,

$$\mathbf{A} = \begin{pmatrix} \mathbf{A_1} & \dots & \mathbf{A_j} & \dots & \mathbf{A_n} \end{pmatrix}_{1 \times n}$$
 and

(3.1.10.2)

$$\mathbf{X_{j}} = \begin{pmatrix} x_{1} \\ \vdots \\ x_{j} \\ \vdots \\ x_{n} \end{pmatrix}, \text{ where } x_{i} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

$$(3.1.10.3)$$

If $\mathbf{A}_{m \times n}$ is zero transformation, then for any vector $\mathbf{X}_{n \times 1}$, $\mathbf{A}\mathbf{X} = \mathbf{0}$. Consider,

$$\mathbf{AX_j} = \mathbf{0}_{m \times 1} \qquad (3.1.10.4)$$

$$\left(\mathbf{A_1} \dots \mathbf{A_j} \dots \mathbf{A_n}\right) \begin{pmatrix} x_1 \\ \vdots \\ x_j \\ \vdots \\ x_n \end{pmatrix} = \mathbf{0}_{m \times 1} \qquad (3.1.10.5)$$

From (3.1.10.3) and (3.1.10.5)

$$\mathbf{A_i} = \mathbf{0}_{m \times 1} \text{ for } j = 1, 2, ...n$$
 (3.1.10.6)

Substitute (3.1.10.6) in (3.1.10.2)

$$\mathbf{A} = \begin{pmatrix} \mathbf{0}_{m \times 1} & \mathbf{0}_{m \times 1} & \dots & \mathbf{0}_{m \times 1} \end{pmatrix}_{1 \times n} \quad (3.1.10.7)$$

$$\therefore \mathbf{A} = \mathbf{0}_{m \times n} \tag{3.1.10.8}$$

Hence A is zero matrix.

Let us assume $A_{m\times n}$ is a zero matrix

$$\mathbf{A} = \mathbf{0}_{m \times n} \tag{3.1.10.9}$$

Then,

$$T(\mathbf{X}) = \mathbf{AX} \tag{3.1.10.10}$$

$$= 0.X (3.1.10.11)$$

$$= \mathbf{0}_{m \times 1} , \forall \mathbf{X} \in F$$
 (3.1.10.12)

Hence $T(\mathbf{X}) = \mathbf{A}\mathbf{X}$ is the zero transformation.

From (3.1.10.8) and (3.1.10.12) it is proved that T is the zero transformation if and only if **A** is the zero matrix.

3.1.11. Let V be an *n*-dimensional vector space over

the field \mathbf{F} and let \mathbf{T} be a linear transformation from \mathbf{V} into \mathbf{V} such that the range and null space of \mathbf{T} are identical. Prove that n is even. (Can you give an example of such a linear transformation \mathbf{T})?

Solution: Let **V** and **W** be vector spaces over the field **F** and let **T** be a linear transformation from **V** into **W**. Then,

$$rank(\mathbf{T}) + nullity(\mathbf{T}) = \dim \mathbf{V}$$
 (3.1.11.1)

It is given that range and null space of T are same, let us assume it to be m. Substituting in equation (3.1.11.1)

$$m + m = n \tag{3.1.11.2}$$

$$\implies n = 2m \tag{3.1.11.3}$$

From equation (3.1.11.3), we can say that n is even

Example: Let us consider a vector space \mathbf{V} , such that $\mathbf{V} \in \mathbb{R}^2$ and let us consider a linear transformation $\mathbf{T} : \mathbf{V} \to \mathbf{V}$ defined by $\mathbf{T} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ 0 \end{pmatrix}$ and is given by matrix \mathbf{M}

$$\mathbf{T} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ 0 \end{pmatrix} \tag{3.1.11.4}$$

$$\mathbf{M} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \tag{3.1.11.5}$$

Let us consider basis of \mathbb{R}^2 $\left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\}$ and apply linear transformation on it.

$$\mathbf{T} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{3.1.11.6}$$

$$\mathbf{T} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{3.1.11.7}$$

From (3.1.11.5),

The range of matrix can be found from row reduced echelon form. But as matrix **M** is in RREF form,

the basis for range is given by $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$.

The null space of matrix is,

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{3.1.11.8}$$

$$\implies x_1 = t \quad x_2 = 0 \tag{3.1.11.9}$$

$$\implies \mathbf{X} = \begin{pmatrix} t \\ 0 \end{pmatrix} = t \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{3.1.11.10}$$

The basis for null space is $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$.

$$rank(T) = 1$$
 $nullity(T) = 1$ (3.1.11.11)

$$\dim(\mathbf{V}) = 2 \tag{3.1.11.12}$$

Thus the range and null space are equal, and n is even.

- 3.1.12. Let \mathbb{V} be a vector space and \mathbf{T} a linear transformation from \mathbb{V} into \mathbb{V} . Prove that the following two statements about \mathbf{T} are equivalent.
 - (a) The intersection of the range of T and null space of T is the zero subspace of V.
 - (b) If $\mathbf{T}(\mathbf{T}\alpha) = 0$, then $\mathbf{T}\alpha = 0$.

Solution:

Given	$\mathbf{T}:\mathbb{V} o\mathbb{V}$
To prove	a) $range(\mathbf{T}) \cap nullspace(\mathbf{T}) = \{0\}$ b) If $\mathbf{T}(\mathbf{T}\alpha) = 0$, then $\mathbf{T}\alpha = 0$.

- 3.2 The Algebra of Linear Transformations
- 3.2.1. Let **T** and **U** be the linear operators on \mathbb{R}^2 defined by $\mathbf{T}(x_1, x_2) = (x_2, x_1)$ and $\mathbf{U}(x_1, x_2) = (x_1, 0)$.

$$\mathbf{T}(x_1, x_2) = (x_2, x_1) \tag{3.2.1.1}$$

and

$$\mathbf{U}(x_1, x_2) = (x_1, 0) \tag{3.2.1.2}$$

How would you describe T and U geometrically ?

Solution: Geometrically, in the x-y plane, T is the reflection about the diagonal x = y

Proof(a)	Let $\mathbf{x} \in \mathbb{V}$ and $\mathbf{x} \in range(\mathbf{T}) \cap nullspace(\mathbf{T})$
	then,
	$\mathbf{x} \in range(\mathbf{T})$
	$\mathbf{x} \in nullspace(\mathbf{T})$
	Consider $\mathbf{y} \in \mathbb{V}$ whose
	linear transformation into \mathbb{V} is \mathbf{x} .
	\implies T(y) = x
	since $x \in null space(T)$
	and the sub space is linearly independent
	$\mathbf{T}(\mathbf{x}) = 0$
	from above equations
	$\mathbf{T}(\mathbf{T}(y)) = 0$
	from the definition of linear
	transformation of independent vector space
	$\mathbf{T}(y) = 0$ $\implies \mathbf{x} = 0$
	·
	$\implies \{0\} \subseteq range(\mathbf{T}) \cap nullspace(\mathbf{T})$ $: range(\mathbf{T}) \cap nullspace(\mathbf{T}) = \{0\}$
	$\therefore range(\mathbf{T}) \cap nullspace(\mathbf{T}) = \{0\}$ Hence Proved.
Proof(b)	If $\mathbf{T}(\mathbf{T}\alpha) = 0$
	then, from the definition of linear
	transformation, $\mathbf{T}\alpha$ will
	be in the null space of linear
	transformation T and is linearly
	independent
	$\therefore \mathbf{T}\alpha = 0$

and U is a projection onto the x-axis.

i) Reflection

Let Consider Matrix A as

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \tag{3.2.1.3}$$

The matrix **A** is representation of the linear transformation T across the line y=x with respect to the standard basis.

Eg:

Let $\alpha \in \mathbb{V}$ and

$$\alpha = \begin{pmatrix} 1 & 7 & -1 & -1 \\ -1 & 1 & 2 & 1 \\ 4 & -2 & 0 & -4 \\ 2 & 3 & 4 & -2 \end{pmatrix}$$

linear transformation of α into \mathbb{V} $\mathbf{T}(\alpha) = c\alpha$

then row reduced echelon form of T is

$$rref(\mathbf{T}) = \begin{pmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\implies rank(\mathbf{T}) = 3,$$

 $nullity(\mathbf{T}) = 1$

$$\Rightarrow range(\mathbf{T}) = \begin{pmatrix} 1 & 7 & -1 \\ -1 & 1 & 2 \\ 4 & -2 & 0 \\ 2 & 3 & 4 \end{pmatrix},$$

$$nullspace(\mathbf{T}) = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

 $\therefore range(\mathbf{T}) \cap nullspace(\mathbf{T}) = \{0\}$

Hence proved that the intersection of the range of T and null space of T is the zero subspace of V.

Let suppose

$$\mathbf{x_1} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \mathbf{x_2} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 3 \\ 4 \end{pmatrix}$$
(3.2.1.4)

After applying linear operator T on it,

$$\mathbf{T}(x_1, x_2) = \mathbf{A} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$(3.2.1.5)$$

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

$$(3.2.1.6)$$

Similarly

$$\mathbf{A} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 4 \\ 3 \end{pmatrix}$$
(3.2.1.7)

Hence after applying Operator T on x_1 and x_2

$$\mathbf{x_1} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \mathbf{x_2} = \begin{pmatrix} 4 \\ 3 \end{pmatrix}$$
 (3.2.1.8)

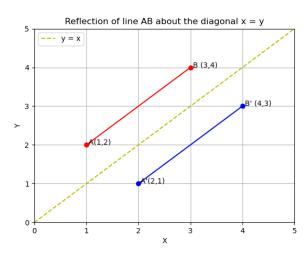


Fig. 3.2.1.1: Reflection of line AB about the x = y

ii) Projection

For projection let Consider Matrix B as

$$\mathbf{B} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.2.1.9}$$

The matrix \mathbf{B} is representation of the linear transformation \mathbf{U} that is projection on x-axis.

Let suppose

$$\mathbf{x_1} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \mathbf{x_2} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 3 \\ 4 \end{pmatrix}$$
 (3.2.1.10)

After applying linear operator U on x_1 and

 x_2

$$\mathbf{T}(x_1, x_2) = \mathbf{U} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (3.2.1.11)$$

$$\implies \mathbf{B} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad (3.2.1.12)$$

Similarly

$$\mathbf{A} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 3 \\ 0 \end{pmatrix} \quad (3.2.1.13)$$

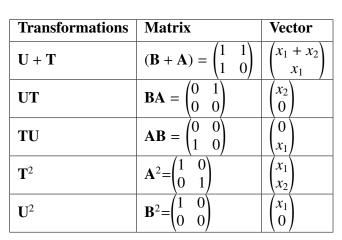


TABLE 3.2.1.1: Summary

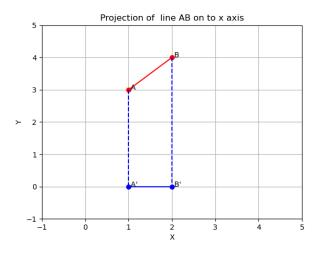


Fig. 3.2.1.2: Projection of AB onto x-axis

Hence after applying Operator U on x_1 and x_2

$$\mathbf{x_1} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \mathbf{x_2} = \begin{pmatrix} 3 \\ 0 \end{pmatrix} \tag{3.2.1.14}$$

b) Give rules like the ones defining **T** and **U** for each of the transformations U + T, UT, TU, T^2 , U^2 . \mathbb{R}^2 into \mathbb{R}^2 is linear transformation? **Solution:** Let **T** and **U** defined by matrices **A** and **B** such that ,

$$T(x) = Ax;$$
 $U(x) = Bx$ (3.2.1.15)

Where.

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}; \quad \mathbf{B} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \qquad (3.2.1.16)$$

Table 3.2.1.1 lists the summary of each Transformations.

3.2.2. Let T be the unique linear operator on C^3 for

which

$$T(\epsilon_1) = \begin{pmatrix} 1 & 0 & i \end{pmatrix}, T(\epsilon_2) = \begin{pmatrix} 0 & 1 & 1 \end{pmatrix},$$
$$T(\epsilon_3) = \begin{pmatrix} i & 1 & 0 \end{pmatrix}$$
$$(3.2.2.1)$$

Is T invertible?

Solution: Let ϵ_i is basis for C^3 such that $T(\epsilon_i)$ is basis for C^3 T is said to be singular if

$$T(\epsilon) = 0 \implies \epsilon \neq 0$$
 (3.2.2.2)

now,

$$\begin{pmatrix} 1 & 0 & i \\ 0 & 1 & 1 \\ i & 1 & 0 \end{pmatrix} \epsilon = 0 \tag{3.2.2.3}$$

consider the row reduced matrix

$$\begin{pmatrix} 1 & 0 & i \\ 0 & 1 & 1 \\ i & 1 & 0 \end{pmatrix} \xrightarrow[R_3 \to R_3 - R_2]{R_3 \to R_3 - iR_1} \begin{pmatrix} 1 & 0 & i \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix} (3.2.2.4)$$

$$\epsilon = c \begin{pmatrix} -i \\ -1 \\ 1 \end{pmatrix} \tag{3.2.2.5}$$

Hence it holds the condition of singularity therefore T is not invertible .

3.2.3. For the linear operator T

$$\mathbf{T} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 3x_1 \\ x_1 - x_2 \\ 2x_1 + x_2 + x_3 \end{pmatrix}$$
 (3.2.3.1)

3.2.4. Let **T** be a linear operator on \mathbb{R}^3 defined by

$$\mathbf{T} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 3x_1 \\ x_1 - x_2 \\ 2x_1 + x_2 + x_3 \end{pmatrix}$$

Is **T** invertible? If so, find a rule for T^{-1} like the one which defines T.

Solution: The transformed vector can be rewritten by expanding the columns as follows

$$\begin{pmatrix} 3x_1 \\ x_1 - x_2 \\ 2x_1 + x_2 + x_3 \end{pmatrix} = \begin{pmatrix} 3 \\ 1 \\ 2 \end{pmatrix} x_1 + \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix} x_2 + \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} x_3$$

$$(3.2.4.1)$$

$$= \begin{pmatrix} 3 & 0 & 0 \\ 1 & -1 & 0 \\ 2 & 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \quad (3.2.4.2)$$

$$\implies \mathbf{T} = \begin{pmatrix} 3 & 0 & 0 \\ 1 & -1 & 0 \\ 2 & 1 & 1 \end{pmatrix} \quad (3.2.4.3)$$

Using Gauss-Jordan Elimination to find the inverse of T, if it exists

$$\begin{pmatrix}
3 & 0 & 0 & | & 1 & 0 & 0 \\
1 & -1 & 0 & | & 0 & 1 & 0 \\
2 & 1 & 1 & | & 0 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_1 \leftarrow \frac{R_1}{3}} \begin{pmatrix}
1 & 0 & 0 & | & \frac{1}{3} & 0 & 0 \\
1 & -1 & 0 & | & 0 & 1 & 0 \\
2 & 1 & 1 & | & 0 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_2 \leftarrow R_2 - R_1} \begin{pmatrix}
1 & 0 & 0 & | & \frac{1}{3} & 0 & 0 \\
0 & -1 & 0 & | & -\frac{1}{3} & 1 & 0 \\
0 & 1 & 1 & | & -\frac{2}{3} & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_2 \leftarrow -R_2} \begin{pmatrix}
1 & 0 & 0 & | & \frac{1}{3} & 0 & 0 \\
0 & 1 & 0 & | & \frac{1}{3} & -1 & 0 \\
0 & 1 & 1 & | & -\frac{2}{3} & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2} \begin{pmatrix}
1 & 0 & 0 & | & \frac{1}{3} & 0 & 0 \\
0 & 1 & 0 & | & \frac{1}{3} & -1 & 0 \\
0 & 1 & 0 & | & \frac{1}{3} & -1 & 0 \\
0 & 0 & 1 & | & -1 & 1 & 1
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2} \begin{pmatrix}
1 & 0 & 0 & | & \frac{1}{3} & 0 & 0 \\
0 & 1 & 0 & | & \frac{1}{3} & -1 & 0 \\
0 & 0 & 1 & | & -1 & 1 & 1
\end{pmatrix}$$

$$\xrightarrow{(3.2.4.8)}$$

Since rank(T) = 3, T is invertible and the

inverse is

$$\mathbf{T}^{-1} = \begin{pmatrix} \frac{1}{3} & 0 & 0\\ \frac{1}{3} & -1 & 0\\ -1 & 1 & 1 \end{pmatrix}$$
 (3.2.4.9)

Now consider any vector $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in \mathbf{R}^3$, then

$$\mathbf{T}^{-1} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} \frac{1}{3} & 0 & 0 \\ \frac{1}{3} & -1 & 0 \\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$
 (3.2.4.10)
$$= \begin{pmatrix} \frac{x_1}{3} - x_2 \\ -x_1 + x_2 + x_3 \end{pmatrix}$$
 (3.2.4.11)

Therefore the transformation T^{-1} is defined on \mathbb{R}^3 as

$$\mathbf{T}^{-1} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} \frac{x_1}{3} \\ \frac{x_1}{3} - x_2 \\ -x_1 + x_2 + x_3 \end{pmatrix}$$
(3.2.4.12)

Prove that

$$(\mathbf{T}^2 - I)(\mathbf{T} - 3I) = 0$$
 (3.2.4.13)

Solution: Expressing (3.2.3.1) in matrix form

$$\mathbf{T} = \begin{pmatrix} 3 & 0 & 0 \\ 1 & -1 & 0 \\ 2 & 1 & 1 \end{pmatrix} \tag{3.2.4.14}$$

The characteristic equation of **T** is given as follows,

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

$$(3.2.4.15)$$

$$\implies (3 - \lambda)(-1 - \lambda)(1 - \lambda) = 0$$

$$\implies (\lambda - 3)(1 + \lambda)(1 - \lambda) = 0$$

$$\implies (\lambda - 3)(1 - \lambda^{2}) = 0$$

$$\implies (\lambda^{2} - 1)(\lambda - 3) = 0 \quad (3.2.4.16)$$

By the Cayley-Hamilton theorem, We can write (3.2.4.16) as

$$(\mathbf{T}^2 - I)(\mathbf{T} - 3I) = 0$$
 (3.2.4.17)

3.2.5. Let \mathbb{C} be the complex vector space of 2×2

matrices with complex entries. Let

$$\mathbf{B} = \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix} \tag{3.2.5.1}$$

and let **T** be the linear operator on $\mathbb{C}^{2\times 2}$ defined by $\mathbf{T}(\mathbf{A}) = \mathbf{B}\mathbf{A}$. What is the rank of **T**? Can you describe \mathbf{T}^2 ?

Solution: An ordered basis for $\mathbb{C}^{2\times 2}$ is given by

$$\mathbf{A_{11}} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \qquad \mathbf{A_{12}} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \qquad (3.2.5.2)$$

$$\mathbf{A_{21}} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \qquad \mathbf{A_{22}} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \qquad (3.2.5.3)$$

Now, we compute

$$T(A_{11}) = BA_{11} (3.2.5.4)$$

$$= \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.2.5.5}$$

$$= \begin{pmatrix} 1 & 0 \\ -4 & 0 \end{pmatrix} \tag{3.2.5.6}$$

from (3.2.5.6) we have

$$\mathbf{T}(\mathbf{A}_{11}) = \mathbf{A}_{11} - 4\mathbf{A}_{21} \tag{3.2.5.7}$$

$$T(A_{12}) = BA_{12} (3.2.5.8)$$

$$= \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$
 (3.2.5.9)

$$= \begin{pmatrix} 0 & 1 \\ 0 & -4 \end{pmatrix} \tag{3.2.5.10}$$

from (3.2.5.10), we have

$$\mathbf{T}(\mathbf{A}_{12}) = \mathbf{A}_{12} - 4\mathbf{A}_{22} \tag{3.2.5.11}$$

$$T(A_{21}) = BA_{21} (3.2.5.12)$$

$$= \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$$
 (3.2.5.13)

$$= \begin{pmatrix} -1 & 0\\ 4 & 0 \end{pmatrix} \tag{3.2.5.14}$$

from (3.2.5.14), we have

$$\mathbf{T}(\mathbf{A}_{21}) = -\mathbf{A}_{11} + 4\mathbf{A}_{21} \tag{3.2.5.15}$$

$$T(A_{22}) = BA_{22} (3.2.5.16)$$

$$= \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
 (3.2.5.17)

$$= \begin{pmatrix} 0 & -1 \\ 0 & 4 \end{pmatrix} \tag{3.2.5.18}$$

from (3.2.5.18), we have

$$\mathbf{T}(\mathbf{A}_{22}) = -\mathbf{A}_{12} + 4\mathbf{A}_{22} \tag{3.2.5.19}$$

Now, by (3.2.5.7), (3.2.5.11), (3.2.5.15) and (3.2.5.19) we write matrix of the linear transformation as follows

$$\mathbf{P} = \begin{pmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ -4 & 0 & 4 & 0 \\ 0 & -4 & 0 & 4 \end{pmatrix}$$
 (3.2.5.20)

Also, we know that the rank of a linear transformation is same as the rank of the matrix of the linear transformation. Thus, we find the rank of matrix **P**.

$$\begin{pmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ -4 & 0 & 4 & 0 \\ 0 & -4 & 0 & 4 \end{pmatrix} \xrightarrow{r_3 = r_3 + 4r_1} \begin{pmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 0 \\ 0 & -4 & 0 & 4 \end{pmatrix}$$

$$(3.2.5.21)$$

$$\begin{pmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 0 \\ 0 & -4 & 0 & 4 \end{pmatrix} \xrightarrow{r_4 = r_4 + 4r_1} \begin{pmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(3.2.5.22)$$

from (3.2.5.22), we found out that $rank(\mathbf{T}) = 2$. Now, we compute

$$T^{2}(A) = T(T(A))$$
 (3.2.5.23)

$$= \mathbf{T}(\mathbf{B}\mathbf{A}) \tag{3.2.5.24}$$

$$= \mathbf{B}^2 \mathbf{A} \tag{3.2.5.25}$$

where

$$\mathbf{B^2} = \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix}$$
 (3.2.5.26)

$$= \begin{pmatrix} 5 & -5 \\ -20 & 20 \end{pmatrix} \tag{3.2.5.27}$$

3.2.6. Let T be a linear transformation from \mathbb{R}^3 into \mathbb{R}^2 , and let U be a linear transformation from \mathbb{R}^2 into \mathbb{R}^3 . Prove that the transformation UT is not invertible. Generalize the theorem.

Solution: Let $\mathbf{v}, \mathbf{x} \in \mathbb{R}^3$ and $\mathbf{w} \in \mathbb{R}^2$. Table 3.2.6.1 shows that maximum rank the transformation matrix \mathbf{C} can have is 2.

$$Rank(\mathbf{C}) = 2$$
 (3.2.6.1)

$$dim(\mathbf{C}) = 3$$
 (3.2.6.2)

$$\implies Rank(\mathbf{C}) < dim(\mathbf{C})$$
 (3.2.6.3)

Therefore from the equation (3.2.6.3), we can say transformation UT is not invertible. Generalizing the proof, for n > m and considering vectors $\mathbf{v}, \mathbf{x} \in \mathbb{R}^n$ and $\mathbf{w} \in \mathbb{R}^m$. From the Table 3.2.6.2,

$$Rank(\mathbf{C}) = m \tag{3.2.6.4}$$

$$dim(\mathbf{C}) = n \tag{3.2.6.5}$$

$$\implies Rank(\mathbf{C}) < dim(\mathbf{C})$$
 (3.2.6.6)

From equation (3.2.6.6)we can say that the transformation UT is not invertible. Let the

vectors
$$\mathbf{v} = \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix} \in \mathbb{R}^3$$
 and $\mathbf{w} = \begin{pmatrix} 4 \\ 2 \end{pmatrix} \in \mathbb{R}^2$.

a) Calculating transformation matrix A,

$$T(\mathbf{v}) = \mathbf{A}\mathbf{v} \tag{3.2.6.7}$$

$$\begin{pmatrix} 4 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 0 & -1 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix}$$
 (3.2.6.8)

$$\begin{pmatrix} 1 & 0 & 1 \\ 1 & 0 & -1 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} = rref(\mathbf{A})$$

$$(3.2.6.9)$$

$$\implies$$
 Rank(**A**) = 2 (3.2.6.10)

b) Calculating transformation matrix **B**,

$$U(\mathbf{w}) = \mathbf{B}\mathbf{w} \tag{3.2.6.11}$$

$$\begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix} = \begin{pmatrix} \frac{3}{4} & 2 \\ 1 & -1 \\ 0 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 4 \\ 2 \end{pmatrix}$$
 (3.2.6.12)

$$\begin{pmatrix} \frac{3}{4} & 2\\ 1 & -1\\ 0 & \frac{1}{2} \end{pmatrix} \sim \begin{pmatrix} 1 & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} = rref(\mathbf{B}) \quad (3.2.6.13)$$

$$\implies Rank(\mathbf{B}) = 2 \quad (3.2.6.14)$$

c) Now for the transformation UT, calculating

the transformation matrix C,

$$UT: \mathbb{R}^3 \to \mathbb{R}^3 \tag{3.2.6.15}$$

$$\implies UT(\mathbf{x}) = \mathbf{C}\mathbf{x} \tag{3.2.6.16}$$

$$U(T(\mathbf{x})) = \mathbf{B}(\mathbf{A}\mathbf{x}) \tag{3.2.6.17}$$

$$\implies$$
 C = **BA** (3.2.6.18)

$$\mathbf{C} = \begin{pmatrix} \frac{3}{4} & 2\\ 1 & -1\\ 0 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 1 & 0 & 1\\ 1 & 0 & -1 \end{pmatrix} = \begin{pmatrix} \frac{3}{4} & 0 & \frac{3}{4}\\ 0 & 0 & 2\\ \frac{1}{2} & 0 & \frac{-1}{2} \end{pmatrix}$$
(3.2.6.19)

$$\begin{pmatrix} \frac{3}{4} & 0 & \frac{3}{4} \\ 0 & 0 & 2 \\ \frac{1}{2} & 0 & \frac{-1}{2} \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} = rref(\mathbf{C})$$
(3.2.6.20)

$$\implies Rank(\mathbf{C}) = 2$$
 (3.2.6.21)

$$dim(\mathbf{C}) = 3$$
 (3.2.6.22)

As $Rank(\mathbb{C}) < dim(\mathbb{C})$, transformation UT is not invertible.

3.2.7. Find two linear operators \mathbf{T} and \mathbf{U} on \mathbf{R}^2 such that $\mathbf{T}\mathbf{U} = 0$ but $\mathbf{U}\mathbf{T} \neq 0$

Solution: Let,

$$\mathbf{x}, \mathbf{y} \in \mathbf{R}^2 \tag{3.2.7.1}$$

Let T and U be given by the matrices

$$T(x) = Ax;$$
 $U(x) = Bx$ (3.2.7.2)

where,

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}; \quad \mathbf{B} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.2.7.3}$$

$$\mathbf{T}(a\mathbf{x} + \mathbf{v}) = a\mathbf{T}\mathbf{x} + \mathbf{T}\mathbf{v} \tag{3.2.7.4}$$

$$\mathbf{U}(a\mathbf{x} + \mathbf{y}) = a\mathbf{U}\mathbf{x} + \mathbf{U}\mathbf{y} \tag{3.2.7.5}$$

From (3.2.7.4) and (3.2.7.5), we can tell that **T** and **U** are linear operators. Now,

$$\mathbf{TU} = \mathbf{AB} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \mathbf{0}$$
(3.2.7.6)

$$\mathbf{UT} = \mathbf{BA} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \neq \mathbf{0}$$
(3.2.7.7)

From (3.2.7.6) and (3.2.7.7) it can be observed that TU = 0 but $UT \neq 0$

Transformation	Matrix Representation	Dimension	Max Rank of transformation matrix
$T: \mathbb{R}^3 \to \mathbb{R}^2$	$T(\mathbf{v}) = \mathbf{A}\mathbf{v}$	$\mathbf{A}: 2 \times 3$	$Rank(\mathbf{A}) = 2$
$U: \mathbb{R}^2 \to \mathbb{R}^3$	$U(\mathbf{w}) = \mathbf{B}\mathbf{w}$	$\mathbf{B}: 3 \times 2$	$Rank(\mathbf{B}) = 2$
$UT: \mathbb{R}^3 \to \mathbb{R}^3$	$UT(\mathbf{x}) = \mathbf{C}\mathbf{x}$	$\mathbf{C}: 3 \times 3$	$Rank(\mathbf{C}) \leq min(Rank(\mathbf{B}), Rank(\mathbf{A}))$
	$U(T(\mathbf{x})) = \mathbf{B}(\mathbf{A}\mathbf{x})$		$Rank(\mathbf{C}) = 2$
	C = AB		

TABLE 3.2.6.1: Proof for non-invertibility of the transformation UT where $T: \mathbb{R}^3 \to \mathbb{R}^2$ and $U: \mathbb{R}^2 \to \mathbb{R}^3$

Transformation	Matrix Representation	Dimension	Max Rank of transformation matrix
$T:\mathbb{R}^n\to\mathbb{R}^m$	$T(\mathbf{v}) = \mathbf{A}\mathbf{v}$	$\mathbf{A}: m \times n$	$Rank(\mathbf{A}) = m$
$U:\mathbb{R}^m o \mathbb{R}^n$	$U(\mathbf{w}) = \mathbf{B}\mathbf{w}$	$\mathbf{B}: n \times m$	$Rank(\mathbf{B}) = m$
$UT: \mathbb{R}^n \to \mathbb{R}^n$	$UT(\mathbf{x}) = \mathbf{C}\mathbf{x}$	$\mathbf{C}: n \times n$	$Rank(\mathbf{C}) \leq min(Rank(\mathbf{B}), Rank(\mathbf{A}))$
	$U(T(\mathbf{x})) = \mathbf{B}(\mathbf{A}\mathbf{x})$		$Rank(\mathbf{C}) = m$
	C = AB		

TABLE 3.2.6.2: Generalization of the proof

3.2.8. Let **V** be a vector space over the field **F** and **T** is a linear operator on **V**. If $\mathbf{T}^2 = 0$, what can you say about the relation of the range of **T** to the null space of **T**? Give an example of linear operator **T** on \mathbf{R}^2 such that $\mathbf{T}^2 = 0$ but $\mathbf{T} \neq 0$.

Solution: Given,

$$\mathbf{T}: \mathbf{V} \to \mathbf{V} \tag{3.2.8.1}$$

Now, T^2 is also a linear operator as,

$$\mathbf{T}^{2}(c\alpha) = \mathbf{T}(\mathbf{T}(c\alpha)) = \mathbf{T}(c\mathbf{T}(\alpha)) \quad (3.2.8.2)$$
$$= c\mathbf{T}(\mathbf{T}(\alpha)) = c\mathbf{T}^{2}(\alpha) \quad (3.2.8.3)$$

Let some vector $y \in \text{Range}(T)$ then there exists $x \in V$ such that,

$$\mathbf{T}(\mathbf{x}) = \mathbf{v} \tag{3.2.8.4}$$

Now given that,

$$T^{2}(x) = 0$$
 (3.2.8.5)
 $\Rightarrow T(T(x)) = 0$ (3.2.8.6)
 $T(y) = 0$ (3.2.8.7)

 \therefore y lies in the Null space of T. Hence T is singular. Thus, the range of T must be contained in Null space of T i.e., Range(T) \subseteq NullSpace(T)

Example:

$$\mathbf{T}: \mathbf{R}^2 \to \mathbf{R}^2 \tag{3.2.8.8}$$

Consider,

$$\mathbf{T}(\mathbf{x}) = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \mathbf{x} \tag{3.2.8.9}$$

$$\implies$$
 T \neq 0 (3.2.8.10)

Now,

$$\mathbf{T}^2: \mathbf{R}^2 \to \mathbf{R}^2 \tag{3.2.8.11}$$

$$\mathbf{T}^{2}(\mathbf{x}) = \mathbf{T}(\mathbf{T}(\mathbf{x})) \tag{3.2.8.12}$$

$$= \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \mathbf{x} = \mathbf{0} \tag{3.2.8.13}$$

$$\implies \mathbf{T}^2(\mathbf{x}) = \mathbf{0} \tag{3.2.8.14}$$

Thus T^2 is a zero transformation, $T^2 = 0$. Now, Kernel of **T** is given by,

$$\mathbf{T}(\mathbf{x}) = \mathbf{0} \tag{3.2.8.15}$$

$$\begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{3.2.8.16}$$

$$\implies x = 0 \tag{3.2.8.17}$$

Thus,

$$\mathbf{Ker}(\mathbf{T}) = y \begin{pmatrix} 0 \\ 1 \end{pmatrix}; y \in \mathbf{R}$$
 (3.2.8.18)

Now,

$$\label{eq:Range} \begin{aligned} \textbf{Range}(\textbf{T}) &= \textbf{ColumnSpace} \left\{ \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \right\} \\ & (3.2.8.19) \end{aligned}$$

$$=k \begin{pmatrix} 0\\1 \end{pmatrix}; k \in \mathbf{R} \tag{3.2.8.20}$$

Thus for the example, Range(\mathbf{T}) = Kernel(\mathbf{T}) and from (3.2.8.10), (3.2.8.14) it is clear that $\mathbf{T}^2 = 0$ but $\mathbf{T} \neq 0$.

3.2.9. Let **T** be a linear operator on the finite-dimensional space **V**. Support there is a linear operator **U** on **V** such that TU = I. Prove that **T** is invertible and $U = T^{-1}$. Give an example which shows that this is false when **V** is not finite-dimensional.

Solution: Theorem

Theorem 3.1. Let f be a function from X into Y. We say that f is invertible if there is a function g from Y to X such that

- a) $g \circ f$ is the identity function on X i.e. $g \circ f = I$. Here, g will be onto and f will be one-one.
- b) $f \circ g$ is the identity function on Y i.e. $f \circ g = I$. Here, f will be onto and g will be one-one.

Theorem 3.2. Let V and W be finite dimensional vector spaces such that dim $V = \dim W$. If T is a linear transformation from V into W, then the following are equivalent:

- a) T is non-singular
- b) T is onto

If any of the above two condition is satisfied then T is invertible.

a) We are given **V** which is a finite dimensional vector space, with the following linear operators defined as:-

$$\mathbf{T}: \mathbf{V} \to \mathbf{V} \tag{3.2.9.1}$$

$$\mathbf{U}: \mathbf{V} \to \mathbf{V} \tag{3.2.9.2}$$

The linear operators also satisfies the condition

$$TU = I$$
 (3.2.9.3)

Where I is an Identity transformation. This

identity transformation can be written as

$$\mathbf{I}: \mathbf{V} \to \mathbf{V} \tag{3.2.9.4}$$

$$\implies$$
 TU: **V** \rightarrow **V** (3.2.9.5)

$$\implies$$
 T[U(V)] = V (3.2.9.6)

From theorem (3.1) we can say that **U** must be one-one and **V** must be onto.

From theorem (3.2) we can say that **T** is invertible.

Now we know that

$$\mathbf{T}\mathbf{T}^{-1} = \mathbf{I} \tag{3.2.9.7}$$

Comparing (3.2.9.3) and (3.2.9.7) we get

$$TT^{-1} = I = TU$$
 (3.2.9.8)

Multiply both sides with T^{-1}

$$\mathbf{T}^{-1}(\mathbf{T}\mathbf{T}^{-1}) = \mathbf{T}^{-1}(\mathbf{T}\mathbf{U})$$
 (3.2.9.9)

$$\mathbf{T}^{-1}\mathbf{I} = (\mathbf{T}^{-1}\mathbf{T})\mathbf{U} \qquad (3.2.9.10)$$

$$T^{-1} = IU$$
 (3.2.9.11)

$$\therefore \mathbf{T}^{-1} = \mathbf{U} \tag{3.2.9.12}$$

b) Let **D** be a differential operator $\mathbf{D}: \mathbf{V} \to \mathbf{V}$, where **V** is a space of polynomial functions in one variable over **R**.

$$\mathbf{D}(c_0 + c_1 x + \dots + c_n x^n) = c_1 + c_2 x + \dots + c_n x^{n-1}$$
(3.2.9.13)

And $U: V \rightarrow V$ be linear operator such that

$$\mathbf{U}(c_0 + c_1 x + \dots + c_n x^n) = c_0 x + \frac{c_1 x^2}{2} + \dots + \frac{c_n x^{n+1}}{n+1}$$
(3.2.9.14)

Therefore, the linear operator $UD : V \rightarrow V$ will be $UD(c_0 + c_1x + ... + c_nx^n)$

$$= \mathbf{U} \left[\mathbf{D} \left(c_0 + c_1 x + \dots + c_n x^n \right) \right]$$

$$= \mathbf{U} \left[c_1 + c_2 x + \dots + c_n x^{n-1} \right]$$

$$= c_1 x + \frac{c_2 x^2}{2} + \dots + \frac{c_n x^n}{n}$$

$$= c_1 x + c_2 x^2 + \dots + c_n x^n$$

$$\neq \mathbf{I}$$
(3.2.9.15)

singular	A linear transformation $T: \mathbb{R}^n \to \mathbb{R}^n$ is said to be singular if \exists some non-zero $\mathbf{X} \in \mathbb{R}^n$ s.t $\mathbf{A}\mathbf{X} = 0$ i.e $Nullity(A) \neq 0$.
	From rank-nullity theorem we can say $rank(A) < n$
non-singular	A linear transformation $T: \mathbb{R}^n \to \mathbb{R}^m$ is said to be non-singular if $\mathbf{AX} = 0$ implies $\mathbf{X} = 0$ i.e $Nullity(A) = 0$
onto	A linear transformation $T: \mathbb{R}^n \to \mathbb{R}^m$, $m \le n$ is said to be onto if for every $\mathbf{b} \in \mathbb{R}^m$, $\mathbf{A}\mathbf{X} = \mathbf{b}$ has at least one solution $\mathbf{X} \in \mathbb{R}^n$
	i.e $dim(Col(\mathbf{A})) = m$ or $Rank(\mathbf{A}) = m$
	If $m > n$, then $\mathbf{AX} = \mathbf{b}$ has no solution because rank-nullity theorem is not satisfied.

TABLE 3.2.10.1

Now, the linear operator $\mathbf{DU} : \mathbf{V} \to \mathbf{V}$ will be $\mathbf{DU}(c_0 + c_1 x + ... + c_n x^n)$

$$= \mathbf{D} \left[\mathbf{U} \left(c_0 + c_1 x + \dots + c_n x^n \right) \right]$$

$$= \mathbf{D} \left[c_0 x + \frac{c_1 x^2}{2} + \dots + \frac{c_n x^{n+1}}{n+1} \right]$$

$$= c_0 + \frac{2c_2 x}{2} + \dots + \frac{(n+1)c_n x^n}{n+1}$$

$$= c_0 + c_1 x + c_2 x^2 + \dots + c_n x^n$$

$$= \mathbf{I} \qquad (3.2.9.16)$$

From (3.2.9.15) and (3.2.9.16) we see that DU = I, but $UD \neq I$.

- 3.2.10. Let **A** be an $m \times n$ matrix with entries in F and let T be the linear transformation from $F^{n \times 1}$ into $F^{m \times l}$ defined by $T(\mathbf{X}) = \mathbf{A}\mathbf{X}$. Show that
 - a) if m < n it may happen that T is onto without being non-singular
 - b) if m > n we may have T non-singular but not onto.

Solution: Proof

a) m < n

Let,
$$T: \mathbb{R}^3 \to \mathbb{R}^2$$
 (3.2.10.1)

$$T(\mathbf{X}) = \mathbf{AX} = \mathbf{b} \tag{3.2.10.2}$$

Let,
$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix}$$
 (3.2.10.3)

Consider,
$$\mathbf{X} = \begin{pmatrix} 2\\4\\1 \end{pmatrix}$$
 (3.2.10.4)

$$\implies \mathbf{AX} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 4 \\ 1 \end{pmatrix} \quad (3.2.10.5)$$

$$= \begin{pmatrix} 6 \\ 5 \end{pmatrix} \tag{3.2.10.6}$$

Hence T is onto.

Consider,
$$\mathbf{X} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$
 (3.2.10.7)

$$\implies \mathbf{AX} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \quad (3.2.10.8)$$
$$= \mathbf{0} \qquad (3.2.10.9)$$

Since $\exists X \neq 0$ such that AX = 0, T is singular.

.. T is both onto and singular.

b) m > n

Let,
$$T : \mathbb{R}^3 \to \mathbb{R}^2$$
 (3.2.10.10)

$$T(\mathbf{X}) = \mathbf{AX} = \mathbf{b} \tag{3.2.10.11}$$

Let,
$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}$$
 (3.2.10.12)

Consider,
$$\mathbf{X} = \begin{pmatrix} -1 \\ 2 \end{pmatrix}$$
 (3.2.10.13)

$$\implies \mathbf{AX} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \end{pmatrix} \quad (3.2.10.14)$$

$$= \begin{pmatrix} -1\\2\\-1 \end{pmatrix}$$
 (3.2.10.15)

(3.2.10.16)

- .. T is not onto, and is also non-singular.
- 3.2.11. Let V be a finite-dimensional vector space and let T be a linear operator on V. Suppose that rank (T^2) = rank (T). Prove that the range and null space of T are disjoint, i.e., have only the zero vector in common. **Solution:** See Table 3.2.11.1.
- 3.2.12. Let p, m, n be positive integers and \mathbb{F} a field.Let \mathbf{V} be the space of $m \times n$ matrices over \mathbb{F} and \mathbf{W} the space of $p \times n$ matrices over \mathbb{F} .Let \mathbf{B} be a fixed $p \times m$ matrix and let \mathbb{T} be the linear transformation from \mathbf{V} into \mathbf{W} defined by $\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A}$.Prove that \mathbb{T} is invertible if and only if p = m and \mathbf{B} is an invertible $m \times m$ matrix. **Solution:**

$$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A} \tag{3.2.12.1}$$

So, **B** is the transformation matrix. **B** is invertible if

a) \mathbb{T} is one to one mapping, that is

$$BA = BA'$$
 (3.2.12.2)

$$\implies \mathbf{A} = \mathbf{A}' \tag{3.2.12.3}$$

b) \mathbb{T} must be onto, that is range(**B**)=**W**

Case 1: Let us assume that \mathbb{T} is invertible with inverse transformation \mathbb{T}_1 from \mathbf{W} to \mathbf{V} that satisfies

$$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A} \in \mathbf{W} \qquad (3.2.12.4)$$

$$\implies \mathbb{T}_1(\mathbf{B}\mathbf{A}) = \mathbf{A} \in \mathbf{V}$$
 (3.2.12.5)

$$\dim(\mathbf{V}) = mn, \dim(\mathbf{W}) = pn \qquad (3.2.12.6)$$

Since \mathbb{T} is one-one mapping, the zero vector in \mathbf{V} , $\mathbf{0}_{m \times n}$ is uniquely mapped to

$$\mathbb{T}(\mathbf{0}_{m\times n}) = \mathbf{B}\mathbf{0}_{m\times n} = \mathbf{0}_{p\times n} \tag{3.2.12.7}$$

So,
$$BA = 0 \iff A = 0$$
 (3.2.12.8)

Let $\{V_1, V_2, \dots, V_{mn}\}$ be the basis for V

$$c_1 \mathbf{V}_1 + c_2 \mathbf{V}_2 + \ldots + c_{mn} \mathbf{V}_{mn} = \mathbf{0}$$
 (3.2.12.9)

$$\iff c_1, c_2, \dots, c_{mn} \in \mathbb{F} = 0 \quad (3.2.12.10)$$

Any matrix $A \in V$ can be written as

$$\mathbf{A} = \sum_{i=1}^{mn} \alpha_i \mathbf{V}_i \tag{3.2.12.11}$$

Since \mathbb{T} is onto, any matrix $\mathbf{C} \in \mathbf{W}$ can be expressed as

$$\mathbf{C} = \mathbf{B} \left(\sum_{i=1}^{mn} \alpha_i \mathbf{V}_i \right)$$
 (3.2.12.12)

$$=\sum_{i=1}^{mn}\alpha_i(\mathbf{B}\mathbf{V}_i) \tag{3.2.12.13}$$

So, the set $S = \{BV_1, BV_2, ..., BV_{mn}\}$ forms basis of **W** if all matrices in it are linearly

Let A be an $m \times n$ matrix with entries in F and			
	let T be the linear transformation from $F^{n\times 1}$ into $F^{m\times l}$		
	defined by $T(\mathbf{X}) = \mathbf{A}\mathbf{X}$. If,		
	m < n $m > n$		
singular Since $rank(\mathbf{A}) < n$, by definition T is singular Consider an non-singular T such that $rank(\mathbf{A}) > n$			
onto	onto Since $m < n$, by definition T can be onto Since $m > n$, by definition T is not onto.		

Given	$T: V \to V$ be a linear operator. $Rank(T^2)=Rank(T)$
Rank of T and T^2	Let $(e_1, e_2,, e_n)$ be a basis for T , then $Rank(T)$ is linearly independent vectors in the set $(Te_1, Te_2,, Te_n)$ Let, $Rank(T) = r = Rank(T^2)$
Rank Nullity Theorem	If Rank(T)=r then $(Te_1, Te_2,, Te_r)$ is the basis of range T. Similarly for T^2 , $(T^2e_1, T^2e_2,, T^2e_r)$ is the basis of range T^2
v∈ range(T) v∈ nullspace(T)	$\mathbf{v} = c_1 T e_1 + c_2 T e_2 + \dots + c_r T e_r$ $T(\mathbf{v}) = 0$ $T(c_1 T e_1 + c_2 T e_2 + \dots + c_r T e_r) = 0$ $c_1 T^2 e_1 + c_2 T^2 e_2 + \dots + c_r T^2 e_r = 0$ But, $(T^2 e_1, T^2 e_2, \dots, T^2 e_r)$ is the basis of range T^2 So, $c_1 = c_2 = \dots = c_r = 0$ Substituting these in \mathbf{v} we get $\mathbf{v} = 0$
Conclusion	Hence from above it can be seen that when \mathbf{v} belongs to both $\mathrm{range}(T)$ and $\mathrm{nullspace}(T)$ then \mathbf{v} is a zero vector. Hence, $\mathrm{range}(T)$ and $\mathrm{nullspace}(T)$ are disjoint.

TABLE 3.2.11.1

Parameter	Description
p, m, n	Positive integers
F	Field
V	Space of $m \times n$ matrices
	over F
W	Space of $p \times n$ matrices
	over F
В	Fixed $p \times m$ matrix
Linear transformation	$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A}$
$\mathbb{T}:\mathbf{V} o\mathbf{W}$	

TABLE 3.2.12.1: Input Parameters

independent.

$$c_{1}(\mathbf{B}\mathbf{V}_{1}) + c_{2}(\mathbf{B}\mathbf{V}_{2}) + \dots + c_{mn}(\mathbf{B}\mathbf{V}_{mn}) = \mathbf{0}$$

$$(3.2.12.14)$$

$$\mathbf{B}(c_{1}\mathbf{V}_{1} + c_{2}\mathbf{V}_{2} + \dots + c_{mn}\mathbf{V}_{mn}) = \mathbf{0}$$

$$(3.2.12.15)$$

$$(3.2.12.8) \implies c_{1}\mathbf{V}_{1} + \dots + c_{mn}\mathbf{V}_{mn} = 0$$

$$(3.2.12.16)$$

$$\iff c_{1}, c_{2}, \dots, c_{mn} = 0 \text{ (from (3.2.12.10))}$$

$$(3.2.12.17)$$

So, the set S with cardinality mn is basis for W

$$(3.2.12.6) \implies pn = mn$$
 (3.2.12.18)
 $p = m$ (3.2.12.19)

(3.2.12.8),(3.2.12.19) prove that **B** is invertible $m \times m$ matrix. Case 2: Consider p = m and **B** is an invertible $m \times m$ matrix.

Verifying if \mathbb{T} is onto,

Let the set of matrices $\{A_1, A_2, \dots, A_{mn}\}$ be the basis for V

Any matrix $A \in V$ can be written as

$$\mathbf{A} = \sum_{i=1}^{mn} \alpha_i \mathbf{A}_i \tag{3.2.12.20}$$

where $\alpha_i \in \mathbb{F}$ The set $\mathbf{M} = \{\mathbf{B}\mathbf{A}_1, \mathbf{B}\mathbf{A}_2, \dots, \mathbf{B}\mathbf{A}_{mn}\}$ lie in \mathbf{W}

$$c_1(\mathbf{B}\mathbf{A}_1) + c_2(\mathbf{B}\mathbf{A}_2) + \ldots + c_{mn}(\mathbf{B}\mathbf{A}_{mn}) = \mathbf{0}$$
(3.2.12.21)

$$\implies$$
 B $(c_1$ **A**₁ + c_2 **A**₂ + ... + c_{mn} **A**_{mn} $) =$ **0** (3.2.12.22)

Since **B** is non-singular,

$$(c_1\mathbf{A}_1 + c_2\mathbf{A}_2 + \dots + c_{mn}\mathbf{A}_{mn}) = \mathbf{0} \quad (3.2.12.23)$$

 $\iff c_1, c_2, \dots, c_{mn} = 0 \quad (3.2.12.24)$

because $\{A_1, A_2, \dots, A_{mn}\}$ are linearly independent

So,M forms basis for W

Any vector $C \in W$ can be written as

$$\mathbf{C} = \sum_{i=1}^{mn} \beta_i \mathbf{B} \mathbf{A}_i \text{ where } \beta_i \in \mathbb{F} \qquad (3.2.12.25)$$
$$= \mathbf{B}(\sum_{i=1}^{mn} \beta_i \mathbf{A}_i) \qquad (3.2.12.26)$$

$$=\mathbf{B}(\sum_{i=1}^{mn}\beta_i\mathbf{A}_i) \qquad (3.2.12.26)$$

$$=$$
 BA (from (3.2.12.20)) (3.2.12.27)

So,range(B)=W

Consider the matrix $A, A' \in V$ such that

$$BA = BA'$$
 (3.2.12.28)

$$\mathbf{B}^{-1}(\mathbf{B}\mathbf{A}) = \mathbf{B}^{-1}(\mathbf{B}\mathbf{A}') \tag{3.2.12.29}$$

$$(\mathbf{B}^{-1}\mathbf{B})\mathbf{A} = (\mathbf{B}^{-1}\mathbf{B})\mathbf{A}'$$
 (3.2.12.30)

$$\implies \mathbf{A} = \mathbf{A}' \qquad (3.2.12.31)$$

So, \mathbb{T} is invertible. Conclusion: From case 1,case 2 \mathbb{T} is invertible if and only if p = mand **B** is an invertible $m \times m$ matrix. Example: Let p = m = 3, n = 4 Let $\mathbb{T} : \mathbf{V} \to \mathbf{W}$ adds row 2 to row 3 for a matrix $A \in V$

The elementary matrix that performs this is

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \tag{3.2.12.32}$$

Let
$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 4 & 9 & 2 & 6 \end{pmatrix}$$
 (3.2.12.33)

$$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A} \qquad (3.2.12.34)$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 4 & 9 & 2 & 6 \end{pmatrix}$$
 (3.2.12.35)

$$= \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 5 & 12 & 8 & 13 \end{pmatrix}$$
 (3.2.12.36)

$$= \mathbf{C} \in \mathbf{W}$$
 (3.2.12.37)

Let transformation $\mathbb{T}_1 : \mathbf{W} \to \mathbf{V}$ subtracts row2 from row 3 for a matrix $C \in W$ and is performed by elementary matrix

$$\mathbf{U} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix}$$

$$(3.2.12.38)$$

$$\mathbf{Let C} = \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 5 & 12 & 8 & 13 \end{pmatrix}$$

$$(3.2.12.39)$$

$$\mathbb{T}_{1}(\mathbf{C}) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 5 & 12 & 8 & 13 \end{pmatrix}$$

$$(3.2.12.40)$$

$$= \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 4 & 9 & 2 & 6 \end{pmatrix}$$

$$(3.2.12.41)$$

$$= \mathbf{A}$$

$$(3.2.12.42)$$

$$\implies \mathbb{T}_{1}(\mathbf{C}) = \mathbf{A}$$

$$(3.2.12.43)$$

$$\mathbb{T}_{1}(\mathbb{T}(\mathbf{A})) = \mathbf{A}$$

$$(3.2.12.44)$$
and
$$\mathbb{T}(\mathbf{A}) = \mathbf{C}$$

$$(3.2.12.45)$$

$$\implies \mathbb{T}(\mathbb{T}_{1}(\mathbf{C})) = \mathbf{C}$$

$$(3.2.12.46)$$

So, \mathbb{T}_1 is the inverse transformation of \mathbb{T} and

$$\mathbb{T}_1 = \mathbb{T}^{-1} \qquad (3.2.12.47)$$

$$\mathbf{UB} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix}$$
 (3.2.12.48)

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \qquad (3.2.12.49)$$

$$\mathbf{BU} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix}$$
 (3.2.12.50)

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
 (3.2.12.51)

$$\implies \mathbf{B}^{-1} = \mathbf{U} \qquad (3.2.12.52)$$

So, \mathbb{T} is invertible and \mathbf{B} is an invertible 3×3 matrix.

3.3 Isomorphism

3.3.1. Let V be the set of complex numbers and let F be the field of real numbers. With the usual operations, V is a vector space over F. Describe explicitly an isomorphism of this space onto \mathbb{R}^2 .

Solution: Let,

$$\mathbf{T}: \mathbf{V} \to \mathbb{R}^2 \tag{3.3.1.1}$$

$$\mathbf{T}(x+iy) = \begin{pmatrix} x \\ y \end{pmatrix} \tag{3.3.1.2}$$

$$x, y \in \mathbb{R} \quad i \in \mathbb{C}$$
 (3.3.1.3)

Consider two vectors,

$$\mathbf{u}, \mathbf{v} \in \mathbf{V} \tag{3.3.1.4}$$

Using (3.3.1.2) if **T** is Linear Transformation.

$$\mathbf{T}(\mathbf{u} + c\mathbf{v}) = \mathbf{T}(\mathbf{u}) + \mathbf{T}(c\mathbf{v}) = \mathbf{T}(\mathbf{u}) + c\mathbf{T}(\mathbf{v})$$
(3.3.1.5)

Hence this is a Linear transformation. Now, checking if **T** is one-one. let,

$$\mathbf{u} = 0 = 0 + j0 \tag{3.3.1.6}$$

(3.3.1.7)

from (3.3.1.2),

$$\mathbf{T}(\mathbf{u}) = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{3.3.1.8}$$

From (3.3.1.8) we see **T** is one-one. Now, checking if **T** is onto. let,

$$\mathbf{T}(\mathbf{u}) = \begin{pmatrix} a \\ b \end{pmatrix} \tag{3.3.1.9}$$

where, a, b are scalars. Using (3.3.1.2), we see there exists a solution for (3.3.1.8) for all a, b. Hence **T** is onto. Therefore **T** is isomorphic over **V** and \mathbb{R}^2

3.3.2. Let V be a vector space over the field of complex numbers, and suppose there is an isomorphism T of V onto C^3 . Let α_1 , α_2 ,

 α_3 , α_4 be vectors in V such that

$$T(\alpha_1) = \begin{pmatrix} 1\\0\\i \end{pmatrix}, T(\alpha_2) = \begin{pmatrix} -2\\1+i\\0 \end{pmatrix},$$
$$T(\alpha_3) = \begin{pmatrix} -1\\1\\1 \end{pmatrix}, T(\alpha_4) = \begin{pmatrix} \sqrt{2}\\i\\3 \end{pmatrix}$$
(3.3.2.1)

- a) Is α₁ in the subspace spanned by α₂ and α₃?
 Solution: T: V → W is an isomorphism if
 (1) T is one one.
 - (2) T is onto.

$$\begin{pmatrix} 1 & -2 & -1 & \sqrt{2} \\ 0 & 1+i & 1 & i \\ i & 0 & 1 & 3 \end{pmatrix} \xrightarrow{ref} \begin{pmatrix} 1 & 0 & -i & -3i \\ 0 & 2 & 1-i & i+1 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

(3.3.2.2)

T is one one over C^3 if

$$T(\alpha) = 0 \implies \alpha = 0 \tag{3.3.2.3}$$

now,

$$\begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ i & 0 & 3 \end{pmatrix} \alpha = 0$$
 (3.3.2.4)

consider the row reduced matrix

$$\begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ i & 0 & 3 \end{pmatrix} \xrightarrow{R_3 \to R_3 - iR_1} \begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ 0 & -2 & \sqrt{2} + 3i \end{pmatrix}$$
(3.3.2.5)

$$\stackrel{R_2 \leftarrow (1-i)R_2}{\underset{R_3 \leftarrow R_3 + R_2}{\longleftarrow}} \begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 2 & i+1 \\ 0 & 0 & \sqrt{2} + 4i + 1 \end{pmatrix} \\
(3.3.2.6)$$

$$\alpha = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \tag{3.3.2.7}$$

Therefore it holds the condition of one one and the rank = no. of pivot columns = 3 (equal to no. of columns). Thus the vectors are linearly independent hence it is onto . Since T is an isomorphoism onto C^3 .

$$T(\alpha_1) = c_1 T(\alpha_2) + c_2 T(\alpha_3)$$
 (3.3.2.8)

 c_1 and c_2 are scalar.

$$\begin{pmatrix} 1 \\ 0 \\ i \end{pmatrix} = c_1 \begin{pmatrix} -2 \\ 1+i \\ 0 \end{pmatrix} + c_2 \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$
 (3.3.2.9)

$$\begin{pmatrix} 1 \\ 0 \\ i \end{pmatrix} = \begin{pmatrix} -2 & -1 \\ 1+i & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}$$
 (3.3.2.10)

Now we find c_i by row reducing augmented matrix.

$$\begin{pmatrix}
-2 & -1 & 1 \\
1+i & 1 & 0 \\
0 & 1 & i
\end{pmatrix}
\xrightarrow{R_1 \to -R_1/2}
\begin{pmatrix}
1 & \frac{1}{2} & -\frac{1}{2} \\
0 & 1 & i \\
1+i & 1 & 0
\end{pmatrix}$$

$$(3.3.2.11)$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_2/2}
\xrightarrow{R_3 \leftarrow R_3 - (1+i)R_1}
\begin{pmatrix}
1 & 0 & \frac{-1-i}{2} \\
0 & 1 & i \\
0 & \frac{1-i}{2} & \frac{1+i}{2}
\end{pmatrix}$$

$$(3.3.2.12)$$

$$\xrightarrow{R_3 \leftarrow R_3 - (1-i)/2R_2}
\begin{pmatrix}
1 & 0 & \frac{-1-i}{2} \\
0 & 1 & i \\
0 & 0 & 0
\end{pmatrix}$$

$$(3.3.2.13)$$

Therefore the coordinate matrix of the vector is

$$\begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = \begin{pmatrix} \frac{-1-i}{2} \\ i \end{pmatrix}$$
 (3.3.2.14)

substituting the c_i in (3.3.2.8)

$$T(\alpha_1) = -\frac{1+i}{2}T(\alpha_2) + iT(\alpha_3)$$
 (3.3.2.15)

Hence α_1 belongs to the subspace spanned by α_2 and α_3 .

b) Let V be a vector space over the field of complex numbers, and suppose there is an isomorphism T of V onto C^3 . Let α_1 , α_2 , α_3 , α_4 be vectors in V such that

$$T(\alpha_1) = \begin{pmatrix} 1\\0\\i \end{pmatrix}, T(\alpha_2) = \begin{pmatrix} -2\\1+i\\0 \end{pmatrix},$$
$$T(\alpha_3) = \begin{pmatrix} -1\\1\\1 \end{pmatrix}, T(\alpha_4) = \begin{pmatrix} \sqrt{2}\\i\\3 \end{pmatrix}$$
(3.3.2.16)

Let W_1 be the subspace spanned by α_1 and α_2 , and let W_2 be the subspace spanned by α_3 and α_4 . What is the intersection of W_1

and W_2 .

Solution: $T: V \rightarrow W$ is an isomorphism if (1) T is one one.

(2) T is onto. If W_1 and W_2 are finite-dimensional subspaces of a vector space V, then $W_1 + W_2$ is finite-dimensional and

$$dim(W_1)+dim(W_2)=dim(W_1\cap W_2)+dim(W_1+W_2)$$

(3.3.2.17)

(3.3.2.19)

We have,

$$T = \begin{pmatrix} 1 & -2 & -1 & \sqrt{2} \\ 0 & 1+i & 1 & i \\ i & 0 & 1 & 3 \end{pmatrix}$$

$$(3.3.2.18)$$

$$\begin{pmatrix} 1 & -2 & -1 & \sqrt{2} \\ 0 & 1+i & 1 & i \\ i & 0 & 1 & 3 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 1 & 0 & -i & 0 \\ 0 & 1 & \frac{1-i}{2} & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

T is one one over C^3 if

$$T(\alpha) = 0 \implies \alpha = 0 \tag{3.3.2.20}$$

now,

$$\begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ i & 0 & 3 \end{pmatrix} \alpha = 0$$
 (3.3.2.21)

consider the row reduced matrix

$$\begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ i & 0 & 3 \end{pmatrix} \xrightarrow{R_3 \to R_3 - iR_1} \begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ 0 & -2 & \sqrt{2} + 3i \end{pmatrix}$$

$$(3.3.2.22)$$

$$\xrightarrow{R_2 \leftarrow (1-i)R_2}
\xrightarrow{R_3 \leftarrow R_3 + R_2}
\begin{pmatrix}
1 & -2 & \sqrt{2} \\
0 & 2 & i+1 \\
0 & 0 & \sqrt{2} + 4i + 1
\end{pmatrix}$$
(3.3.2.23)

$$\alpha = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \tag{3.3.2.24}$$

Therefore it holds the condition of one one and the rank = no. of pivot columns = 3 (equal to no. of columns). Thus the vectors are linearly independent hence it is onto.

Hence, T is an isomorphism onto C^3 . Also, from (3.3.2.19), we observe,

$$T(\alpha_3) = -iT(\alpha_1) + \frac{i-1}{2}T(\alpha_2)$$
 (3.3.2.25)

Hence $T(\alpha_3)$ belongs to the subspace spanned by $T(\alpha_1)$ and $T(\alpha_2)$.

Therefore, α_3 is in subspace spanned by α_1 and α_2 . Therefore,

$$\alpha_3 \in W_1 \qquad (3.3.2.26)$$

$$\implies \alpha_3 \in W_1 \cap W_2 \qquad (3.3.2.27)$$

Since $T(\alpha_1)$ and $T(\alpha_2)$ are linearly independent, and $T(\alpha_3)$ and $T(\alpha_4)$ are linearly independent, we have,

$$dim(W_1) = dim(W_2) = 2$$
 (3.3.2.28)

From (3.3.2.17),

$$dim(W_1) + dim(W_2) = dim(W_1 \cap W_2) + dim(W_1 + W_2)$$

$$(3.3.2.29)$$

$$dim(W_1 + W_2) = 3$$

$$(3.3.2.30)$$

$$\implies dim(W_1 \cap W_2) = 2 + 2 - 3$$

$$(3.3.2.31)$$

$$\implies dim(W_1 \cap W_2) = 1$$

$$(3.3.2.32)$$

Therefore,

$$W_1 \cap W_2 = c\alpha_3 \tag{3.3.2.33}$$

c) Find a basis for the subspace of V spanned by the 4 vectors α_i .

isomorphic to C^3 via isomorphism T which implies that C^3 is also isomorphic to V via isomorphism T^{-1} .

As V is isomorphic to C^3 , so

$$dim(V) = dim(C^3) = 3$$
 (3.3.2.34)

Now,

$$\begin{pmatrix}
1 & 0 & i \\
-2 & 1+i & 0 \\
-1 & 1 & 1 \\
\sqrt{2} & i & 3
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 + R_1}
\begin{pmatrix}
1 & 0 & i \\
-2 & 1+i & 0 \\
0 & 1 & 1+i \\
2 & i\sqrt{2} & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_4 \leftarrow R_4 + R_2}
\begin{pmatrix}
1 & 0 & i \\
-2 & 1+i & 0 \\
0 & 1 & 1+i \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_2 \leftarrow R_2 + 2R_1}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow (1+i)R_3}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 1+i & 2i \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_3 - R_2}
\begin{pmatrix}
1 & 0 & i \\
0 & 1+i & 2i \\
0 & 0 & 0 \\
0 & 1+i(1+\sqrt{2}) & 3\sqrt{2}
\end{pmatrix}$$

From here we can get that $T\alpha_3$ is dependent vector while $T\alpha_1$, $T\alpha_2$ and $T\alpha_4$ are independent vector. These $T\alpha_1$, $T\alpha_2$ and $T\alpha_4$ also span the vector space C^3 , so these 3 vectors are the basis of C^3 .

As dim(V) = 3, so it must have 3 basis and as V and C^3 are isomorphic so α_1 , α_2 and α_4 are the basis of V.

Solution: V is a vector space and V is 3.3.3. Let \mathbb{W} be the set of all 2×2 complex Hermitian matrices, that is the sset of 2×2 complex matrices A ssuch that $A_{ij} = \overline{A_{ji}}$ (the bar denoting complex conjugation). W is a vector space over the field of real numbers, under the usual operations. Verify that

$$\begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} \rightarrow \begin{pmatrix} t + x & y + iz \\ y - iz & t - x \end{pmatrix}$$
 (3.3.3.1)

is an isomorphism of \mathbb{R}^4 onto \mathbb{W} .

Solution:

a) Check for linearity: The transformation T

is given by

$$T: \mathbb{R}^4 \to \mathbb{W} \tag{3.3.3.2}$$

$$T \begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} = \begin{pmatrix} t + x & y + iz \\ y - iz & t - x \end{pmatrix}$$
 (3.3.3.3)

Let $\mathbf{x} = \begin{bmatrix} x \\ y \\ t \end{bmatrix}$. Expressing R.H.S of equation

(3.3.3.3) using Kronecker Product,

$$T(\mathbf{x}) = \begin{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 1 \end{pmatrix} \mathbf{x} & \begin{pmatrix} 0 & 1 & i & 0 \end{pmatrix} \mathbf{x} \\ \begin{pmatrix} 0 & 1 & -i & 0 \end{pmatrix} \mathbf{x} & \begin{pmatrix} -1 & 0 & 0 & 1 \end{pmatrix} \mathbf{x} \end{pmatrix}$$

$$= \left(\begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & -i & 0 \end{pmatrix}\right) \mathbf{x} \quad \begin{pmatrix} 0 & 1 & i & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix} \mathbf{x}$$

$$= \left(\begin{pmatrix} 1 & 0 & 0 & 1 & 0 & 1 & i & 0 \\ 0 & 1 & -i & 0 & -1 & 0 & 0 & 1 \end{pmatrix}\right) \begin{pmatrix} x & 0 \\ y & 0 \\ z & 0 \\ t & 0 \\ 0 & x \\ 0 & y \\ 0 & z \\ 0 & t \end{pmatrix}$$

$$= \left(\begin{pmatrix} 1 & 0 & 0 & 1 & 0 & 1 & i & 0 \\ 0 & 1 & -i & 0 & -1 & 0 & 0 & 1 \end{pmatrix}\right) \begin{pmatrix} x & 0 \\ y & 0 \\ z & 0 \\ t & 0 \\ 0 & x \\ 0 & y \\ 0 & z \\ 0 & t \end{pmatrix}$$

$$= \left(3.3.3.6\right)$$

$$\implies T(\mathbf{x}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{x} & \mathbf{0}_{4\times 1} \\ \mathbf{0}_{4\times 1} & \mathbf{x} \end{pmatrix} (3.3.3.7)$$

Where **A** and **B** are block matrices.

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

$$\mathbf{B} = \begin{pmatrix} 0 & 1 & i & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix}$$
(3.3.3.8) one-one and that implies $T : \mathbb{R}^4$ isomorphism.

3.3.4. Show that $\mathbf{F}^{\mathbf{m} \times \mathbf{n}}$ is isomorphic to $\mathbf{F}^{\mathbf{m} \cdot \mathbf{n}}$.

Solution: See Tables 3.3.4.1, 3.3.4.

$$\mathbf{B} = \begin{pmatrix} 0 & 1 & i & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix} \tag{3.3.3.9}$$

The Kronecker Product of I_2 and x gives the block matrix in equation (3.3.3.7).

$$\mathbf{I}_{2\times2}\otimes\mathbf{x}_{4\times1} = \begin{pmatrix} \mathbf{x} & \mathbf{0} \\ \mathbf{0} & \mathbf{x} \end{pmatrix}_{8\times2} \tag{3.3.3.10}$$

Hence we can write equation (3.3.3.7) as,

$$T(\mathbf{x}) = (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{x}) \tag{3.3.3.11}$$

Let $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{R}^4$ and $\alpha, \beta \in \mathbb{R}$.

$$T (\alpha \mathbf{x}_1 + \beta \mathbf{x}_2) = (\mathbf{A} \quad \mathbf{B}) (\mathbf{I} \otimes (\alpha \mathbf{x}_1 + \beta \mathbf{x}_2))$$

$$= \alpha (\mathbf{A} \quad \mathbf{B}) (\mathbf{I} \otimes \mathbf{x}_1) + \beta (\mathbf{A} \quad \mathbf{B}) (\mathbf{I} \otimes \mathbf{x}_2)$$

$$= \alpha T \mathbf{x}_1 + \beta T \mathbf{x}_2$$

$$(3.3.3.14)$$

Therefore from equation (3.3.3.14), we can say T is linear transformation.

b) Check for one-one property: For transformation T to be one-one, we can prove if $T(\mathbf{x}) = \mathbf{0}$, that implies $\mathbf{x} = \mathbf{0}$. From the equation (3.3.3.11),

$$T\left(\mathbf{x}\right) = \mathbf{0} \tag{3.3.3.15}$$

$$(\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{x}) = \mathbf{0} \tag{3.3.3.16}$$

$$\implies \begin{pmatrix} 1 & 0 & 0 & 1 & 0 & 1 & i & 0 \\ 0 & 1 & -i & 0 & -1 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x & 0 \\ y & 0 \\ z & 0 \\ t & 0 \\ 0 & x \\ 0 & y \\ 0 & z \\ 0 & t \end{pmatrix} = \mathbf{0}_{2 \times 2}$$

$$(3.3.3.17)$$

From equation (3.3.3.3),

$$\begin{pmatrix} t + x & y + iz \\ y - iz & t - x \end{pmatrix} = \mathbf{0}_{2 \times 2}$$
 (3.3.3.18)

$$\implies x = 0, y = 0, z = 0, t = 0$$
 (3.3.3.19)

 \implies **x** = **0** (3.3.3.20)

Hence from (3.3.3.15) and (3.3.3.20), T is one-one and that implies $T: \mathbb{R}^4 \to \mathbb{W}$ is

Solution: See Tables 3.3.4.1, 3.3.4.2

 $\mathbb{R}^{2\times 2}$ is isomorphic to \mathbb{R}^4 ie, $\mathbb{R}^{2\times 2} \cong \mathbb{R}^4$.

3.3.5. Let V be the set of complex numbers regarded as a vector space over the field of real numbers. We define a function T from V into the space of 2×2 real matrices, as follows. If z = x + iy

Invertible Linear Map	A linear map $T \in L(V, W)$ is called invertible if there exists a linear map $S \in L(W, V)$ such that ST equals the identity map on V and TS equals the identity map on W . A linear map $S \in L(W, V)$ satisfying $ST = I_V$ and $TS = I_W$ is called an inverse of T .
Isomorphic Vector Spaces	Two vector spaces V and W are called isomorphic if there is an isomorphism from one vector space onto the other one. An isomorphism is an invertible linear map.
Rank Nullity Theorem	Let V and W be finite dimensional vector spaces. Let $T\colon V\to W$ be a linear transformation $\text{Rank}(T) + \text{Nullity}(T) = \text{dim } V$

TABLE 3.3.4.1: Definition

Result 1	The space of all $m \times n$ matrices over the field F has dimension mn .
Result 2	Let V and W be finite-dimensional vector spaces over the field F such that dim V = dim W. If T is a linear transformation from V into W, then the following are equivalent: (a). T is invertible. (b). T is non-singular. (c). T is onto, that is, range of T is W.

TABLE 3.3.4.2: Results Used

with x and y real numbers, then

$$\mathbf{T}(z) = \begin{pmatrix} x + 7y & 5y \\ -10y & x - 7y \end{pmatrix}$$

a) Verify that T is a one-one (real) linear transformation of V into the space of 2×2 real matrices.

Solution: The kronecker product also called as matrix direct product is defined as

$$\mathbf{A} \otimes \mathbf{B} = \begin{pmatrix} a_{11}\mathbf{B} & \cdots & a_{1n}\mathbf{B} \\ \vdots & \ddots & \vdots \\ a_{m1}\mathbf{B} & \cdots & a_{mn}\mathbf{B} \end{pmatrix}$$
(3.3.5.1)

Also,

$$\mathbf{A} \otimes (\mathbf{B} + \mathbf{C}) = \mathbf{A} \otimes \mathbf{B} + \mathbf{A} \otimes \mathbf{C}$$
 (3.3.5.2)

$$\mathbf{A} \otimes (k\mathbf{B}) = k(\mathbf{A} \otimes \mathbf{B}) \tag{3.3.5.3}$$

Given,

$$\mathbf{T}: \mathbf{C} \to \mathbf{R}^{2\times 2}$$

$$\mathbf{T}(x+iy) = \begin{pmatrix} x+7y & 5y \\ -10y & x-7y \end{pmatrix}$$
 (3.3.5.4)

Let,

$$z = x + iy;$$
 $w = a + ib;$ $z, w \in \mathbb{C}$

Also the RHS of (3.3.5.4) can be expressed as,

$$\mathbf{T}(\mathbf{z}) = \begin{pmatrix} 1 & 7 \\ 0 & -10 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} & \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \end{pmatrix}$$
$$= \begin{pmatrix} 1 & 7 & 0 & 5 \\ 0 & -10 & 1 & -7 \end{pmatrix} \begin{pmatrix} x & 0 \\ y & 0 \\ 0 & x \\ 0 & y \end{pmatrix}$$
$$= \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{x} & 0 \\ 0 & \mathbf{x} \end{pmatrix} \tag{3.3.5.5}$$

where **A** and **B** are block matrices and,

$$\mathbf{x} = \begin{pmatrix} x \\ y \end{pmatrix}$$

The diagonal block matrix can be expressed as the kronecker product of ${\bf I}$ and ${\bf x}$

$$\mathbf{I} \otimes \mathbf{x} = \begin{pmatrix} \mathbf{x} & 0 \\ 0 & \mathbf{x} \end{pmatrix} \tag{3.3.5.6}$$

Defining Sets	We define set S and set T as $S = \{(a,b): a,b \in \mathbb{N}, 1 \le a \le m, 1 \le b \le n\}, T = \{1,2,,mn\}$	
Defining Bijection	We now define a bijection $\sigma: S \to T$ as $(a,b) \to (a-1)\mathbf{n} + \mathbf{b}$	
Defining Function G	We now define a function G from $F^{m\times n}$ to F^{mn} as follows. Let $\mathbf{A} \in F^{m\times n}$. Then map \mathbf{A} to the mn tupple that has \mathbf{A}_{ij} in the $\sigma(i,j)$ position. In other words, $\mathbf{A} \to (\mathbf{A}_{11},\mathbf{A}_{12},,\mathbf{A}_{1n},,\mathbf{A}_{m1},\mathbf{A}_{m2},,\mathbf{A}_{mn})$	
Proving <i>G</i> to be Linear	Since, addition in $F^{m\times n}$ and in F^{mn} is performed component-wise, $G(\mathbf{A} + \mathbf{B}) = G(\mathbf{A}) + G(\mathbf{B})$ and scalar multiplication in $F^{m\times n}$ and in F^{mn} is also defined as $G(c\mathbf{A}) = cG(\mathbf{A})$.	
Proving <i>G</i> to be One-One	$G(\mathbf{A}) = G(\mathbf{B})$ $\implies (\mathbf{A}_{11}, \mathbf{A}_{12},, \mathbf{A}_{1n},, \mathbf{A}_{m1}, \mathbf{A}_{m2},, \mathbf{A}_{mn}) = (\mathbf{B}_{11}, \mathbf{B}_{12},, \mathbf{B}_{1n},, \mathbf{B}_{m1}, \mathbf{B}_{m2},, \mathbf{B}_{mn})$ $\implies \mathbf{A}_{i,j} = \mathbf{B}_{ij} \forall 1 \le i \le m, 1 \le j \le n$ $\implies \mathbf{A} = \mathbf{B}$	
Proving G to be Onto	Since G is one to one, so $\text{Null}(G) = 0$. Thus, by Rank-Nullity Theorem $\dim(\text{Range}(G)) = mn$, proving G to be a surjective (onto) map as by Result 1 dimension of $F^{m \times n} = mn$	
$F^{m \times n} \cong F^{mn}$	Since G has an inverse and is an isomorphism of \mathbf{T} . Thus, by Result 2 $F^{m \times n} \cong F^{mn}$	

TABLE 3.3.4.3: Proof

Where I is an identity matrix. (3.3.5.5) can be rewritten as,

$$\mathbf{T}(\mathbf{z}) = (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{x}) \tag{3.3.5.7}$$

Consider.

$$\mathbf{T}(\alpha \mathbf{z} + \mathbf{w}) = (\mathbf{A} \ \mathbf{B})(\mathbf{I} \otimes (\alpha \mathbf{z} + \mathbf{w}))$$

Using properties (3.3.5.2), (3.3.5.3), the above equation can be expressed as,

$$\mathbf{T}(\alpha \mathbf{z} + \mathbf{w}) = (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes (\alpha \mathbf{z})) + (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{x})$$
$$= \alpha (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{z}) + (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{w})$$
$$= \alpha \mathbf{T}(\mathbf{z}) + \mathbf{T}(\mathbf{w}) \qquad (3.3.5.8)$$

From (3.3.5.8), it can be proved that **T** is a linear operator.

b) Verify that

$$\mathbf{T}(z_1 z_2) = \mathbf{T}(z_1) \mathbf{T}(z_2)$$
 (3.3.5.9)

Solution: The product of two Kronecker products yields another Kronecker product:

$$(\mathbf{A} \otimes \mathbf{B})(\mathbf{C} \otimes \mathbf{D}) = (\mathbf{AC}) \otimes (\mathbf{BD})$$
 (3.3.5.10)

Given,

$$\mathbf{T}(\mathbf{z}) = \begin{pmatrix} x + 7y & 5y \\ -10y & x - 7y \end{pmatrix} (3.3.5.11)$$
$$\mathbf{x} = \begin{pmatrix} x \\ y \end{pmatrix} (3.3.5.12)$$

$$\mathbf{T}(\mathbf{x}) = \begin{pmatrix} \begin{pmatrix} 1 & 7 \end{pmatrix} \mathbf{x} & \begin{pmatrix} 0 & 5 \end{pmatrix} \mathbf{x} \\ \begin{pmatrix} 0 & -10 \end{pmatrix} \mathbf{x} & \begin{pmatrix} 1 & -7 \end{pmatrix} \mathbf{x} \end{pmatrix} (3.3.5.13)$$

$$= \begin{pmatrix} 1 & 7 \\ 0 & -10 \end{pmatrix} \mathbf{x} \quad \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \mathbf{x}$$
 (3.3.5.14)

Let
$$\mathbf{A} = \begin{pmatrix} 1 & 7 \\ 0 & -10 \end{pmatrix}$$
 (3.3.5.15)

$$\mathbf{B} = \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \tag{3.3.5.16}$$

$$\implies \mathbf{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A}\mathbf{x} & \mathbf{B}\mathbf{x} \end{pmatrix} \qquad (3.3.5.17)$$

$$\mathbf{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{x} & \mathbf{0} \\ \mathbf{0} & \mathbf{x} \end{pmatrix} \qquad (3.3.5.18)$$

The diagonal block matrix can be expressed as the kronecker product of \mathbf{I} and \mathbf{x}

$$\mathbf{I} \otimes \mathbf{x} = \begin{pmatrix} \mathbf{x} & \mathbf{0} \\ \mathbf{0} & \mathbf{x} \end{pmatrix} \tag{3.3.5.19}$$

We can write (3.3.5.11) as

$$\mathbf{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes \mathbf{x}) \tag{3.3.5.20}$$

Starting with RHS of (3.3.5.10)

$$\mathbf{T}(z_1)\mathbf{T}(z_2) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes \mathbf{z_1}) \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes \mathbf{z_2})$$
(3.3.5.21)

If

$$(\mathbf{I} \otimes \mathbf{z}_1) (\mathbf{A} \quad \mathbf{B}) (\mathbf{I} \otimes \mathbf{z}_2) = (\mathbf{I} \otimes \mathbf{z}_1 \mathbf{z}_2)$$

$$(3.3.5.22)$$

then, we can write (3.3.5.21) as

$$\mathbf{T}(z_1)\mathbf{T}(z_2) = (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{z_1}\mathbf{z_2}) = \mathbf{T}(z_1z_2)$$
(3.3.5.23)

c) How would you describe the range of T? **Solution:**

$$\mathbb{T}: \mathbf{V} \to \mathbb{R}^{2 \times 2} \tag{3.3.5.24}$$

where $\mathbb{R}^{2\times 2}$, is the space of all 2×2 real matrices

$$\mathbb{T}(z) = \begin{pmatrix} x + 7y & 5y \\ -10y & x - 7y \end{pmatrix} \quad (3.3.5.25)$$

$$\mathbf{x} = \begin{pmatrix} x \\ y \end{pmatrix} \quad (3.3.5.26)$$

$$\mathbb{T}(\mathbf{x}) = \begin{pmatrix} (1 & 7)\mathbf{x} & (0 & 5)\mathbf{x} \\ (0 & -10)\mathbf{x} & (1 & -7)\mathbf{x} \end{pmatrix} \quad (3.3.5.27)$$

$$= \begin{pmatrix} (1 & 7 \\ 0 & -10 \end{pmatrix} \mathbf{x} & \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \mathbf{x} \end{pmatrix} \quad (3.3.5.28)$$

$$\text{Let } \mathbf{A} = \begin{pmatrix} 1 & 7 \\ 0 & -10 \end{pmatrix} \quad (3.3.5.29)$$

$$\mathbf{B} = \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \quad (3.3.5.30)$$

$$\implies \mathbb{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A}\mathbf{x} & \mathbf{B}\mathbf{x} \end{pmatrix} \quad (3.3.5.31)$$

$$\mathbb{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{x} & \mathbf{0}_{2\times 1} \\ \mathbf{0}_{2\times 1} & \mathbf{x} \end{pmatrix} \quad (3.3.5.32)$$

The kronecker product of \mathbf{I} , \mathbf{x} gives the block matrix

$$\mathbf{I}_{2\times 2} \otimes \mathbf{x}_{2\times 1} = \begin{pmatrix} \mathbf{x} & \mathbf{0} \\ \mathbf{0} & \mathbf{x} \end{pmatrix}_{4\times 2}$$

$$(3.3.5.33)$$

$$(3.3.5.32) \implies \mathbb{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \mathbf{x}$$

$$(3.3.5.34)$$

$$= \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{bmatrix} \begin{pmatrix} x \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ y \end{pmatrix} \end{bmatrix}$$

$$(3.3.5.35)$$

$$= x \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} + y \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$(3.3.5.36)$$

$$\mathbf{I} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix}$$

$$(3.3.5.37)$$

$$\mathbf{I} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{pmatrix}$$

$$(3.3.5.38)$$

Kronecker product in the first term of (3.3.5.36) picks out 1st columns of **A**, **B** and in the second term picks out 2nd columns of

A, B so basis for range(\mathbb{T}) is

$$\left\{ \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\} = (3.3.5.39)$$

$$\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 7 & 5 \\ -10 & -7 \end{pmatrix} \right\}$$

$$(3.3.5.40)$$

$$\operatorname{range}(\mathbb{T}) = (3.3.5.41)$$

$$\operatorname{span of} \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 7 & 5 \\ -10 & -7 \end{pmatrix} \right\}$$

$$(3.3.5.42)$$

3.3.6. Let **V** and **W** be finite-dimensional vector spaces over the field **F**. Prove that **V** and **W** are isomorphic if and only if dim **V** = dim **W Solution:** Refer Table 3.3.6.1.

Therefore from equation (3.3.6.4) and (3.3.6.7) in Table 3.3.6.1, $\{T(\mathbf{v_1}), T(\mathbf{v_2}), \cdots, T(\mathbf{v_n})\}$ are linearly independent and span **W**.

$$\implies$$
 dim $\mathbf{W} = \mathbf{n} = \text{dim } \mathbf{V}$

V and W are isomorphic if and only if dim V = dim W

3.3.7. Let V and W be finite-dimensional vector spaces over the field F and let U be an isomorphism of V and W. Prove that $T \to UTU^{-1}$ is an isomorphism of L(V, V) onto L(W, W).

Solution:

Example Let

$$U = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \tag{3.3.7.1}$$

here U is an isomorphism from $\mathbb{R}^{2\times 2}$ to $\mathbb{R}^{2\times 2}$ since inverse of U exists and

$$U^{-1} = \begin{pmatrix} -2 & -\frac{3}{2} \\ -1 & -\frac{1}{2} \end{pmatrix}$$
 (3.3.7.2)

Consider

$$T = \begin{pmatrix} -1 & 2 \\ 3 & 1 \end{pmatrix} \in \mathbb{R}^{2 \times 2} \tag{3.3.7.3}$$

Now

$$UTU^{-1} = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} -1 & 2 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} -2 & -1 \\ -\frac{3}{2} & -\frac{1}{2} \end{pmatrix}$$
 (3.3.7.4)
= $\begin{pmatrix} -16 & -7 \\ -33 & -14 \end{pmatrix} \in \mathbb{R}^{2 \times 2}$ (3.3.7.5)

Also inverse exists for T

$$S = T^{-1} = \begin{pmatrix} -\frac{1}{2} & \frac{2}{7} \\ \frac{3}{7} & \frac{1}{7} \end{pmatrix}$$
 (3.3.7.6)

Since T inverse exists $\mathcal{T}(T) = UTU^{-1}$ is an isomorphism from $\mathbb{R}^{2\times 2}$ onto $\mathbb{R}^{2\times 2}$.

3.3.8. Let T be a linear operator on the finite-dimensional space \mathbb{V} . Suppose there is a linear operator U on \mathbb{V} such that TU = I. Prove that T is invertible and $U = T^{-1}$. Give an example which shows that this is false when \mathbb{V} is not finite-dimensional.

Solution: Let $T: \mathbb{V} \to \mathbb{V}$ be a linear operator, where \mathbb{V} is a finite dimensional vectors space and $U: \mathbb{V} \to \mathbb{V}$ is also a linear operator such that,

$$TU = I \tag{3.3.8.1}$$

Where, I is an identity transformation. Now we know that linear transformations are functions. Hence,

$$TU = I$$
 is a function (3.3.8.2)

$$\Longrightarrow I: \mathbb{V} \to \mathbb{V} \tag{3.3.8.3}$$

Such that T(V) = V. Defining $TU : \mathbb{V} \to \mathbb{V}$ to be a linear operator, we have,

$$T[U(V_i)] = V_i \qquad [V_i \in \mathbb{V}] \qquad (3.3.8.4)$$

Now we show in the below Table that T is one-one and onto as follows,

Hence we get from Table 3.3.8.1 that, *T* is invertible. Hence we get the following,

$$TT^{-1} = I$$
 (3.3.8.5)

Where T^{-1} is an inverse function of linear operator T. Hence,

$$TT^{-1} = I = TU$$
 (3.3.8.6)

$$\implies T^{-1}(TT^{-1}) = T^{-1}(TU)$$
 (3.3.8.7)

$$\implies T^{-1}(I) = IU \tag{3.3.8.8}$$

$$\implies T^{-1} = U \tag{3.3.8.9}$$

Hence from (3.3.8.9) it is proven that T is invertible and $T^{-1} = U$

Example: Let D be the differential operator $D: \mathbb{V} \to \mathbb{V}$ where \mathbb{V} is a space of polynomial

$T: \mathbf{V} \to \mathbf{W} \mathbf{V} = \{\mathbf{v}_1, \mathbf{v}_2, \cdots, \mathbf{v}_n\} \mathbf{W} = \{T(\mathbf{v}_1), T(\mathbf{v}_2), \cdots, T(\mathbf{v}_n)\}$			
Property Used	Derivation		
T is one-one	Linear combination of vectors in W		
	$\sum_{k=1}^{n} \alpha_k T(\mathbf{v_k}) = 0 $ (3.3.6.1)		
	$\sum_{k=1}^{n} T(\alpha_k \mathbf{v_k}) = 0 $ (3.3.6.2)		
	$\implies \sum_{k=1}^{n} \alpha_k \mathbf{v_k} = 0 \tag{3.3.6.3}$		
	$\implies \alpha_1 = \alpha_2 = \dots = \alpha_n = 0 \tag{3.3.6.4}$		
	From equation (3.3.6.1) and (3.3.6.4), the set of vectors $\{T(\mathbf{v_1})\}$ are linearly independent), $T(\mathbf{v_2}), \cdots, T(\mathbf{v_n})$	
T is onto	For any $y \in W$ there exists an $x \in V$ such that $T(x) = y$.		
	$\mathbf{x} = \sum_{k=1}^{n} \alpha_k \mathbf{v_k} \tag{3.3.6.5}$		
	$T(\mathbf{x}) = T(\sum_{k=1}^{n} \alpha_k \mathbf{v_k}) = \mathbf{y} $ (3.3.6.6)		
	$\sum_{k=1}^{n} \alpha_k T(\mathbf{v_k}) = \mathbf{y} $ (3.3.6.7)		
	From equation (3.3.6.7), any vector in W can be represented as linear combination of $\{T(\mathbf{v_1}), T(\mathbf{v_2}), \cdots, T(\mathbf{v_n})\}$. That is it spans W .		

TABLE 3.3.6.1: Derivation

functions in one variable x over \mathbb{R} as follows,

$$D(c_0 + c_1 x + \dots + c_n x^n) = c_1 + c_2' x + \dots + c_n' x^{n-1}$$
(3.3.8.10)

We first prove that the vector space V is infinite dimensional.

Suppose to the contrary that V is finite dimensional vector space and is given by the span of k polynomials in V as follows,

$$span(\mathbb{V}) = \{p_1, p_2, \dots, p_k\}$$
 (3.3.8.11)

Also let m be the maximum of the degree of these k polynomials in (3.3.8.11). Now let an element of the vector space \mathbb{V} be,

$$cx^{m+1} \in \mathbb{V} \tag{3.3.8.12}$$

As maximum degree of the basis of \mathbb{V} is m hence cx^{m+1} cannot be represented by any

linear combination of the basis of \mathbb{V} . If \mathbb{F} is field corresponding to \mathbb{V} then we have,

$$cx^{m+1} \neq \sum_{i=1}^{k} \alpha_i p_i \quad [\alpha_i \in \mathbb{F} \ \forall i] \qquad (3.3.8.13)$$

Hence, cx^{m+1} is not in the span of p_1, p_2, \ldots, p_k . Hence, \mathbb{V} is infinite dimensional vector space.

Next we prove that D is not one-one operator. Let, two different elements from the vector space \mathbb{V} be as follows,

$$c_1 + x^m \in \mathbb{V} \tag{3.3.8.14}$$

$$c_2 + x^m \in \mathbb{V} \tag{3.3.8.15}$$

From definition (3.3.8.10) of operator D we

Given	$\mathcal{T}(T): T \to UTU^{-1}$
	U is isomorphism of V onto W that means U is $one - one$
	$\mathcal{T}: L(V, V) \to L(W, W)$
To prove	\mathcal{T} is isomorphism of $L(V, V)$ onto $L(W, W)$
	It is same as proving \mathcal{T} is invertible, because
	$isomorphim \implies one - one$ $\implies invertible$ by definition
Proof	Consider inverse transformation $S: L(W, W) \to L(V, V)$ $S: S \to U^{-1}SU$
	where $U^{-1}SU$ is a composition of 3 linear transformations $V \xrightarrow{U} W \xrightarrow{S} W \xrightarrow{U^{-1}} V$
	Now consider $S(UTU^{-1})$,
	$S(UTU^{-1}) = U^{-1}(UTU^{-1})U = T$
	Similarly consider $\mathcal{T}(U^{-1}SU)$,
	$\mathcal{T}(U^{-1}SU) = U(U^{-1}SU)U^{-1} = S$
	$\implies TS = I \text{ and } ST = I$
	we can say $\mathcal T$ is invertible since we have found an inverse $\mathcal S$
	Hence \mathcal{T} is one-one implies \mathcal{T} isomorphism of V onto W

TABLE 3.3.7.1: Proof

have,

$$D(c_1 + x^m) = mx^{m-1} (3.3.8.16)$$

$$D(c_2 + x^m) = mx^{m-1} (3.3.8.17)$$

From (3.3.8.16) and (3.3.8.17),

$$c_1 + x^m \neq c_2 + x^m$$
 (3.3.8.18)

$$D(c_1 + x^m) = D(c_2 + x^m)$$
 (3.3.8.19)

Hence from (3.3.8.19) we see that D is not One-One operator.

And, $U: \mathbb{V} \to \mathbb{V}$ is another linear operator such that,

$$U(c_0 + c_1 x + \dots + c_n x^n) = c_0 x + c_1 \frac{x^2}{2} + \dots + c_n \frac{x^{n+1}}{n+1}$$
(3.3.8.20)

Now, $DU : \mathbb{V} \to \mathbb{V}$ is a linear operator such that,

$$DU(c_0 + c_1 x + \dots + c_n x^n)$$
 (3.3.8.21)
= $D[U(c_0 x + c_1 \frac{x^2}{2} + \dots + c_n \frac{x^{n+1}}{n+1})]$ (3.3.8.22)

$$= D[c_0x + c_1\frac{x^2}{2} + \dots + c_n\frac{x^{n+1}}{n+1}] \quad (3.3.8.23)$$

$$= c_0 + c_1 \frac{2x}{2} + \dots + c_n \frac{(n+1)x^n}{n+1}$$
 (3.3.8.24)

$$= c_0 + c_1 x + \dots + c_n x^n \tag{3.3.8.25}$$

Hence, from (3.3.8.25),

$$DU = I$$
 (3.3.8.26)

Again $UD : \mathbb{V} \to \mathbb{V}$ is a linear operator such that,

$$UD(c_0 + c_1 x + \dots + c_n x^n)$$
 (3.3.8.27)

$$= U[D(c_0 x + c_1 \frac{x^2}{2} + \dots + c_n \frac{x^{n+1}}{n+1})]$$
 (3.3.8.28)

$$= U[c_1 + c_2' x + \dots + c_n' x^{n-1}]$$
 (3.3.8.29)

$$= c_1 x + c_2 \frac{x^2}{2} + \dots + c_n \frac{x^n}{n}$$
 (3.3.8.30)

Hence, from (3.3.8.30),

$$UD \neq I$$
 (3.3.8.31)

Hence, from (3.3.8.26) and (3.3.8.31), D is not invertible.

- 3.4 Representation of Transformations by Matrices
- 3.4.1. Let T be the linear operator on \mathbb{C}^2 defined by $T(x_1, x_2) = (x_1, 0)$. Let β be the standard ordered basis for \mathbb{C}^2 and $\beta' = \{\alpha_1, \alpha_2\}$ be the ordered basis defined by $\alpha_1 = (1, i), \alpha_2 = (-i, 2)$.

linear transformation	Let V and W be vector spaces over field F . A linear transformation V into W is a function T from V into W such that $T(c\alpha + \beta) = c(T\alpha) + T\beta$ for all α and β in V and all scalars in c in F .
isomorphism	If V and W are vector spaces over the field F , any $one-one$ linear transformation $T:V\to W$ is called isormorphism of V onto W
one-one	A linear transformation $T: \mathbb{R}^n \to \mathbb{R}^m$ is said to be one-one if for every $\mathbf{b} \in \mathbb{R}^m$, $\mathbf{A}\mathbf{X} = \mathbf{b}$ has at most one solution in \mathbb{R}^n . Equivalently, if $T(\mathbf{u}) = T(\mathbf{v})$, then $u = v$. By definition, all <i>invertible</i> transformations are one-one
invertible	A linear transformation $T: V \to W$ is invertible if there exists another linear transformation $U: W \to V$ such that UT is the <i>identity</i> transformation on V and TU is the identity transformation on W . T is invertible if and only if T is $one - one$ and $onto$

TABLE 3.3.7.2: Definitions

Proof	Conclusion	
Let $V_1, V_2 \in \mathbb{V}$ then,		
If $V_1 \neq V_2$ then,	T is one-one function	
$T[U(\mathbf{V_1})] \neq T[U(\mathbf{V_2})]$		
T is linear operator on		
finite dimensional	T is onto function	
vector space		

TABLE 3.3.8.1: Proof of Invertibility of transformation

a) What is the matrix of **T** relative to the pair β, β' ?

Solution:

$$\beta = \{\epsilon_1, \epsilon_2\} \implies \epsilon_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \epsilon_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$
(3.4.1.1)

Hence, β as matrix

$$\beta = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.4.1.2}$$

$$\beta' = \{\alpha_1, \alpha_2\} \implies \alpha_1 = \begin{pmatrix} 1 \\ i \end{pmatrix}, \alpha_2 = \begin{pmatrix} -i \\ 2 \end{pmatrix}$$
(3.4.1.3)

Hence, β' as matrix

$$\beta' = \begin{pmatrix} 1 & -i \\ i & 2 \end{pmatrix} \tag{3.4.1.4}$$

Geometrically, **T** is a projection onto the x-axis.

For projection, let Consider Matrix A as

$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.4.1.5}$$

The matrix A is representation of the linear transformation T that is projection on x-axis. After applying linear operator T on it,

$$\mathbf{T}(\beta) = \mathbf{A}\beta = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$
(3.4.1.6)

$$\begin{pmatrix} 1 & -i & 1 & 0 \\ i & 2 & 0 & 0 \end{pmatrix} \xrightarrow{R_2 = R_2 - iR_1} \begin{pmatrix} 1 & -i & 1 & 0 \\ 0 & 1 & -i & 0 \end{pmatrix}$$
(3.4.1.7)

$$\stackrel{R_1=R_1+iR_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 2 & 0 \\ 0 & 1 & -i & 0 \end{pmatrix} \tag{3.4.1.8}$$

Therefore matrix of relative to the pair β , β'

$$\mathbf{T}(\beta) = \begin{pmatrix} 2 & 0 \\ -i & 0 \end{pmatrix} \beta' \tag{3.4.1.9}$$

b) What is the matrix of T relative to the pair β , β' ?

Solution: Transformation T from C^2 to C^2 . Let

$$\beta = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.4.1.10}$$

$$\beta' = \begin{pmatrix} 1 & -i \\ i & 2 \end{pmatrix} \tag{3.4.1.11}$$

T is defined by

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.4.1.12}$$

$$T(\beta) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \beta \tag{3.4.1.13}$$

$$T(\beta) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \quad (3.4.1.14)$$

To find relative matrix we will use row reduce augmented matrix.

$$\begin{pmatrix} 1 & -i & 1 & 0 \\ i & 2 & 0 & 0 \end{pmatrix} \tag{3.4.1.15}$$

$$\begin{pmatrix} 1 & -i & 1 & 0 \\ i & 2 & 0 & 0 \end{pmatrix} \xrightarrow[R_1 \to R_1 + iR_2]{} \xrightarrow{R_2 \to R_2 - iR_1} \begin{pmatrix} 1 & 0 & 2 & 0 \\ 0 & 1 & -i & 0 \end{pmatrix}$$
(3.4.1.16)

Hence the matrix of T in the order basis of β'

$$\mathbf{B} = \begin{pmatrix} 2 & 0 \\ -i & 0 \end{pmatrix} \tag{3.4.1.17}$$

Therefore matrix of relative to the pair β ,

$$\beta'$$

$$T(\beta) = \mathbf{A}\beta = \mathbf{B}\beta' = \begin{pmatrix} 2 & 0 \\ -i & 0 \end{pmatrix} \beta' \quad (3.4.1.18)$$

c) What is the matrix of **T** in the ordered basis β'

Solution:

Let

$$\beta = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.4.1.19}$$

We have

$$\alpha_1 = \begin{pmatrix} 1 \\ i \end{pmatrix}, \alpha_2 = \begin{pmatrix} -i \\ 2 \end{pmatrix} \tag{3.4.1.20}$$

So,

$$\beta' = \begin{pmatrix} 1 & -i \\ i & 2 \end{pmatrix} \tag{3.4.1.21}$$

Geometrically, **T** is a projection onto the x-axis.

For projection, let Consider Matrix A as

$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.4.1.22}$$

The matrix **A** is representation of the linear transformation **T** that is projection on x-axis. After applying linear operator **T** on it,

$$\mathbf{T}(\beta') = \mathbf{A}\beta' = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & -i \\ i & 2 \end{pmatrix} = \begin{pmatrix} 1 & -i \\ 0 & 0 \end{pmatrix}$$
(3.4.1.23)

Now, for finding the matrix of **T** in the ordered basis β' , we combine the 3.4.1.20 and 3.4.1.23 and use concept of row-reduction of the augmented matrix:

$$\begin{pmatrix} 1 & -i & 1 & -i \\ i & 2 & 0 & 0 \end{pmatrix} \xrightarrow{R_2 = R_2 - iR_1} \begin{pmatrix} 1 & -i & 1 & -i \\ 0 & 1 & -i & -1 \end{pmatrix}$$

$$(3.4.1.24)$$

$$\xrightarrow{R_1 = R_1 + iR_2} \begin{pmatrix} 1 & 0 & 2 & -2i \\ 0 & 1 & -i & -1 \end{pmatrix}$$

$$(3.4.1.25)$$

Hence, the matrix of **T** in the ordered basis

 β' is

$$\mathbf{B} = \begin{pmatrix} 2 & 2i \\ -i & -1 \end{pmatrix} \tag{3.4.1.26}$$

And this can also be represented using β and β' , which shows the relation between **T**, β and β' .

$$\mathbf{T}(\beta) = \mathbf{A}\beta = \mathbf{B}\beta' \tag{3.4.1.27}$$

d) What is the matrix of T in the ordered basis $\{\alpha_2, \alpha_1\}$?

Solution: Transformation T from \mathbb{C}^2 to \mathbb{C}^2 . Let

$$\mathbf{e_1} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \mathbf{e_2} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \tag{3.4.1.28}$$

$$\beta = \begin{pmatrix} \mathbf{e_1} & \mathbf{e_2} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.4.1.29}$$

$$\beta' = \begin{pmatrix} \alpha_2 & \alpha_1 \end{pmatrix} = \begin{pmatrix} -i & 1\\ 2 & i \end{pmatrix} \qquad (3.4.1.30)$$

T in the ordered basis β is:

$$[\mathbf{T}]_{\beta} = \begin{pmatrix} -i & 1\\ 2 & i \end{pmatrix} \tag{3.4.1.31}$$

T is defined by

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.4.1.32}$$

$$T(\mathbf{x}) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \mathbf{x} \tag{3.4.1.33}$$

$$T(\alpha_2) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} -i \\ 2 \end{pmatrix} = \begin{pmatrix} -i \\ 0 \end{pmatrix} \qquad (3.4.1.34)$$

$$T(\alpha_1) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ i \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$
 (3.4.1.35)

$$[\alpha]_{\beta} = \begin{pmatrix} -i & 1\\ 0 & 0 \end{pmatrix} \tag{3.4.1.36}$$

The matrix of T in the ordered basis $\{\alpha_2, \alpha_1\}$ is given as:

$$[\mathbf{T}_{\alpha}]_{\beta} = [\mathbf{T}]_{\beta}[\alpha]_{\beta} = \begin{pmatrix} -i & 1\\ 2 & i \end{pmatrix} \begin{pmatrix} -i & 1\\ 0 & 0 \end{pmatrix} = \begin{pmatrix} -1 & -i\\ -2i & 2 \end{pmatrix}$$
(3.4.1.37)

3.4.2. Let **T** be the linear transformation from \mathbb{R}^3 into \mathbb{R}^2 defined by

$$\mathbf{T}(x_1, x_2, x_3) = (x_1 + x_2, 2x_3 - x_1) \quad (3.4.2.1)$$

If
$$\beta = (\alpha_1, \alpha_2, \alpha_3)$$
 and $\beta' = (\beta_1, \beta_2)$ where $\alpha_1 = (1, 0, -1), \alpha_2 = (1, 1, 1), \alpha_3 = (1, 0, 0)$

$$\beta_1 = (0,1), \beta_2 = (1,0)$$

a) What is the matrix of **T** relative to the pair β , β'

Solution: Let

$$\beta = (\alpha_1, \alpha_2, \alpha_3) \tag{3.4.2.2}$$

$$\implies \beta = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ -1 & 1 & 0 \end{pmatrix} \tag{3.4.2.3}$$

and

$$\beta' = \{\beta_1, \beta_2\} \tag{3.4.2.4}$$

$$\implies \beta' = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \tag{3.4.2.5}$$

T is defined by

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.4.2.6}$$

using equation (3.4.2.1)

$$T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ 2x_3 - x_1 \end{pmatrix}$$
 (3.4.2.7)

R.H.S of the equation can be written as a product of 2×3 and 3×1 matrices,

$$= \begin{pmatrix} 1 & 1 & 0 \\ -1 & 0 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$
 (3.4.2.8)

$$\implies \mathbf{A} = \begin{pmatrix} 1 & 1 & 0 \\ -1 & 0 & 2 \end{pmatrix} \tag{3.4.2.9}$$

Now,

$$T(\beta) = \begin{pmatrix} 1 & 1 & 0 \\ -1 & 0 & 2 \end{pmatrix} \beta \tag{3.4.2.10}$$

$$T(\beta) = \begin{pmatrix} 1 & 1 & 0 \\ -1 & 0 & 2 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ -1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 1 \\ -3 & 1 & -1 \end{pmatrix}$$
(3.4.2.11)

To find relative matrix we will use row

reduce augmented matrix.

$$\begin{pmatrix} 1 & 2 & 1 & 0 & 1 \\ -3 & 1 & -1 & 1 & 0 \end{pmatrix} \tag{3.4.2.12}$$

$$\begin{pmatrix} 1 & 2 & 1 & | & 0 & 1 \\ -3 & 1 & -1 & | & 1 & 0 \end{pmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{pmatrix} -3 & 1 & -1 & | & 1 & 0 \\ 1 & 2 & 1 & | & 0 & 1 \end{pmatrix}$$

$$(3.4.2.13)$$

Hence the matrix of **T** in the order basis of β'

$$\mathbf{B} = \begin{pmatrix} -3 & 1 & -1 \\ 1 & 2 & 1 \end{pmatrix} \tag{3.4.2.14}$$

Therefore matrix of relative to the pair β , β'

$$T(\beta) = \mathbf{A}\beta = \mathbf{B}\beta' = \begin{pmatrix} -3 & 1 & -1 \\ 1 & 2 & 1 \end{pmatrix}\beta'$$
(3.4.2.15)

3.4.3. Let T be the linear transformation from \mathbb{R}^3 into \mathbb{R}^2 defined by,

$$T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ 2x_3 - x_1 \end{pmatrix}$$
 (3.4.3.1)

- a) If β is the standard ordered basis for \mathbb{R}^3 and β' is the standard ordered basis for \mathbb{R}^2 , what is the matrix of T relative to the pair β , β'
- 3.4.4. Let T be a linear operator on $\mathbf{F^n}$, let \mathbf{A} be the matrix of T in the standard ordered basis for $\mathbf{F^n}$, and let W be the subspace of $\mathbf{F^n}$ spanned by the column vectors of \mathbf{A} . What does W have to do with T?

Solution: See Table 3.4.4.1.

3.4.5. Let *V* be a two-dimensional vector space over the field *F* and let *B* be an ordered basis for *V*. If *T* is a linear operator on *V* and

$$[T]_B = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \tag{3.4.5.1}$$

Prove that

$$T^{2} - (a+d)T + (ad - bc)I = 0 (3.4.5.2)$$

Solution: Here T is a linear operator on V and B is an ordered basis of V. Let us consider $[T]_B = A$, so $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$.

Now, the characteristic equation of A is:

$$\begin{vmatrix} A - \lambda I \end{vmatrix} = 0 \quad (3.4.5.3)$$

$$\implies \begin{vmatrix} a - \lambda & b \\ c & d - \lambda \end{vmatrix} = 0 \quad (3.4.5.4)$$

$$\implies \lambda^2 - (a+d)\lambda + (ad-bc) = 0 \quad (3.4.5.5)$$

According to the Cayley-Hamilton's Theorem, every square matrix satisfies its own characteristic equation. Here *A* is a 2x2 square matrix, so it should also satisfy its characteristic equation. Now.

$$\lambda^{2} - (a+d)\lambda + (ad-bc) = 0$$

$$(3.4.5.6)$$

$$\implies A^{2} - (a+d)A + (ad-bc)I = 0$$

$$(3.4.5.7)$$

We can also write the equation 3.4.5.7 as:

$$[T]_B^2 - (a+d)[T]_B + (ad-bc)I = 0$$
 (3.4.5.8)
or, $T^2 - (a+d)T + (ad-bc)I = 0$ (3.4.5.9)

3.4.6. Let T be a linear operator on \mathbb{R}^3 , the matrix of which in the standard ordered basis is,

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ -1 & 3 & 4 \end{pmatrix} \tag{3.4.6.1}$$

Find a basis for the range of T and a basis for the null-space of T.

Solution: The basis of the range of linear transformation T is the basis of the column-space of A or basis of C(A). Hence the basis of the range of the linear transformation T is derived by reducing A into Reduced-Row Echelon form as follows,

$$\begin{pmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ -1 & 3 & 4 \end{pmatrix} \xrightarrow{R_3 = R_3 + R_1} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 5 & 5 \end{pmatrix}$$
 (3.4.6.2)
$$\xrightarrow{R_3 = R_3 - 5R_2} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$
 (3.4.6.3)

From (3.4.6.3) the basis of the range of linear operator T are as follows,

$$\mathbf{a_1} = \begin{pmatrix} 1 & 0 & 0 \end{pmatrix} \tag{3.4.6.4}$$

$$\mathbf{a_2} = \begin{pmatrix} 0 & 1 & 0 \end{pmatrix} \tag{3.4.6.5}$$

Again, the basis for null-space of linear oper-

Let $\{\alpha_1, \alpha_2, \cdots, \alpha_n\}$ be an ordered basis of \mathbf{F}^n					
Given	Explanation				
T is a linear operator F ⁿ	As T is linear,				
	$T(\mathbf{x}) = \mathbf{A}\mathbf{x}$	(3.4.4.1)			
	$\mathbf{A} = \begin{pmatrix} T\epsilon_1 & T\epsilon_2 & \cdots & T\epsilon_n \end{pmatrix}$	(3.4.4.2)			
	From equation (3.4.4.2), columns of $\bf A$ are the images of the standard belements of $\bf F^n$.				
Range of T	range(T) = { $T \epsilon_1, T \epsilon_2, \cdots, T \epsilon_n$ }	(3.4.4.3)			
	From equation (3.4.4.2) and (3.4.4.3), column	mns of A generate the range of T.			
W spanned by column vectors A	Since any generating set contains a basis for the generated space, we can say that the columns of A contains a basis of the range of T. As W is spanned by column vectors of A , we can say that W contains a basis for the range of T.				

TABLE 3.4.4.1: Expanation

ator T or $N(\mathbf{A})$ is a solution of the equation $\mathbf{A}\mathbf{x} = 0$. From (3.4.6.3) we have,

$$\mathbf{A}\mathbf{x} = 0$$
 (3.4.6.6)

$$\implies \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = 0 \tag{3.4.6.7}$$

Setting the value of the free variable $x_3 = 1$ we get the solution,

$$\mathbf{x} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \tag{3.4.6.8}$$

Hence, the basis of the null-space of the linear operator T is given by,

$$\mathbf{b} = \begin{pmatrix} 1 & -1 & 1 \end{pmatrix} \tag{3.4.6.9}$$

3.4.7. Let T be the linear operator on \mathbb{R}^2 defined by

$$T(x_1, x_2) = (-x_2, x_1)$$
 (3.4.7.1)

a) What is the matrix of **T** in the standard ordered basis of \mathbb{R}^2 ?

Solution:

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \qquad (3.4.7.2)$$

$$\mathbf{T}(\mathbf{x}) = \mathbf{T}\mathbf{x} \qquad (3.4.7.3)$$

$$\implies \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -x_2 \\ x_1 \end{pmatrix} \qquad (3.4.7.4)$$

The matrix of **T** in the standard ordered basis

from (3.4.7.1) is

$$\mathbf{T} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \tag{3.4.7.5}$$

b) Let T be the linear operator on \mathbb{R}^2 defined by

$$T\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -x_2 \\ x_1 \end{pmatrix} \tag{3.4.7.6}$$

What is the matrix of T in the ordered basis $\mathbf{B} = \{\alpha_1, \alpha_2\}$, where $\alpha_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ and $\alpha_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$?

Solution:

Let \mathbf{B}' be the standard ordered basis for \mathbf{R}^2 . Then,

$$T(\alpha_1') = T \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} = 0\alpha_1 + 1\alpha_2 \quad (3.4.7.7)$$

$$T(\alpha_2') = T \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \end{pmatrix} = -1\alpha_1 + 0\alpha_2 \quad (3.4.7.8)$$

Hence, the matrix of T in the standard ordered basis \mathbf{B}' is

$$T_{\mathbf{B}'} = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \tag{3.4.7.9}$$

Given,

$$\alpha_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \tag{3.4.7.10}$$

$$\alpha_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix} \tag{3.4.7.11}$$

Then,

$$\alpha_1' = 1\alpha_1 + 2\alpha_2 \tag{3.4.7.12}$$

$$\alpha_2' = 1\alpha_1 - 1\alpha_2 \tag{3.4.7.13}$$

So the **P** matrix is

$$\mathbf{P} = \begin{pmatrix} 1 & 1 \\ 2 & -1 \end{pmatrix}$$
 (3.4.7.14) 3.4.8. Let \mathbb{T} be the linear operator on \mathbb{R}^3 defined by

And

$$\mathbf{P}^{-1} = \begin{pmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{2}{3} & -\frac{1}{3} \end{pmatrix}$$
 (3.4.7.15)

Hence

$$[T]_{\mathbf{B}} = \mathbf{P}^{-1}[T]_{\mathbf{B}'}\mathbf{P} \qquad (3.4.7.16)$$

$$= \begin{pmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{2}{3} & -\frac{1}{3} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 2 & -1 \end{pmatrix} \qquad (3.4.7.17)$$

$$= \begin{pmatrix} -\frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{2}{3} \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 2 & -1 \end{pmatrix} \qquad (3.4.7.18)$$

$$[T]_{\mathbf{B}} = \begin{pmatrix} \frac{1}{3} & -\frac{2}{3} \\ \frac{2}{3} & -\frac{1}{3} \end{pmatrix} \qquad (3.4.7.19)$$

Hence, $[T]_{\mathbf{B}}$ is the required matrix for the given ordered basis B.

c) Prove that for every real number c, the operator (T - cI) is invertible.

Solution: From the equation (3.4.7.1), the 3.4.9. The linear operator \mathbf{T} on \mathbf{R}^2 defined by matrix of T in standard order basis is,

$$\mathbf{T} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \tag{3.4.7.20}$$

To find the invertibility of the operator $(\mathbf{T} - c\mathbf{I})$ for every real number c, let us start with

$$(\mathbf{T} - c\mathbf{I}) (\mathbf{T} + c\mathbf{I})$$
 (3.4.7.21)^{3.4.10}
= $\mathbf{T}^2 - c^2\mathbf{I}$ (3.4.7.22)

Consider T^2

$$\mathbf{T}^2 = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \Longrightarrow \mathbf{T}^2 = -\mathbf{I}$$
(3.4.7.23)

Substituting equation (3.4.7.23)in (3.4.7.22),

$$(\mathbf{T} - c\mathbf{I})(\mathbf{T} + c\mathbf{I}) = -(1 + c^2)\mathbf{I}$$
 (3.4.7.24)

As c is a real number, $c^2 \ge 0$ and hence factor $-(1+c^2)$ is always non-zero. Therefore, from the equation (3.4.7.24),

$$(\mathbf{T} - c\mathbf{I})^{-1} = \frac{-1}{1+c^2} (\mathbf{T} + c\mathbf{I})$$
 (3.4.7.25)

Hence the operator (T - cI) is invertible and its inverse is given by the equation (3.4.7.25)

$$\mathbb{T}(x_1, x_2, x_3) =$$
(3.4.8.1)

$$(3x_1 + x_3, -2x_1 + x_2, -x_1 + 2x_2 + 4x_3)$$
 (3.4.8.2)

What is the matrix of \mathbb{T} in the standard ordered basis of \mathbb{R}^3 ?

Solution:

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \tag{3.4.8.3}$$

$$\mathbb{T}(\mathbf{x}) = \mathbf{T}\mathbf{x} \tag{3.4.8.4}$$

The matrix of \mathbb{T} in the standard ordered basis from (3.4.8.2) is

$$\mathbf{T} = \begin{pmatrix} 3 & 0 & 1 \\ -2 & 1 & 0 \\ -1 & 2 & 4 \end{pmatrix} \tag{3.4.8.5}$$

$$\mathbf{T} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ 0 \end{pmatrix} \tag{3.4.9.1}$$

is represented by the matrix

$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.4.9.2}$$

 $(3.4.7.21)^{3.4.10}$. Let θ be a real number. Prove that the following two matrices are similar over the field of complex numbers:

$$\begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}, \quad \begin{pmatrix} e^{i\theta} & 0 \\ 0 & e^{-i\theta} \end{pmatrix}$$

Solution: Two matrices **A** and **B** are said to

be similar iff there exists a invertible matrix **P** such that:

$$\mathbf{B} = \mathbf{P}^{-1}\mathbf{A}\mathbf{P} \tag{3.4.10.1}$$

Let,

$$\mathbf{A} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}; \quad \mathbf{B} = \begin{pmatrix} e^{i\theta} & 0 \\ 0 & e^{-i\theta} \end{pmatrix}$$
(3.4.10.2)

Finding the characteristic polynomial of A,

$$|A - \lambda I| = \begin{vmatrix} \cos \theta - \lambda & -\sin \theta \\ \sin \theta & \cos \theta - \lambda \end{vmatrix}$$
$$= (\cos \theta - \lambda)^2 + \sin^2 \theta$$
$$= 1 + \lambda^2 + 2\lambda \cos \theta \qquad (3.4.10.3)$$

The eigenvalues can be calculated by equating the characteristic polynomial to zero. The eigenvalues are,

$$\lambda_1 = \cos \theta + i \sin \theta; \ \lambda_2 = \cos \theta - i \sin \theta$$
(3.4.10.4)

The eigenvectors corresponding to (3.4.10.4) are,

$$\alpha_1 = \begin{pmatrix} i \\ 1 \end{pmatrix}; \ \alpha_2 = \begin{pmatrix} -i \\ 1 \end{pmatrix}$$
 (3.4.10.5)

$$\mathbf{P} = \begin{pmatrix} \alpha_1 & \alpha_2 \end{pmatrix} = \begin{pmatrix} i & -i \\ 1 & 1 \end{pmatrix} \tag{3.4.10.6}$$

Now,

$$\mathbf{P}^{-1}\mathbf{A}\mathbf{P} = \frac{1}{2i} \begin{pmatrix} 1 & i \\ -1 & i \end{pmatrix} \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} i & -i \\ 1 & 1 \end{pmatrix}$$
$$= \begin{pmatrix} \cos \theta + i \sin \theta & 0 \\ 0 & \cos \theta - i \sin \theta \end{pmatrix}$$
$$= \begin{pmatrix} e^{i\theta} & 0 \\ 0 & e^{-i\theta} \end{pmatrix} = \mathbf{B}$$
(3.4.10.7)

Hence, from (3.4.10.7), **A** and **B** are similar matrices.

3.4.11. Let \mathbb{V} be finite dimensional vector space over the field \mathbb{F} , and let \mathbf{S} and \mathbf{T} be linear operators on \mathbb{V} . When do there exist ordered bases \mathcal{B} and \mathcal{B}' for \mathbb{V} such that $[S]_{\mathcal{B}} = [T]'_{\mathcal{B}}$? Prove that such bases exist if and only if there is an invertible linear operator \mathbf{U} on \mathbb{V} such that $\mathbf{T} = \mathbf{USU}^{-1}$

Solution: See Table 3.4.11.1.

3.4.12. Prove that if **S** is a linear operator on \mathbb{R}^2 such that $\mathbb{S}^2 = \mathbb{S}$, then $\mathbb{S} = \mathbb{O}$, or $\mathbb{S} = \mathbb{I}$, or there is

an ordered basis **B** for \mathbb{R}^2 such that $[S]_B = A$. **Solution:** If a linear operator **S** is defined on \mathbb{R}^2 such that $\mathbb{S}^2 = \mathbb{S}$, then

$$S^2 - S = 0 (3.4.12.1)$$

$$S(S - I) = 0 (3.4.12.2)$$

$$\implies \mathbf{S} = \mathbf{0}, \mathbf{S} = \mathbf{I} \tag{3.4.12.3}$$

The transformation of a vector $\mathbf{x} \in \mathbf{R}^2$ can be represented as

$$Sx = y$$
 (3.4.12.4)

$$\implies$$
 S(Sx) = Sy (3.4.12.5)

$$\implies \mathbf{S}^2 \mathbf{x} = \mathbf{S} \mathbf{y} \tag{3.4.12.6}$$

$$\implies$$
 Sx = Sy (3.4.12.7)

$$\implies \mathbf{x} = \mathbf{y} \tag{3.4.12.8}$$

Therefore the transformation of a vector $\mathbf{x} \in \mathbf{R}^2$ can be given as

$$\mathbf{S}\mathbf{x} = \mathbf{x} \ \forall \ \mathbf{x} \in \mathbf{R}^2 \tag{3.4.12.9}$$

Consider the ordered basis set

$$B = \{\epsilon_1, \epsilon_2\} \in \mathbf{R}^2$$
 (3.4.12.10)

and if

$$[S]_B = A$$
 (3.4.12.11)

$$\Longrightarrow [\mathbf{S}]_{\mathbf{B}} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.4.12.12}$$

Thus we can re-write the column vectors of $[S]_B$ using (3.4.12.9) as

$$\mathbf{S}\begin{pmatrix}1\\0\end{pmatrix} = \begin{pmatrix}1\\0\end{pmatrix} = 1\begin{pmatrix}1\\0\end{pmatrix} + 0\begin{pmatrix}0\\1\end{pmatrix} \qquad (3.4.12.13)$$

$$\mathbf{S}\begin{pmatrix}0\\0\end{pmatrix} = \begin{pmatrix}0\\0\end{pmatrix} = 0\begin{pmatrix}1\\0\end{pmatrix} + 0\begin{pmatrix}0\\1\end{pmatrix} \qquad (3.4.12.14)$$

Therefore, any vector \mathbf{x} in column space of $[\mathbf{S}]_{\mathbf{B}}$ can be uniquely expressed by $\left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}$, hence it forms the basis for column space of $[\mathbf{S}]_{\mathbf{B}}$. Therefore one of the basis vector of B is $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$. The other basis vector can be any vector

which is linearly independent to $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$. One of the ordered basis set can be

$$B = \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\} \tag{3.4.12.15}$$

Assume $[S]_{\mathcal{B}} = [T]_{\mathcal{B}'}$	Assume $T = USU^{-1}$
Given	Given
\mathbb{V} is a finite dimensional vector space over field \mathbb{F} S and T are linear operators on \mathbb{V} \mathcal{B} and \mathcal{B}' are ordered bases for \mathbb{V} $[S]_{\mathcal{B}} = [T]'_{\mathcal{B}}$	$\mathbb V$ is a finite dimensional vector space over field $\mathbb F$ $\mathbf S$ and $\mathbf T$ are linear operators on $\mathbb V$ $\mathcal B$ and $\mathcal B'$ are ordered bases for $\mathbb V$ There is an invertible linear operator $\mathbf U$ on $\mathbb V$ such that $\mathbf T = \mathbf U \mathbf S \mathbf U^{-1}$
To prove	To prove
There is an invertible linear operator \mathbf{U} on \mathbb{V} such that $\mathbf{T} = \mathbf{U}\mathbf{S}\mathbf{U}^{-1}$	$[S]_{\mathcal{B}} = [T]_{\mathcal{B}}'$
Assumptions	Assumptions
Let U be the operator which carries \mathcal{B} to \mathcal{B}' $\mathcal{B} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n\}$ $\mathcal{B}' = \{\mathbf{x}'_1, \mathbf{x}'_2, \dots, \mathbf{x}'_n\}$	Let U be the operator which carries \mathcal{B} to \mathcal{B}' $\mathcal{B} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n\}$ $\mathcal{B}' = \{\mathbf{x}'_1, \mathbf{x}'_2, \dots, \mathbf{x}'_n\}$
Proof	Proof
For $\mathbf{v} \in \mathbb{V}$, expressed as a linear combination of the vectors of \mathcal{B} $\mathbf{v} = a_1 \mathbf{x}_1 + a_2 \mathbf{x}_2 + \ldots + a_n \mathbf{x}_n$ $\mathbf{w} = \mathbf{U}(\mathbf{v}) = a_1 \mathbf{x}'_1 + a_2 \mathbf{x}'_2 + \ldots + a_n \mathbf{x}'_n$	For $\mathbf{v} \in \mathbb{V}$, expressed as a linear combination of the vectors of \mathcal{B}' $\mathbf{v} = a_1 \mathbf{x}'_1 + a_2 \mathbf{x}'_2 + \ldots + a_n \mathbf{x}'_n$ $\mathbf{USU}^{-1}(\mathbf{v}) = \mathbf{T}(\mathbf{v})$
$: [S]_{\mathcal{B}} = [T]'_{\mathcal{B}} $ $ \mathbf{S}(\mathbf{v}) = c_1 \mathbf{x}_1 + c_2 \mathbf{x}_2 + \ldots + c_n \mathbf{x}_n \text{and} $ $ \mathbf{T}(\mathbf{w}) = c_1 \mathbf{x}'_1 + c_2 \mathbf{x}'_2 + \ldots + c_n \mathbf{x}'_n $	$= \mathbf{T}(a_1\mathbf{x}_1' + a_2\mathbf{x}_2' + \dots + a_n\mathbf{x}_n')$ $= c_1\mathbf{x}_1' + c_2\mathbf{x}_2' + \dots + c_n\mathbf{x}_n'$ But we know that $\mathbf{USU}^{-1}(\mathbf{x}) = \mathbf{USU}^{-1}(a_1\mathbf{x}_1' + a_2\mathbf{x}_2' + \dots + a_n\mathbf{x}_n')$
$\mathbf{U}^{-1}\mathbf{T}\mathbf{U}(\mathbf{v})$ $=\mathbf{U}^{-1}\mathbf{T}(\mathbf{w})$	$\mathbf{USU}^{-1}(\mathbf{v}) = \mathbf{USU}^{-1}(a_1\mathbf{x}_1' + a_2\mathbf{x}_2' + \dots + a_n\mathbf{x}_n')$ = $\mathbf{US}(a_1\mathbf{x}_1 + a_2\mathbf{x}_2 + \dots + a_n\mathbf{x}_n)$
$= \mathbf{U}^{-1}(c_1\mathbf{x}'_1 + c_2\mathbf{x}'_2 + \dots + c_n\mathbf{x}'_n)$ $= c_1\mathbf{x}_1 + c_2\mathbf{x}_2 + \dots + c_n\mathbf{x}_n$ $= \mathbf{S}(\mathbf{v})$	So, S in basis \mathcal{B} has the same entries as T in basis \mathcal{B} $\therefore [S]_{\mathcal{B}} = [T]_{\mathcal{B}'}$
$\implies \mathbf{U}^{-1}\mathbf{T}\mathbf{U} = \mathbf{S}$ $\implies \mathbf{T} = \mathbf{U}\mathbf{S}\mathbf{U}^{-1}$	

TABLE 3.4.11.1

3.4.13. Let **W** be the space of all $n \times 1$ column matrices over a field **F**. If **A** is an $n \times n$ matrix over **F**, then **A** defines a linear operator $\mathbf{L}_{\mathbf{A}}$ on **W** through left multiplication: $\mathbf{L}_{\mathbf{A}}(\mathbf{X}) = \mathbf{A}\mathbf{X}$.

Prove that every linear operator on **W** is left multiplication by some $n \times n$ matrix, i.e., is $\mathbf{L}_{\mathbf{A}}$ for some **A**.

Now suppose V is an n-dimensional vector

Defining Linear Map T	Let $\mathbf{T}: \mathbf{W} \to \mathbf{W}$ be a linear operator and $(e_1, e_2,, e_n)$ be a basis for \mathbf{W} . Now, $\mathbf{Te_1} = \alpha_{11}\mathbf{e_1} + \alpha_{12}\mathbf{e_2} + + \alpha_{1n}\mathbf{e_n}$ $\mathbf{Te_2} = \alpha_{21}\mathbf{e_1} + \alpha_{22}\mathbf{e_2} + + \alpha_{2n}\mathbf{e_n}$ \vdots $\mathbf{Te_n} = \alpha_{n1}\mathbf{e_1} + \alpha_{n2}\mathbf{e_2} + + \alpha_{nn}\mathbf{e_n}$
Matrix of Linear Map T	Let A be matrix of linear transformation T . Then A is $\mathbf{A} = \begin{pmatrix} \alpha_{11} & \alpha_{21} & \dots & \alpha_{n1} \\ \alpha_{12} & \alpha_{22} & \dots & \alpha_{n2} \\ \vdots & \vdots & \dots & \vdots \\ \alpha_{1n} & \alpha_{2n} & \dots & \alpha_{nn} \end{pmatrix}$
Proving every linear operator on W is left multiplication by some $n \times n$ matrix, i.e, is L_A for some A	$\begin{aligned} \mathbf{A}\mathbf{e_1} &= (\alpha_{11}, \alpha_{12},, \alpha_{1n}) \\ &= \alpha_{11}\mathbf{e_1} + \alpha_{12}\mathbf{e_2} + + \alpha_{1n}\mathbf{e_n} \\ &= \mathbf{T}\mathbf{e_i} \end{aligned}$ Hence, $\begin{aligned} \mathbf{T}\mathbf{e_i} &= \mathbf{A}\mathbf{e_i} \\ \text{Since T and A are linear. Then} \\ \mathbf{T}\mathbf{x} &= \mathbf{A}\mathbf{x} \\ \mathbf{T}\mathbf{x} &= \mathbf{L_A}\mathbf{x} \end{aligned}$

TABLE 3.4.13.1

space over the field \mathbf{F} , and let β be an ordered basis for \mathbf{V} . For each α in \mathbf{V} , define $\mathbf{U}_{\alpha} = [\alpha]_{\beta}$. Prove that \mathbf{U} is an isomorphism of \mathbf{V} onto \mathbf{W} . If \mathbf{T} is a linear operator on \mathbf{V} , then $\mathbf{U}\mathbf{T}\mathbf{U}^{-1}$ is a linear operator on \mathbf{W} . Accordingly, $\mathbf{U}\mathbf{T}\mathbf{U}^{-1}$ is left multiplication by some $n \times n$ matrix \mathbf{A} . What is \mathbf{A} ?

Solution: See Tables 3.4.13.1 and 3.4.13.2.

3.4.14. Let V be an n-dimensional vector space over the field F, and let $\mathcal{B} = \{\alpha_1, \dots, \alpha_n\}$ be an ordered basis for V then there is a unique linear operator T on V such that

$$T\alpha_j = \alpha_{j+1}, j = 1, \dots, n-1$$
 (3.4.14.1)
 $T\alpha_n = 0.$ (3.4.14.2)

a) What is the matrix A of T in the ordered basis \mathcal{B} ?

Solution: Given that,

$$T: V \to V \tag{3.4.14.3}$$

$$[T(\alpha)]_{\mathcal{B}} = A[\alpha]_{\mathcal{B}} \tag{3.4.14.4}$$

$$T\alpha_i = \alpha_{i+1} \tag{3.4.14.5}$$

$$T\alpha_n = 0 \tag{3.4.14.6}$$

where j = 1, ..., n - 1. The matrix A of T in the ordered basis \mathcal{B} is given by,

$$\implies A = ([T\alpha_1]_{\mathcal{B}} \cdots [T\alpha_n]_{\mathcal{B}})$$
(3.4.14.7)

For $j = 1, \ldots, n-1$ we have,

$$T\alpha_i = \alpha_{i+1} \tag{3.4.14.8}$$

we can write,

$$T\alpha_{j} = 0\alpha_{1} + \ldots + 0\alpha_{j} + 1\alpha_{j+1} + \ldots + 0\alpha_{n}$$

$$(3.4.14.9)$$

$$\implies [T\alpha_{j}]_{\mathcal{B}} = (0, \ldots, 0, 1, 0, \ldots, 0)^{T}$$

$$(3.4.14.10)$$

Proving U as Linear and defining a linear map T	$\mathbf{U}(c\alpha_{1} + \alpha_{2}) = [c\alpha_{1} + \alpha_{2}]_{\beta}$ $= c[\alpha_{1}]_{\beta} + [\alpha_{2}]_{\beta}$ $= c\mathbf{U}(\alpha_{1}) + \mathbf{U}(\alpha_{2})$ Suppose $\beta = \{\alpha_{1}, \alpha_{2},, \alpha_{n}\}$ be the ordered basis for \mathbf{V} . Let \mathbf{T} be the function from \mathbf{W} to \mathbf{V} as follows: $\begin{pmatrix} a_{1} \\ a_{2} \\ \vdots \\ a_{n} \end{pmatrix} \rightarrow a_{1}\alpha_{1} + a_{2}\alpha_{2} + + a_{n}\alpha_{n}$
Proving U to be an isomorphism	For isomorphism, we must show that TU is identity map on V and UT is an identity map on W .
TU = I _V	$\mathbf{TU}(\mathbf{x}) = \mathbf{TU}(a_1\alpha_1 + a_2\alpha_2 + \dots + a_n\alpha_n)$ $= \mathbf{TU}(a_1\alpha_1 + a_2\alpha_2 + \dots + a_n\alpha_n)$ $= a_1\mathbf{T} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + a_2\mathbf{T} \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} + \dots + a_n\mathbf{T} \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$ $= a_1\alpha_1 + a_2\alpha_2 + \dots + a_n\alpha_n$ Hence, \mathbf{TU} is identity map on \mathbf{V} .
$\mathbf{UT} = \mathbf{I}_{\mathbf{W}}$	$\mathbf{UT}(\mathbf{x}) = \mathbf{UT}(a_{1}e_{1} + a_{2}e_{2} + \dots + a_{n}e_{n})$ $= \mathbf{UT}((a_{1}e_{1} + a_{2}e_{2} + \dots + a_{n}e_{n})$ $= \mathbf{U}(a_{1}\mathbf{T} \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} + a_{2}\mathbf{T} \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} + \dots + a_{n}\mathbf{T} \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$ $= a_{1}\mathbf{U}(\alpha_{1}) + a_{2}\mathbf{U}(\alpha_{2}) + \dots + a_{n}\mathbf{U}(\alpha_{n})$ $= a_{1}e_{1} + a_{2}e_{2} + \dots + a_{n}e_{n}$ Hence, \mathbf{UT} is identity map on \mathbf{W} .
Matrix of UTU ⁻¹	Now, we define the matrix of $\mathbf{U}\mathbf{T}\mathbf{U}^{-1}$. Since $\mathbf{U}\alpha_i$ is the standard $n\times 1$ matrix with all zeros except in the ith place which equals one. Let β' be the standard basis for \mathbf{W} . Then the matrix of \mathbf{U} with respect to β and β' is the identity matrix. Likewise the matrix of \mathbf{U}^{-1} with respect to β' and β is the identity matrix. Thus, $[\mathbf{U}\mathbf{T}\mathbf{U}^{-1}]_{\beta} = \mathbf{I}[\mathbf{T}]_{\beta}\mathbf{I}^{-1} = [T]_{\beta}$ Thus, the matrix \mathbf{A} is simply $[\mathbf{T}]_{\beta}$, the matrix of \mathbf{T} with respect to β .

where 1 is in (j + 1)th position. Now,

$$T\alpha_n = 0$$
 (3.4.14.11)

$$\implies [T\alpha_n]_{\mathcal{B}} = 0 \tag{3.4.14.12}$$

Thus from (3.4.14.7), (3.4.14.10) and (3.4.14.12) we get matrix A of T in the ordered basis \mathcal{B} as,

$$A = \begin{pmatrix} 0 & 0 & 0 & 0 & \cdots & 0 & 0 \\ 1 & 0 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 1 & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & 1 & 0 \end{pmatrix}$$
 (3.4.14.13)

3.4.15. Let V and W be finite-dimensional vector spaces over the field F and let T be a linear transformation from V into W. If

$$\mathcal{B} = \{\alpha_1, \dots, \alpha_n\} \text{ and } \mathcal{B}' = \{\beta_1, \dots, \beta_m\}$$
(3.4.15.1)

are ordered bases for V and W, respectively, define the linear transformation $E^{p,q}$ as in the proof of Theorem 5: $E^{p,q}(\alpha_i) = \delta_{iq}\beta_p$. Then the $E^{p,q}$, $1 \le p \le m$, $1 \le q \le n$, form a basis for L(V, W) and so

$$T = \sum_{p=1}^{m} \sum_{q=1}^{n} A_{pq} E^{p,q}$$
 (3.4.15.2)

for certain scalars A_{pq} (the coordinates of T in this basis for $L(V, \hat{W})$. Show that the matrix A with entries $A(p,q) = A_{pq}$ is precisely the matrix of T relative to the pair $\mathcal{B}, \mathcal{B}'$.

Solution: Given,

$$T = \sum_{n=1}^{m} \sum_{q=1}^{n} A_{pq} E^{p,q}$$
 (3.4.15.3)

where

$$E^{p,q}(\alpha_i) = \begin{cases} \beta_p & p = i \\ 0 & \text{otherwise} \end{cases}$$
 (3.4.15.4)

where $\mathcal{B} = \{\alpha_1, \dots, \alpha_n\}$ is basis of V and 3.5.1. In R^3 , let $\alpha_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$, $\alpha_2 = \begin{pmatrix} 0 \\ 1 \\ -2 \end{pmatrix}$ and $\alpha_3 = \begin{pmatrix} -1 \\ -1 \\ 0 \end{pmatrix}$.

Consider a vector $\mathbf{x} = \{x_1, x_2, \dots, x_n\} \in V$,

$$\mathbf{x} = \sum_{q=1}^{n} x_q \alpha_q$$
 (3.4.15.6)

$$\therefore E^{p,q}(\mathbf{x}) = \sum_{q=1}^{n} x_q E^{p,q}(\alpha_q)$$
 (3.4.15.7)

$$= x_q \delta_{iq} \beta_p \tag{3.4.15.8}$$

Consider $T(\mathbf{x})$, from (3.4.15.3)

$$T(\mathbf{x}) = \sum_{p=1}^{m} \sum_{q=1}^{n} A_{pq} E^{p,q}(\mathbf{x})$$
 (3.4.15.9)

Substitute (3.4.15.8) in (3.4.15.9)

$$T(\mathbf{x}) = \sum_{p=1}^{m} \sum_{q=1}^{n} A_{pq} x_p \delta_{iq} \beta_q$$
 (3.4.15.10)

From (3.4.15.5), $\delta_{iq}\beta_q$ is the transformation of basis of V to W. Hence $T: V \to W$ is

$$T = \begin{pmatrix} \sum_{p=1}^{n} A_{p1} x_{p} \\ \sum_{p=1}^{n} A_{p2} x_{p} \\ \vdots \\ \sum_{p=1}^{n} A_{pm} x_{p} \end{pmatrix}$$
(3.4.15.11)

$$T = \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & A_{22} & \dots & A_{2n} \\ \vdots & \vdots & \dots & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$
 (3.4.15.12)

 \therefore We showed that the matrix A with entries $A(p,q) = A_{pq}$ is precisely the matrix of T relative to the pair $\mathcal{B}, \mathcal{B}'$.

3.5 Linear Functionals

a) if f is a linear functional on \mathbb{R}^3 such that $f(\alpha_1) = 1$, $f(\alpha_2) = -1$, $f(\alpha_3) = 3$, And if $\alpha = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$, find $f(\alpha)$.

Solution: Given, $\alpha = \begin{pmatrix} a \\ b \end{pmatrix}$. Let,

$$\mathbf{A} = \begin{pmatrix} \alpha_1 & \alpha_2 & \alpha_3 \end{pmatrix} \qquad (3.5.1.1)$$
$$\mathbf{AX} = \alpha \qquad (3.5.1.2)$$

$$\mathbf{AX} = \alpha \qquad (3.5.1.2)$$

$$\begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \\ 1 & -2 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$
 (3.5.1.3)

 $\mathbf{X} = A^{-1}\alpha$ will give solution of the equation.

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 1 & -2 & 0 & 0 & 0 & 1 \end{pmatrix} \xrightarrow{R_3 \leftarrow R_3 - R_1} (3.5.1.4)$$

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & -2 & 1 & -1 & 0 & 1 \end{pmatrix} \xrightarrow{R_3 \leftarrow R_3 + 2R_2} (3.5.1.5)$$

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & 0 & -1 & -1 & 2 & 1 \end{pmatrix} \xrightarrow{R_3 \leftarrow R_3/(-1)}$$

$$(3.5.1.6)$$

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & -2 & -1 \end{pmatrix} \xrightarrow{R_1 \leftarrow R_1 + R_3}$$

$$(3.5.1.7)$$

$$\begin{pmatrix} 1 & 0 & 0 & 2 & -2 & -1 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & -2 & -1 \end{pmatrix} \xleftarrow{R_2 \leftarrow R_2 + R_3}$$

(3.5.1.8)

$$\begin{pmatrix}
1 & 0 & 0 & 2 & -2 & -1 \\
0 & 1 & 0 & 1 & -1 & -1 \\
0 & 0 & 1 & 1 & -2 & -1
\end{pmatrix}$$
(3.5.1.9)

Thus,

$$A^{-1} = \begin{pmatrix} 2 & -2 & -1 \\ 1 & -1 & -1 \\ 1 & -2 & -1 \end{pmatrix}$$
 (3.5.1.10)

$$\mathbf{X} = A^{-1}\alpha = \begin{pmatrix} 2 & -2 & -1 \\ 1 & -1 & -1 \\ 1 & -2 & -1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$
 (3.5.1.11)

Given, f is a linear functional on \mathbb{R}^3 ,

$$\alpha = \alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 \tag{3.5.1.12}$$

$$\implies f(\alpha) = \mathbf{X}^T \begin{pmatrix} f(\alpha_1) \\ f(\alpha_2) \\ f(\alpha_3) \end{pmatrix}$$
 (3.5.1.13)

Given, $f(\alpha_1) = 1$, $f(\alpha_2) = -1$ and $f(\alpha_3) = 3$.

$$f(\alpha) = \mathbf{X}^T \begin{pmatrix} 1 \\ -1 \\ 3 \end{pmatrix} \tag{3.5.1.14}$$

$$\implies f(\alpha) = \begin{pmatrix} a \\ b \\ c \end{pmatrix}^{T} \begin{pmatrix} 2 & 1 & 1 \\ -2 & -1 & -2 \\ -1 & -1 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 3 \end{pmatrix}$$
(3.5.1.15)

$$f(\alpha) = \begin{pmatrix} a \\ b \\ c \end{pmatrix}^T \begin{pmatrix} 4 \\ -7 \\ -3 \end{pmatrix}$$
 (3.5.1.16)

Hence,

$$f(\alpha) = 4a - 7b - 3c \tag{3.5.1.17}$$

b) Describe explicitly a linear functional f on R^3 such that $f(\alpha_1) = f(\alpha_2) = 0$ but $f(\alpha_3) \neq$

Solution: Let us consider $\alpha = \begin{bmatrix} a \\ b \end{bmatrix}$ such that

$$\alpha_1 x_1 + \alpha_2 x_2 + \alpha_2 x_2 = \alpha \tag{3.5.1.18}$$

$$\alpha_1 x_1 + \alpha_2 x_2 + \alpha_2 x_2 = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$
 (3.5.1.19)

The coefficient matrix is:

$$A = \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \\ 1 & -2 & 0 \end{pmatrix} \tag{3.5.1.20}$$

So,

$$A\mathbf{x} = \alpha \tag{3.5.1.21}$$

$$\implies x = A^{-1}\alpha \tag{3.5.1.22}$$

Now to get A^{-1} , we will consider Gauss-Jordon theorem. So, we will take (A|I), where I is a 3×3 identity matrix.

$$\begin{pmatrix}
1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & -1 & 0 & 1 & 0 \\
1 & -2 & 0 & 0 & 0 & 1
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 - R_1}$$

$$\begin{pmatrix}
1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & -1 & 0 & 1 & 0 \\
0 & -2 & 1 & -1 & 0 & 1
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 + 2R_2}$$

$$\begin{pmatrix}
1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & -1 & 0 & 1 & 0 \\
0 & 0 & -1 & -1 & 2 & 1
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 + 2R_2}$$

$$\begin{pmatrix}
1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & -1 & 0 & 1 & 0 \\
0 & 0 & -1 & -1 & 2 & 1
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 / (-1)}$$

$$\begin{pmatrix}
1 & 0 & -1 & 1 & 0 & 0 \\
0 & 1 & -1 & 0 & 1 & 0 \\
0 & 0 & 1 & 1 & -2 & -1
\end{pmatrix}
\xrightarrow{R_2 \leftarrow R_2 + R_3}
\xrightarrow{R_1 \leftarrow R_1 + R_3}$$

$$\begin{pmatrix}
1 & 0 & 0 & 2 & -2 & -1 \\
0 & 1 & 0 & 1 & -1 & -1 \\
0 & 0 & 1 & 1 & -2 & -1
\end{pmatrix}$$
(3.5.1.24)

Now, we can say that

$$A^{-1} = \begin{pmatrix} 2 & -2 & -1 \\ 1 & -1 & -1 \\ 1 & -2 & -1 \end{pmatrix}$$
 (3.5.1.25)

 $\mathbf{x} = A^{-1}\alpha$

(3.5.1.26)

As

$$\Rightarrow \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = A^{-1} \begin{pmatrix} a \\ b \\ c \end{pmatrix} \qquad (3.5.1.27)$$

$$\Rightarrow \mathbf{x} = \begin{pmatrix} 2 & -2 & -1 \\ 1 & -1 & -1 \\ 1 & -2 & -1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} \qquad (3.5.1.28)$$

Now, as f is a linear functional on R^3 ,

$$\alpha = \alpha_1 x_1 + \alpha_2 x_2 + \alpha_2 x_2$$

$$(3.5.1.29)$$

$$\Rightarrow f(\alpha) = f(\alpha_1 x_1 + \alpha_2 x_2 + \alpha_2 x_2)$$

$$(3.5.1.30)$$

$$\Rightarrow f(\alpha) = x_1 f(\alpha_1) + x_2 f(\alpha_2) + x_3 f(\alpha_3)$$

$$(3.5.1.31)$$

$$\Rightarrow f(\alpha) = \begin{pmatrix} x_1 & x_2 & x_3 \end{pmatrix} \begin{pmatrix} f(\alpha_1) \\ f(\alpha_2) \\ f(\alpha_3) \end{pmatrix}$$

$$(3.5.1.32)$$

$$\Rightarrow f(\alpha) = \mathbf{x}^T \begin{pmatrix} f(\alpha_1) \\ f(\alpha_2) \\ f(\alpha_3) \end{pmatrix}$$

$$(3.5.1.33)$$

As mentioned in the problem statement, $f(\alpha_1) = f(\alpha_2) = 0$ and $f(\alpha_3) \neq 0$. Now,

$$f(\alpha) = \mathbf{x}^{T} \begin{pmatrix} f(\alpha_{1}) \\ f(\alpha_{2}) \\ f(\alpha_{3}) \end{pmatrix}$$

$$(3.5.1.34)$$

$$\Rightarrow f(\alpha) = \mathbf{x}^{T} \begin{pmatrix} 0 \\ 0 \\ f(\alpha_{3}) \end{pmatrix}$$

$$(3.5.1.35)$$

$$\Rightarrow f(\alpha) = \begin{pmatrix} a \\ b \\ c \end{pmatrix}^{T} \begin{pmatrix} 2 & 1 & 1 \\ -2 & -1 & -2 \\ -1 & -1 & -1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ f(\alpha_{3}) \end{pmatrix}$$

$$(3.5.1.36)$$

$$\Rightarrow f(\alpha) = \begin{pmatrix} a \\ b \\ c \end{pmatrix}^{T} \begin{pmatrix} f(\alpha_{3}) \\ -2f(\alpha_{3}) \\ -f(\alpha_{3}) \end{pmatrix}$$

$$(3.5.1.37)$$

$$\Rightarrow f(a, b, c) = f(\alpha_{3}) \begin{pmatrix} a \\ b \\ c \end{pmatrix}^{T} \begin{pmatrix} 1 \\ -2 \\ -1 \end{pmatrix}$$

$$(3.5.1.38)$$

So, the function can be defined as:

$$f(a,b,c) = f(\alpha_3) \begin{pmatrix} a \\ b \\ c \end{pmatrix}^T \begin{pmatrix} 1 \\ -2 \\ -1 \end{pmatrix}$$
 (3.5.1.39)

$$\mathbf{or}, f(\alpha) = f(\alpha_3)(a - 2b - c)$$
 (3.5.1.40)

3.5.2. Let $\mathbf{B} = \{\alpha_1, \alpha_2, \alpha_3\}$ be the basis for \mathbf{C}^3 defined by

$$\alpha_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \alpha_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \alpha_3 = \begin{pmatrix} 2 \\ 2 \\ 0 \end{pmatrix}$$
 (3.5.2.1)

Find the dual basis of **B**.

Solution: Let $\{f_1, f_2, f_3\}$ be the dual basis of **B** such that,

$$f_i(\alpha_j) = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$$
 (3.5.2.2)

and

$$f_i(\alpha_1, \alpha_2, \alpha_3) = \sum_{i=1}^3 \delta_{ij} \alpha_j$$
 (3.5.2.3)

Given,

$$\mathbf{B} = \begin{pmatrix} 1 & 1 & 2 \\ 0 & 1 & 2 \\ -1 & 1 & 0 \end{pmatrix} \tag{3.5.2.4}$$

Then,

$$\mathbf{B}^{-1} = \begin{pmatrix} 1 & -1 & 0 \\ 1 & -1 & 1 \\ -\frac{1}{2} & 1 & -\frac{1}{2} \end{pmatrix}$$
 (3.5.2.5)

So the elements of dual basis are rows of matrix \mathbf{B}^{-1} . Therefore we get,

$$(\delta_{11}, \delta_{12}, \delta_{13}) = (1, -1, 0) \tag{3.5.2.6}$$

$$(\delta_{21}, \delta_{22}, \delta_{23}) = (1, -1, 1)$$
 (3.5.2.7)

$$(\delta_{31}, \delta_{32}, \delta_{33}) = (-\frac{1}{2}, 1, -\frac{1}{2})$$
 (3.5.2.8)

Using (3.5.2.3), we get

Solution:

$$\begin{pmatrix} f_1 \\ f_2 \\ f_3 \end{pmatrix} = \begin{pmatrix} 1 & -1 & 0 \\ 1 & -1 & 1 \\ -\frac{1}{2} & 1 & -\frac{1}{2} \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix}$$
(3.5.2.9)

$$= \begin{pmatrix} \alpha_1 - \alpha_2 \\ \alpha_1 - \alpha_2 + \alpha_3 \\ -\frac{1}{2}\alpha_1 + \alpha_2 - \frac{1}{2}\alpha_3 \end{pmatrix}$$
 (3.5.2.10)

Hence, $\{f_1, f_2, f_3\}$ is the required dual basis for

3.5.3. If **A** and **B** are $n \times n$ matrices over the field **F**, show that trace(AB) = trace(BA). Now show that similar matrices have the same trace.

$$trace(\mathbf{AB}) = \sum_{i=1}^{n} (\mathbf{AB})_{ii}$$
 (3.5.3.1)

$$= \sum_{i=1}^{n} \sum_{k=1}^{n} \mathbf{A}_{ik} \mathbf{B}_{ki}$$
 (3.5.3.2)

$$= \sum_{i=1}^{n} \sum_{k=1}^{n} \mathbf{B}_{ki} \mathbf{A}_{ik}$$
 (3.5.3.3)

$$= \sum_{i=1}^{n} (\mathbf{BA})_{kk}$$
 (3.5.3.4)

$$= trace(\mathbf{BA}) \tag{3.5.3.5}$$

Hence proved trace(AB) = trace(BA). Let A and **B** be similar matrices then \exists **S** such that,

$$\mathbf{A} = \mathbf{S}^{-1}\mathbf{B}\mathbf{S} \tag{3.5.3.6}$$

Taking trace of A we get,

$$trace(\mathbf{A}) = trace(\mathbf{S}^{-1}\mathbf{B}\mathbf{S}) = trace(\mathbf{S}^{-1}(\mathbf{B}\mathbf{S}))$$
(3.5.3.7)

Using (3.5.3.5) in (3.5.3.7)

$$trace(\mathbf{A}) = trace(\mathbf{S}^{-1}\mathbf{S}\mathbf{B}) \tag{3.5.3.8}$$

$$= trace((\mathbf{S}^{-1}\mathbf{S})\mathbf{B}) \qquad (3.5.3.9)$$

$$= trace(\mathbf{IB}) \qquad (3.5.3.10)$$

$$= trace(\mathbf{B}) \tag{3.5.3.11}$$

Hence Proved

(3.5.2.6) Hence Proved (3.5.2.7) 3.5.4. Let **V** be the vector space of all polynomial functions p from R into R which have degree 2 or less:

$$p(x) = c_0 + c_1 x + c_2 x^2$$

Define three linear functionals on V by

$$f_1(p) = \int_0^1 p(x) dx; \ f_2(p) = \int_0^2 p(x) dx;$$
$$f_3(p) = \int_0^{-1} p(x) dx$$

Show that $\{f_1, f_2, f_3\}$ is a basis for \mathbf{V}^* by exhibiting the basis for V of which it is the dual.

Solution: Given the basis **F** and corresponding dual basis G, the defining property of the dual basis states that:

$$\mathbf{G}^{T}\mathbf{F} = \mathbf{I}$$

$$\implies \mathbf{G} = (\mathbf{F}^{-1})^{T}$$
(3.5.4.1)

Let the indexed vector sets,

$$\mathbf{V} = \{f_1, f_2, f_3\}; \ \mathbf{V}^* = \{\alpha_1, \alpha_2, \alpha_3\}$$

1. Let,

$$\mathbf{p}' = \mathbf{c}^T \mathbf{x} \tag{3.5.4.2}$$

where,

$$\mathbf{c} = \begin{pmatrix} c_0 \\ c_1 \\ c_2 \end{pmatrix}; \quad \mathbf{x} = \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix}$$

2. Representing the functionals as vector,

$$\mathbf{f} = \begin{pmatrix} f_1 \\ f_2 \\ f_3 \end{pmatrix} \tag{3.5.4.3}$$

3. Representing the integrations as vector,

$$\mathbf{I} = \begin{pmatrix} \int_0^1 dx \\ \int_0^2 dx \\ \int_0^{-1} dx \end{pmatrix}$$
 (3.5.4.4)

4. So,

$$\mathbf{f} = \mathbf{I}\mathbf{c}^T\mathbf{x} = \mathbf{I}\mathbf{p}' \tag{3.5.4.5}$$

(3.5.4.5) can written in matrix format as,

$$\mathbf{f} = \mathbf{Pc} \tag{3.5.4.6}$$

where,

$$\mathbf{P} = \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{3} \\ 2 & 2 & \frac{8}{3} \\ -1 & \frac{1}{2} & \frac{-1}{2} \end{pmatrix}$$
(3.5.4.7)

5. **P** is one-one if it has a inverse. Calculating the determinant of **P**,

$$\implies |P| = -2 \tag{3.5.4.8}$$

From, (3.5.4.8), **P** is one-one. Also,

$$\mathbf{V} = \mathbf{f}^T = \begin{pmatrix} f_1 & f_2 & f_3 \end{pmatrix} \tag{3.5.4.9}$$

From (3.5.4.6), (3.5.4.8) and (3.5.4.9), the rows of **P** are isomorphic to **V**. So, finding the dual

basis by performing matrix operations on \mathbf{P}^T

$$\begin{pmatrix} 1 & 2 & -1 & 1 & 0 & 0 \\ \frac{1}{2} & 2 & \frac{1}{2} & 0 & 1 & 0 \\ \frac{1}{3} & \frac{8}{3} & \frac{-1}{3} & 0 & 0 & 1 \end{pmatrix} \xrightarrow{R_{3} \leftarrow R_{3} - \frac{R_{1}}{3}} \xrightarrow{R_{2} \leftarrow R_{2} - \frac{R_{1}}{2}}$$

$$\begin{pmatrix} 1 & 2 & -1 & 1 & 0 & 0 \\ 0 & 1 & 1 & \frac{-1}{2} & 1 & 0 \\ 0 & 2 & 0 & \frac{-1}{3} & 0 & 1 \end{pmatrix} \xleftarrow{R_{3} \leftarrow \frac{R_{3} - 2R_{2}}{-2}} \xrightarrow{R_{3} \leftarrow \frac{R_{3} - 2R_{2}}{-2}}$$

$$\begin{pmatrix} 1 & 2 & -1 & 1 & 0 & 0 \\ 0 & 1 & 1 & \frac{-1}{2} & 1 & 0 \\ 0 & 0 & 1 & \frac{-1}{3} & 1 & \frac{-1}{2} \end{pmatrix} \xrightarrow{R_1 \leftarrow R_1 + R_3} \xrightarrow{R_2 \leftarrow R_2 - R_3}$$

$$\begin{pmatrix} 1 & 2 & 0 & \frac{2}{3} & 1 & \frac{-1}{2} \\ 0 & 1 & 0 & \frac{-1}{6} & 0 & \frac{1}{2} \\ 0 & 0 & 1 & \frac{-1}{3} & 1 & \frac{-1}{2} \end{pmatrix} \xrightarrow{R_1 \leftarrow R_1 - 2R_2}$$

$$\begin{pmatrix}
1 & 0 & 0 & 1 & 1 & -\frac{3}{2} \\
0 & 1 & 0 & -\frac{1}{6} & 0 & \frac{1}{2} \\
0 & 0 & 1 & -\frac{1}{3} & 1 & -\frac{1}{2}
\end{pmatrix}$$
(3.5.4.10)

From (3.5.4.10), the dual of $V \implies V^*$ can be written in matrix form as,

$$\mathbf{V}^* = \mathbf{A}\mathbf{x} \tag{3.5.4.11}$$

where,

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & \frac{-3}{2} \\ \frac{-1}{6} & 0 & \frac{1}{2} \\ \frac{-1}{3} & 1 & \frac{-1}{2} \end{pmatrix}; \quad \mathbf{x} = \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix}$$

3.5.5. If **A** and **B** are $n \times n$ matrices, show that

$$\mathbf{AB} - \mathbf{BA} = \mathbf{I} \tag{3.5.5.1}$$

is impossible.

Solution:

$$tr(\mathbf{AB}) = tr(\mathbf{BA}) \tag{3.5.5.2}$$

$$tr(\mathbf{I}) = \sum_{i=1}^{n} I_{jj}$$
 (3.5.5.3)

$$\implies \sum_{i=1}^{n} 1 = n \tag{3.5.5.4}$$

Taking tr on both sides (3.5.5.1)

$$tr(\mathbf{AB} - \mathbf{BA}) = tr(\mathbf{I}) \qquad (3.5.5.5)$$

$$\implies tr(\mathbf{AB}) - tr(\mathbf{BA}) = tr(\mathbf{I})$$
 (3.5.5.6)

From (3.5.5.2) and (3.5.5.4)

$$tr(\mathbf{AB}) - tr(\mathbf{BA}) = 0 \tag{3.5.5.7}$$

$$tr(\mathbf{I}) = n \tag{3.5.5.8}$$

$$= 0 \neq n$$
 (3.5.5.9)

Thus:-

$$tr(\mathbf{AB} - \mathbf{BA}) \neq tr(\mathbf{I}) \tag{3.5.5.10}$$

hence no solution possible.

3.5.6. Let m and n be positive integers and field **F**.Let f_1, \ldots, f_m be linear functions in F^n .For α in F^n define

$$T\alpha = (f_1(\alpha), \dots, f_m(\alpha)) \tag{3.5.6.1}$$

show that T is a linear transformation from F^n into F^m . Then show that every linear transformation from F^n into F^m is of the above form ,for some f_1, \ldots, f_m .

Solution: Let $\mathbf{b}, \alpha \in F^n$ and a is a scalar

$$T(a\alpha + \mathbf{b}) = (f_1(a\alpha + \mathbf{b}), \dots, f_m(a\alpha + \mathbf{b}))$$

$$= (af_1(\alpha) + f_1(\mathbf{b}), \dots, af_m(\alpha) + f_m(\mathbf{b}))$$

$$= a(f_1(\alpha), \dots, f_m(\alpha)) + (f_1(\mathbf{b}), \dots, f_m(\mathbf{b}))$$
(3.5.6.2)

The equation (3.5.6.2) can be written as

$$T(a\alpha + \mathbf{b}) = aT(\alpha) + T(\mathbf{b}) \tag{3.5.6.3}$$

So, T is a linear transformation.

Let the matrix A of order $m \times n$ represent any linear transformation $\mathbf{X} \mapsto A\mathbf{X}$ from F^n into F^m . For $i=1,\ldots,m$, let

$$f_i(x_1, \dots, x_n) = \sum_{i=1}^n A_{ij} x_j$$
 (3.5.6.4)

The transformation into F^m , AX can be written as

$$(f_1(\mathbf{X}),\ldots,f_m(\mathbf{X})) \tag{3.5.6.5}$$

This is of the form (3.5.6.1)

3.5.7. Let $\alpha_1 = (1, 0, -1, 2)$ and $\alpha_2 = (2, 3, 1, 1)$ and let **W** be the subspace of \mathbb{R}^4 spanned by α_1 and α_2 . Which linear functionals **f**:

$$f(x_1, x_2, x_3, x_4) = c_1 x_1 + c_2 x_2 + c_3 x_3 + c_4 x_4$$
(3.5.7.1)

are in the annihilator of W?

Solution: (3.5.7.1), can be expressed as

Linear Functional	If V is a vector space over the field F , field F is also called a linear functional
Dual Space of V	If V is a vector space, the collection of It is the space $L(V, F)$ which we denote
Annihilator of S	${\bf V}$ is a vector space over the field ${\bf F}$ and S^o of linear functionals ${\bf f}$ on ${\bf V}$ such that

$$\mathbf{f}(\mathbf{x}) = \mathbf{x}^T \mathbf{c} \tag{3.5.7.2}$$

where
$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix}$$
 and $\mathbf{c} = \begin{pmatrix} c_1 \\ c_2 \\ c_3 \\ c_4 \end{pmatrix}$ (3.5.7.3)

Given two vectors

$$\alpha_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \\ 2 \end{pmatrix} \quad and \quad \alpha_2 = \begin{pmatrix} 2 \\ 3 \\ 1 \\ 1 \end{pmatrix} \tag{3.5.7.4}$$

Since, α_1 is not a scalar multiple of α_2 . Thus, α_1 and α_2 are linearly independent. Also, given α_1 and α_2 span W, thus $\{\alpha_1, \alpha_2\}$ form basis for W. Hence, W has dimension 2.

Now, a functional \mathbf{f} is in the annihilator of \mathbf{W} if and only if $\mathbf{f}(\alpha_1) = \mathbf{f}(\alpha_2) = 0$. We find such \mathbf{f} by solving the system

$$\mathbf{f}(\alpha_1) = \mathbf{f}(\alpha_2) = 0 \tag{3.5.7.5}$$

From, (3.5.7.2) and (3.5.7.5), we have

$$\begin{pmatrix} 1 & 0 & -1 & 2 \\ 2 & 3 & 1 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \\ c_4 \end{pmatrix} = 0$$
 (3.5.7.6)

Converting (3.5.7.6) into row reduced echelon form

$$\begin{pmatrix} 1 & 0 & -1 & 2 \\ 2 & 3 & 1 & 1 \end{pmatrix} \stackrel{r_2=r_2-2r_1}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -1 & 2 \\ 0 & 3 & 3 & -3 \end{pmatrix}$$

$$(3.5.7.7)$$

$$\begin{pmatrix} 1 & 0 & -1 & 2 \\ 0 & 3 & 3 & -3 \end{pmatrix} \stackrel{r_2=\frac{r_2}{3}}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -1 & 2 \\ 0 & 1 & 1 & -1 \end{pmatrix}$$

$$(3.5.7.8)$$

from (3.5.7.8), we have

$$\begin{pmatrix} 1 & 0 & -1 & 2 \\ 0 & 1 & 1 & -1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \\ c_4 \end{pmatrix} = 0$$
 (3.5.7.9)

The general element of \mathbf{W}^o is therefore

$$\mathbf{f}(\mathbf{x}) = \mathbf{x}^T \mathbf{a} \tag{3.5.7.10}$$

where

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} \quad and \quad \mathbf{a} = \begin{pmatrix} c_3 - 2c_4 \\ -c_3 + c_4 \\ c_3 \\ c_4 \end{pmatrix} \quad (3.5.7.11)$$

for arbitrary constants c_3 and c_4 .

Also, W^o has dimension 2.

3.5.8. Let **W** be the subspace of \mathbb{R}^5 which is spanned by the vectors

$$\alpha_1 = \epsilon_1 + 2\epsilon_2 + \epsilon_3,$$

$$\alpha_2 = \epsilon_2 + 3\epsilon_3 + 3\epsilon_4 + \epsilon_5,$$

$$\alpha_3 = \epsilon_1 + 4\epsilon_2 + 6\epsilon_3 + 4\epsilon_4 + \epsilon_5$$
(3.5.8.1)

Find a basis for W⁰

Solution: If **V** is a vector space over the field \mathbb{F} and **W** is a subset of **V**, the annihilator of **W** is the set \mathbf{W}^0 of linear functionals **f** on **V** such that $\mathbf{f}(\alpha) = 0$ for every α in **W**.

Properties of Annihilator: If f is a linear functional on \mathbb{R}^n :

$$f(x_1, x_2, \dots, x_n) = \sum_{j=1}^{n} c_j x_j$$
 (3.5.8.2)

Then f is in W^0 if and only if

$$\forall \alpha \in W : f(\alpha) = 0 \iff f = 0 \quad (3.5.8.3)$$

(3.5.8.2) can be expressed as:

$$\mathbf{f}(\mathbf{x}) = \mathbf{x}^T \mathbf{c} \qquad (3.5.8.4)$$

where
$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 and $\mathbf{c} = \begin{pmatrix} c_1 \\ c_2 \\ c_3 \\ c_4 \\ c_5 \end{pmatrix}$ (3.5.8.5)

Given three vectors

$$\alpha_1 = \begin{pmatrix} 1\\2\\1\\0\\0 \end{pmatrix}, \quad \alpha_2 = \begin{pmatrix} 0\\1\\3\\3\\1 \end{pmatrix}, \alpha_3 = \begin{pmatrix} 1\\4\\6\\4\\1 \end{pmatrix} \quad (3.5.8.6)$$

Let matrix A with column vectors $\alpha_1, \alpha_2, \alpha_3$:

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 4 \\ 1 & 3 & 6 \\ 0 & 3 & 4 \\ 0 & 1 & 1 \end{pmatrix} \tag{3.5.8.7}$$

Given that f is a linear functional on \mathbb{R}^5 , then f is in W^0 if and only if,

$$f(\alpha_i) = 0, i = 1, 2, 3$$
 (3.5.8.8)

$$\implies \mathbf{A}^T \mathbf{c} = \mathbf{0} \tag{3.5.8.9}$$

Converting the (3.5.8.9) into system of equations, we have,

$$\begin{pmatrix} 1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 3 & 3 & 1 \\ 1 & 4 & 6 & 4 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \\ c_4 \\ c_5 \end{pmatrix} = 0$$
 (3.5.8.10)

Converting (3.5.8.10) into row reduced echelon form,

$$\begin{pmatrix} 1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 3 & 3 & 1 \\ 1 & 4 & 6 & 4 & 1 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 1 & 0 & 0 & 4 & 3 \\ 0 & 1 & 0 & -3 & -2 \\ 0 & 0 & 1 & 2 & 1 \end{pmatrix}$$
(3.5.8.11)

From (3.5.8.11), we have,

$$c_1 = -(4c_4 + 3c_5) \tag{3.5.8.12}$$

$$c_2 = (3c_4 + 2c_5) \tag{3.5.8.13}$$

$$c_3 = -(2c_4 + c_5) \tag{3.5.8.14}$$

Therefore, general element of W^0 is therefore,

$$f(x_1, \dots, x_5) = -(4c_4 + 3c_5)x_1 + (3_4 + 2c_5)x_2$$
$$-(2c_4 + c_5)x_3 + c_4x_4 + c_5x_5$$
$$(3.5.8.15)$$

Therefore, dimension of \mathbf{W}^0 is 2 and a basis $\{f_1, f_2\}$ can be obtained by putting $c_4 = 0, c_5 =$

1 and
$$c_4 = 1, c_5 = 0$$
 in (3.5.8.15)

$$f_1(x_1,\ldots,x_5) = -3x_1 + 2x_2 - x_3 + x_5$$

(3.5.8.16)

$$f_2(x_1, \dots, x_5) = -4x_1 + 3x_2 - 2x_3 + x_4$$
(3.5.8.17)

3.5.9. Let \mathbb{V} be the vector space of all 2×2 matrices over the field of real numbers, and let

$$\mathbf{B} = \begin{pmatrix} 2 & -2 \\ -1 & 1 \end{pmatrix} \tag{3.5.9.1}$$

Let \mathbb{W} be the subspace of \mathbb{V} consisting of all \mathbf{A} such that AB = 0. Let f be a linear functional on \mathbb{V} which is in the annihilator of \mathbb{W} . Suppose that $f(\mathbf{I}) = 0$ and $f(\mathbf{C}) = 3$, where **I** is the 2×2 identity matrix and

$$\mathbf{C} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.5.9.2}$$

Find $f(\mathbf{B})$

V is of the form.

$$f(\mathbf{A}) = aA_{11} + bA_{12} + cA_{21} + dA_{22} \quad (3.5.9.3)$$

for a,b,c,d \in **R** Let **A** \in **W** be,

$$\mathbf{A} = \begin{pmatrix} p & q \\ q & s \end{pmatrix} \tag{3.5.9.4}$$

$$\therefore \mathbf{AB} = 0 \qquad (3.5.9.5)$$

$$\implies \begin{pmatrix} p & q \\ q & s \end{pmatrix} \begin{pmatrix} 2 & -2 \\ -1 & 1 \end{pmatrix} = 0 \qquad (3.5.9.6)$$

$$\implies \begin{pmatrix} 2p - q & -2p + q \\ 2q - s & -2q + s \end{pmatrix} = 0 \qquad (3.5.9.7)$$

 \therefore q=2p and s=2q. Hence **W** consists of all matrices of the form

$$\begin{pmatrix} p & 2p \\ q & 2q \end{pmatrix} \tag{3.5.9.8}$$

Now V is an annihilator of W. Hence, $f \in \mathbf{W}^0$

$$\implies f\begin{pmatrix} p & 2p \\ q & 2q \end{pmatrix} = 0 \forall p, q \in \mathbf{R} \qquad (3.5.9.9)$$

from equation (3.5.9.3)

$$\implies ap + 2bp + cq + 2dq = 0 \forall p, q \in \mathbf{R}$$
(3.5.9.10)

$$\implies (a+2b)p + (c+2d)q = 0, \forall p, q \in \mathbf{R}$$
 3.5. (3.5.9.11)

Hence, $b=\frac{-1}{2}a$ and $d=\frac{-1}{2}c$. Hence general $f \in$ \mathbf{W}^0 is of the form,

$$f(\mathbf{A}) = aA_{11} - \frac{1}{2}aA_{12} + cA_{21} - \frac{1}{2}cA_{22}$$
(3.5.9.12)

Now, $f(\mathbf{C}) = 3 \implies d = 3 \implies c=-6$. Also given that, $f(\mathbf{I}) = 0 \implies a - \frac{1}{2}c = 0 \implies a = -3$. Substituting the above parameters in equation (3.5.9.12) we get,

$$\therefore f(\mathbf{A}) = -3A_{11} + \frac{3}{2}A_{12} - 6A_{21} + 3A_{22}$$
(3.5.9.13)
$$\text{Now, } f(\mathbf{B}) = f\begin{pmatrix} 2 & -2 \\ -1 & 1 \end{pmatrix}$$

$$\implies f(\mathbf{B}) = -3(2) + \frac{3}{2}(-2) - 6(-1) + 3(1) = 0$$
(3.5.9.15)

Solution: The general Linear functional f on 5.10. Let F be a subfield of the complex numbers. We define n linear functionals on $F^n(n \ge 2)$ by

$$f_k(x_1,, x_n) = \sum_{j=1}^n (k-j)x_j, 1 \le k \le n.$$
(3.5.10.1)

What is the dimension of the subspace annihilated by $f_1, f_2, ..., f_n$?

Solution: See Tables 3.5.10.1 and 3.5.10.2.

Given	F be a subfield of the co
	Definition of n linear functionals on $F^n (n \ge 2)$ by
To find	The dimension of the subspace an
f_k	$f_k(x_1,, x_n) = \sum_{j=1}^n f_k(x_1,, x_n) = k \sum_{j=1}^n$
	All f_k are linear combinations of the

TABLE 3.5.10.1

3.5.11. Let ${\bf F}$ be a subfield of the field of complex numbers and let V be any vector space over F.

Vector	Suppose that f and g are linear functionals on $\mathbf{f}_k(x_1,, x_n) = \sum_{j=1}^n (k - \mathbf{V})$ such that the function h defined by $h(\alpha) = \mathbf{f}(\alpha)g(\alpha)$ is also a linear functional on \mathbf{V} . Prove that either $f = 0$ or $g = 0$. Solution: Refer Table 3.5.11.1. 3.5.12. Let \mathbb{F} be a field of characteristic zero and let \mathbf{V} be a finite dimensional vector space over \mathbb{F} . If $\alpha_1, \alpha_2,, \alpha_m$ are finitely many each different from the zero vector, prove that there is a linear functional f on \mathbf{V} such that $f(\alpha_i) \neq 0, i = 1, 2,, m$ $f_1(\mathbf{x}) = 0.x_1 - x_2 - 2x_3 (n-1)x_n$ $f_2(\mathbf{x}) = x_1 + 0.x_2 - 1.x_3$ Solution: \mathbf{x} we will make use of the following theorem $f_n(\mathbf{x}) = (n-1)x_1 + (n-2).x_2 + + \lim_{n \to \infty} f(n) = \lim_{n$
	$(n-1) (n-2) (n-2) 0 f_i(\mathbf{y}_j) = \delta_{ij}$ (3.5.12.2)
	and V^* is the space of all linear functionals on
34.	
Matrix	where the i^{th} row is defined by where the i^{th} row is defined by $A_i = (i-1,i-2,,i \frac{w}{n})$ and finite-dimensional space to the trace of any matrix which represents the operator in a ordered basis. This is well-defined since all such representing matrices for one operator are $\begin{pmatrix} 0 & -1 & -2 & -3 \\ 1 & 0 & -1 & -2 \\ 2 & 1 & 0 & -1 \\ 3 & 2 & 1 & 0 \end{pmatrix}$ Now let V be the space of all 2×2 matrices over the field F and let P be a fixed 2×2 matrix. Let T be the linear operator on V defined by $T(A) = PA$. Prove that $tr(T) = 2tr(P)$. Solution: Given V is the space of all 2×2 matrix.
	$(a)A_i \longrightarrow (1-i)A_2 + A_i$ matrices over field F. P is a 2×2 matrix,
	$A_{i} = (0, i-2, 2(i-2), 3(i-2),, (n-1)(i-2)) \begin{pmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{pmatrix} $ (3.5.13.1)
	$A_{i} \xrightarrow{j-2} A_{i} $ $A_{i} = -A_{1}$ $(c)A_{i} \xrightarrow{\longrightarrow} A_{i} + A_{1} \text{ of V where,}$ $A_{i} = 0$ $(P21 P22)$ $P22)$ be the ordered basis
	Since, A_1 and A_2 are linearly independent $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$, $e_{12} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ (3.5.13.2)
	annihiliated = $n-2$ $e_{21} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, e_{22} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$ (3.5.13.3)

Given	Derivation	
f , g , h are linear functionals of \mathbf{V}	By contradiction, let us assume $f \neq 0$ and	$g \neq 0$. For all $\mathbf{v} \in \mathbf{V}$
	$h(\mathbf{v}) = f(\mathbf{v})g(\mathbf{v})$	(3.5.11.1)
	$h(2\mathbf{v}) = f(2\mathbf{v})g(2\mathbf{v})$	(3.5.11.2)
	$=2f(\mathbf{v})2g(\mathbf{v})$	(3.5.11.3)
	$=4f(\mathbf{v})g(\mathbf{v})$	(3.5.11.4)
	Similarly,	
	$h(2\mathbf{v}) = 2h(\mathbf{v})$	(3.5.11.5)
	$=2f(\mathbf{v})g(\mathbf{v})$	(3.5.11.6)
	From equation (3.5.11.4) and (3.5.11.6),	
	$\implies 4f(\mathbf{v})g(\mathbf{v}) = 2f(\mathbf{v})g(\mathbf{v})$	(3.5.11.7)
	$\implies f(\mathbf{v}).g(\mathbf{v}) = 0$	(3.5.11.8)
Choosing Basis	Let B be a basis for V . Let,	
	$\mathbf{B_1} = \{ \mathbf{b} \in \mathbf{B} \mid f(\mathbf{b}) = 0 \},$	(3.5.11.9)
	$\mathbf{B_2} = \{ \mathbf{b} \in \mathbf{B} \mid g(\mathbf{b}) = 0 \}$	(3.5.11.10)
	Since,	
	$f(\mathbf{b}).g(\mathbf{b}) = 0 \forall \mathbf{b} \in \mathbf{B}$	(3.5.11.11)
	$\implies f(\mathbf{b}) = 0 \text{ or } g(\mathbf{b}) = 0$	(3.5.11.12)
	$\implies \mathbf{b} \in \mathbf{B_1} \text{ or } \mathbf{b} \in \mathbf{B_2}$	(3.5.11.13)
Choosing b ₁ and b ₂ from basis	s Let us choose $\mathbf{b_1} \in \mathbf{B_1} - \mathbf{B_2}$ and $\mathbf{b_2} \in \mathbf{B_2} - \mathbf{B_1}$ $\implies f(\mathbf{b_2}) \neq 0$ and $g(\mathbf{b_1}) \neq 0$	
	$f(\mathbf{b_1} + \mathbf{b_2}).g(\mathbf{b_1} + \mathbf{b_2}) = (f(\mathbf{b_1}) + f(\mathbf{b_2})).($	$(g(\mathbf{b_1}) + g(\mathbf{b_2}))$ (3.5.11.14)
	$= f(\mathbf{b_1}).g(\mathbf{b_1}) + f(\mathbf{b_1}).g(\mathbf{b_2}) + f(\mathbf{b_2}).g(\mathbf{b_1})$	(3.5.11.15)
	= 0 + 0 + f	$f(\mathbf{b_2}).g(\mathbf{b_1}) + 0$ (3.5.11.16)
	=f	$f(\mathbf{b_2}).g(\mathbf{b_1}) \neq 0$ (3.5.11.17)
	Equation (3.5.11.17) is contradiction to th $\implies \boxed{f = 0 \text{ or } g = 0}$	e fact that $f(\mathbf{v}).g(\mathbf{v}) = 0$.

TABLE 3.5.11.1: Expanation

The matrix representation of linear functional

Given,
$$T(A) = PA$$
,

$$T(e_{11}) = Pe_{11} = \begin{pmatrix} p_{11} & 0 \\ p_{21} & 0 \end{pmatrix}$$

$$= p_{11}e_{11} + p_{21}e_{21}$$

$$T(e_{12}) = Pe_{12} = \begin{pmatrix} 0 & p_{11} \\ 0 & p_{21} \end{pmatrix}$$

$$(3.5.13.6)$$

PARAMETERS	DESCRIPTION
F	Field
V	Finite dimensional vector space over F
$\alpha_1, \alpha_2, \ldots, \alpha_m$	non zero vectors in V
$f: \mathbf{V} \to \mathbb{F}$	Linear functional on V

TABLE 3.5.12.1: Input Parameters

T in the ordered basis \mathcal{B} is given as,

$$T = ([T(e_{11})]_{\mathcal{B}} [T(e_{12})]_{\mathcal{B}} [T(e_{21})]_{\mathcal{B}} [T(e_{22})]_{\mathcal{B}})$$
(3.5.13.12)

$$T = \begin{pmatrix} p_{11} & 0 & p_{12} & 0 \\ 0 & p_{11} & 0 & p_{12} \\ p_{21} & 0 & p_{22} & 0 \\ 0 & p_{21} & 0 & p_{22} \end{pmatrix}$$

$$= \begin{pmatrix} p_{11} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} & p_{12} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} & p_{12} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} & such that \mathbf{x}_1 + \mathbf{x}_2 + \mathbf{x}_3 + \cdots + \mathbf{x}_n = 0.$$

$$= \begin{pmatrix} p_{11} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} & p_{12} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} & such that \mathbf{x}_1 + \mathbf{x}_2 + \mathbf{x}_3 + \cdots + \mathbf{x}_n = 0.$$
a) Prove that \mathbf{W}^0 consists of all linear functional f of the form
$$= \begin{pmatrix} p_{11} I & p_{12} I \\ p_{21} I & p_{22} I \end{pmatrix} = P \otimes I \qquad (3.5.13.15)$$

$$f(\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n) = c \sum_{j=1}^{n} \mathbf{x}_j \quad (3.6.1.1)$$

Now by using the property of kronecker product we get,

$$tr(T) = tr(P \otimes I) = tr(P)tr(I)$$
 (3.5.13.16)
= $2tr(P)$ (3.5.13.17)
 $\therefore tr(T) = 2tr(P)$ (3.5.13.18)

3.5.14. Show that the trace functional on $n \times n$ matrices is unique in the following sense. If W is the space of $n \times n$ matrices over the field F and if f is a linear functional on W such that f(AB) =f(BA) for each A and B in W, then f is a scalar multiple of the trace function. If, in addition, f(I) = n, then f is the trace function.

Solution: See Tables 3.5.14.1 and 3.5.14.2

3.5.15. Let **W** be the space of $n \times n$ matrices over the field \mathbf{F} , and let \mathbf{W}_0 be the subspace spanned by the matrices C of the form C = AB - BA. Prove that W_0 is exactly the subspace of matrices which have trace zero.

> **Solution:** Let there be two subspaces defined as

$$\mathbf{W}_0 = \{ A \in \mathbf{W} : trace(A) = 0 \}$$
 (3.5.15.1)

and

$$\mathbf{W}_1 = \{ C \in \mathbf{W} : C = AB - BA \}$$
 (3.5.15.2)

Consider $C \in \mathbf{W}$ such that C = AB - BA where $A, B \in \mathbf{R}^{N \times N}$ and since

$$trace(AB) = trace(BA)$$
 (3.5.15.3)

$$\implies trace(AB) - trace(BA) = 0 \quad (3.5.15.4)$$

$$\implies trace(C) = 0$$
 (3.5.15.5)

Thus any linear combination of matrices C =AB - BA represents the subspace of all $N \times$ N matrices with trace equal to 0. Hence, the subspace \mathbf{W}_0 and \mathbf{W}_1 are equivalent.

- be the set of all vectors $(\mathbf{x_1}, \mathbf{x_2}, \cdots, \mathbf{x_n}) \in \mathbf{F}^n$ such that $\mathbf{x_1} + \mathbf{x_2} + \mathbf{x_3} + \cdots + \mathbf{x_n} = 0$.
 - a) Prove that W^0 consists of all linear functional f of the form

$$f(\mathbf{x_1}, \mathbf{x_2}, \cdots, \mathbf{x_n}) = c \sum_{i=1}^{n} \mathbf{x_j}$$
 (3.6.1.1)

Solution: *Definition 1:* If the vector space **V** is finite dimensional (say dimension =n), the dimension of null-space N_f by rank nullity theorem is given by,

$$|\mathbf{N}_f| = |\mathbf{V}| - 1 = n - 1$$
 (3.6.1.2)

Definition 2: If V is a vector space over the field F and S is a subset of V, the annihilator of S is the set S^0 of linear functional f on Vsuch that $f(\alpha) = 0$ for every α in **S**. Solution Let h be the functional,

$$h(\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n) = \mathbf{x}_1 + \mathbf{x}_2 + \cdots + \mathbf{x}_n$$

$$Let, \mathbf{X} = \begin{pmatrix} \mathbf{x_1} \\ \mathbf{x_2} \\ \vdots \\ \mathbf{x_n} \end{pmatrix}$$

$$(3.6.1.4)$$

$$\implies h(\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n) = 0 \quad (3.6.1.5)$$

Then **W** is in null-space of h. Hence by definition-1, the dimension of **W** is,

$$|W| = n - 1 \tag{3.6.1.6}$$

Now let,

$$a_j = \epsilon_1 - \epsilon_{i+1}$$
, for $i = (1, \dots, n-1)$
(3.6.1.7)

Hence clearly $\{a_1, a_2, \dots, a_{n-1}\}$ are linearly independent. Hence from (3.6.1.6) and above statement we can conclude that $\{a_1, a_2, \dots, a_{n-1}\}$ are all in **W** so they must form basis for **W**. Now, it is given that f is linear functional hence,

$$f\left(\mathbf{x_1}, \mathbf{x_2}, \cdots, \mathbf{x_n}\right) = \sum_{j=1}^{n} c_j \mathbf{x_j}$$
 (3.6.1.8)

$$\implies f(\mathbf{x_1}, \mathbf{x_2}, \cdots, \mathbf{x_n}) = C^T \mathbf{X} \quad (3.6.1.9)$$

Where,

$$C = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix} \tag{3.6.1.10}$$

Now $f \in \mathbf{W}^0$ from definition-2 is given as,

$$f(a_1) = f(a_2) = \dots = f(a_n) = 0 \text{ for every } a_j \in \mathbf{W}$$
(3.6.1.11)

$$\implies c_1 - c_i = 0 \forall i = 2 \cdots n \qquad (3.6.1.12)$$

$$\implies c_i = c_1 = c \forall i \qquad (3.6.1.13)$$

$$\implies C = c \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \quad (3.6.1.14)$$

Hence,

$$f(\mathbf{x_1}, \mathbf{x_2}, \cdots, \mathbf{x_n}) = C^T \mathbf{X}$$
 (3.6.1.15)

PARAMETERS	MATRIX REPRESENTATION
Basis for V	$\mathbf{B} = \begin{pmatrix} \mathbf{v}_1 & \mathbf{v}_2 & \dots & \mathbf{v}_n \end{pmatrix}$
Basis for V*	$\mathbf{B} = \begin{pmatrix} \mathbf{v}_1 & \mathbf{v}_2 & \dots & \mathbf{v}_n \end{pmatrix}$ $\mathbf{B}^* = \begin{pmatrix} \mathbf{f}_1 & \mathbf{f}_2 & \dots & \mathbf{f}_n \end{pmatrix}$
$f_i(\mathbf{v}_j) = \delta_{ij}$	$(\mathbf{f}_i)^T \mathbf{v}_j = \delta_{ij}$
B , B * are dual basis	$(\mathbf{B}^*)^T\mathbf{B} = \mathbf{I}$
$\alpha_i = \sum_{k=1}^n a_k \mathbf{v}_k, i \in [1, m]$	$\alpha_i = \mathbf{Ba} \text{ where } \mathbf{a} = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix}, a_l \in \mathbb{F}, l \in [1, n]$
$\alpha_i \neq 0$	a ≠ 0
Any linear functional f over \mathbf{V}	$\mathbf{f} = \mathbf{B}^* \mathbf{c}$, where $\mathbf{c} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix}$, $c_l \in \mathbb{F}$, $l \in [1, n]$
$f(lpha_i)$	$\mathbf{f}^{T} \alpha_{i}$ $= (\mathbf{B}^{*} \mathbf{c})^{T} \alpha_{i}$ $= \mathbf{c}^{T} (\mathbf{B}^{*})^{T} \mathbf{B} \mathbf{a}$ $= \mathbf{c}^{T} \mathbf{a}$
f with $c_l \neq 0 \forall l \in [1, n]$ lies in $\mathbf{V}^*, \mathbf{a} \neq 0$	$f(\alpha_i) = \mathbf{c}^T \mathbf{a} \neq 0, i = 1, 2, \dots, m$

TABLE 3.5.12.2: Proof

Given	W is the space of $n \times n$ matrices over the field F f is a linear functional on W , that $f(AB) = f(BA)$
To prove	f is scalar multiple of trace $f(A) = c \times \operatorname{trace}(A)$ and If $f(I) = n$, then f is the trace $f(A) = \operatorname{trace}(A)$
Basis of W	Let $\{E^{pq}, 1 \le p, q \le n\}$ be the basis of W where E^{pq} is a $n \times n$ matrix $E^{pq} = \begin{cases} 1 & pq^{th}\text{-position} \\ 0 & \text{everywhere else} \end{cases}$

TABLE 3.5.14.1

Find $f(E^{pq})$	If $p = q$, we know $I = \sum_{p,q=1}^{n} E^{pp}$ then $f(I) = \sum_{p} f(E^{pp}) \implies f(I) = nf(E^{pp})$
	$f(E^{pp}) = \frac{f(I)}{n}$ If $p \neq q$, then $f(E^{pq}) = f(E^{p1}E^{1q}) = f(E^{1q}E^{p1}) = 0$ $\operatorname{trace}(E^{pq}) = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$
	$\therefore f(E^{pq}) = \frac{f(I)}{n} \operatorname{trace}(E^{pq})$
Proof $(find \ f(A))$	Let $A \in W$, then $A = \sum_{p,q=1}^{n} c^{pq} E^{pq}$
	Also, trace(A) = $\sum_{p,q=1}^{n} c^{pq} \text{trace}(E^{pq})$
	Since f is linear, $f(A) = \sum_{p,q=1}^{n} c^{pq} f(E^{pq})$
	$f(A) = \sum_{p,q=1}^{n} c^{pq} \left(\frac{f(I)}{n} \operatorname{trace}(E^{pq}) \right)$
	$f(A) = \frac{f(I)}{n} \operatorname{trace}(A)$
	Hence f is a scalar multiple of trace.
	If $f(I) = n$, then $f(A) = \text{trace}(A)$ Hence f is the trace function

TABLE 3.5.14.2

$$\implies f(\mathbf{x_1}, \mathbf{x_2}, \cdots, \mathbf{x_n}) = c \sum_{j=1}^{n} \mathbf{x_j}$$

Hence, proved

b) Show that the dual space **W*** of **W** can be 'naturally' identified with the linear functionals

$$f(x_1,\ldots,x_n)=c_1x_1+\ldots c_nx_n$$

on \mathbf{F}^n which satisfy $c_1 + \ldots + c_n = 0$ **Solution:** See Table 3.6.1.1 Pictorial Representation

(3.6.1.16) 3.6.2. Use theorem 20 to prove the following. If **W** is a subspace of a finite dimensional vector space \mathbf{V} and if $\{g_1, g_2, \dots, g_r\}$ is any basis for \mathbf{W}^0 then

$$\mathbf{W} = \bigcap_{i=1}^{r} \mathbf{N}_{g_i} \tag{3.6.2.1}$$

Solution:

Theorem 20: Let $g, f_1, f_2, ..., f_r$ be linear functionals on vector space **V** with respective null spaces $\mathbf{N}, \mathbf{N}_1, ..., \mathbf{N}_r$.

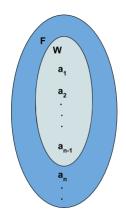


Fig. 3.6.1.1: **W** of dimension n-1, is the null space of **F**, where (a_1, \ldots, a_{n-1}) are basis vectors for **W**

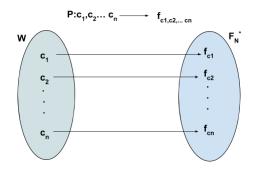


Fig. 3.6.1.2: Mapping from $\mathbf{W} \xrightarrow{P} \mathbf{F_N}^*$, where $P(c_1, ..., c_n) = f_{c_1, ..., c_n}, f_{c_1, ..., c_n}(x_1, ..., x_n) = c_1 x_1 + ... c_n x_n$

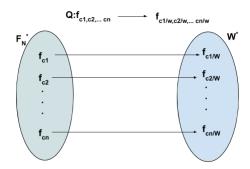


Fig. 3.6.1.3: Mapping from $\mathbf{F_N}^* \xrightarrow{Q} \mathbf{W}^*$, where $f_{c_1,\dots,c_n}(x_1,\dots,x_n) = c_1x_1 + \dots + c_nx_n$ is the linear functional with $c_1 + \dots + c_n = 0$

Then g is a linear combination of f_1, \ldots, f_r if and only if \mathbb{N} contains the intersection

 $\mathbf{N}_1 \cap \mathbf{N}_2 \cap \ldots \cap \mathbf{N}_r$.

See Tables 3.6.2.1, 3.6.2.2 for the proof.

3.6.3. Let S be a set, F a field, and V(S; F) the space of all functions from S into F:

$$(f+g)(x) = f(x) + g(x)$$
$$(cf)(x) = cf(x)$$

Let W be any n-dimensional subspace of V(S, F). Show that there exist points x_1, x_2, \ldots, x_n in S and functions f_1, \ldots, f_n in W such that $f_i(x_i) = \delta_{ij}$

Solution:

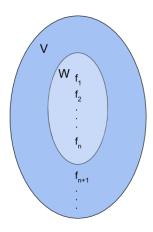


Fig. 3.6.3.1: Vector space of all function *V* and it's *n*-dimensional subspace *W*

Example:

Consider points $\{x_1, x_2\} \in S$, let

$$x_1 = 1 \tag{3.6.3.1}$$

$$x_2 = 0 (3.6.3.2)$$

Also consider functions $\{f_1, f_2\} \in W$ where

$$f_1(x) = x (3.6.3.3)$$

$$f_2(x) = 1 - x \tag{3.6.3.4}$$

Now, we have

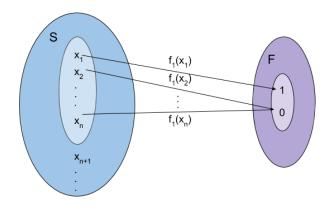
$$f_1(x_1) = f_1(1) = 1$$
 (3.6.3.5)

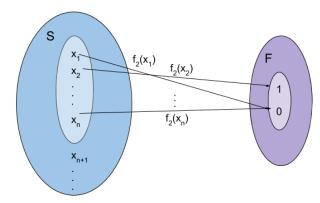
$$f_1(x_2) = f_1(0) = 0$$
 (3.6.3.6)

Also,

$$f_2(x_1) = f_2(1) = 0$$
 (3.6.3.7)

$$f_2(x_2) = f_2(0) = 1$$
 (3.6.3.8)





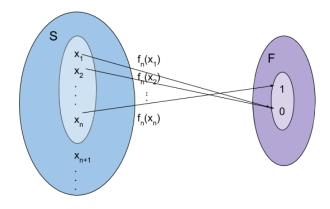


Fig. 3.6.3.2: Functions $f_1, f_2, ..., f_n$ where $f: S \rightarrow F$ and $f_i(x_i) = \delta_{ij}$ for i, j = 1, ..., n

Hence $f_i(x_i) = \delta_{ii}$.

- 3.7 The Transpose of a Linear Transformation
- 3.7.1. Let \mathbb{F} be a field and let f be the linear functional on \mathbb{F}^2 defined by,

$$f(x_1, x_2) = ax_1 + bx_2 (3.7.1.1)$$

a) For the linear operator $T(x_1, x_2) = (x_1, 0)$ Let, $g = T^t y$ and find $g(x_1, x_2)$

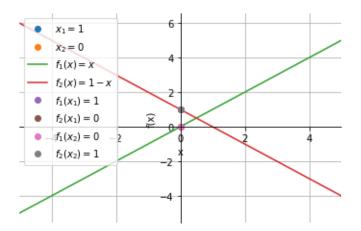


Fig. 3.6.3.3: Functions f_1 , f_2 and points x_1 , x_2

Solution: The linear functional f on \mathbb{F}^2 is defined by,

$$f(x_1, x_2) = \mathbf{a}^{\mathrm{T}} \mathbf{x} \quad \forall (x_1, x_2) \in \mathbb{F}^2 \quad (3.7.1.2)$$

where,

$$\mathbf{a} = \begin{pmatrix} a \\ b \end{pmatrix} \tag{3.7.1.3}$$

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \tag{3.7.1.4}$$

We use the following theorem,

Let \mathbb{V} and \mathbb{W} be vector spaces, over the field F. For each linear transformation $T: \mathbb{V} \to \mathbb{W}$, there is a unique linear transformation $T^t: \mathbb{W}^* \to \mathbb{V}^*$ such that,

$$(T^t g)(\alpha) = g(T\alpha) \tag{3.7.1.5}$$

The given linear operator T defined as,

$$T(x_1, x_2) = \mathbf{A}\mathbf{x} = \begin{pmatrix} x_1 \\ 0 \end{pmatrix} \quad \forall (x_1, x_2) \in \mathbb{F}^2$$
(3.7.1.6)

Where,

$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.7.1.7}$$

(3.7.1.8)

Consider the following mapping,

$$g = T^t f$$
 (3.7.1.9)

Then, $\forall (x_1, x_2) \in \mathbb{F}^2$ we have,

$$g(x_1, x_2) = T^t f(x_1, x_2) \quad [From (3.7.1.9)]$$

$$= f(T(x_1, x_2)) \quad [From (3.7.1.5)]$$

$$= \mathbf{a}^T \mathbf{A} \mathbf{x} \qquad (3.7.1.12)$$

$$= \mathbf{a}^T \begin{pmatrix} x_1 \\ 0 \end{pmatrix} \quad [From (3.7.1.6)]$$

$$= ax_1 \quad [From (3.7.1.2)]$$

$$= (3.7.1.14)$$

Hence, (3.7.1.14) is the required answer.

b) For given linear operator T, such that

$$\mathbf{T}(x_1, x_2) = (-x_2, x_1) \tag{3.7.1.15}$$

Let

$$g = \mathbf{T}^t f \tag{3.7.1.16}$$

Then find $g(x_1, x_2)$

Solution:

The linear operator **T** can be represented as a matrix A as follows

$$\mathbf{A} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \tag{3.7.1.17}$$

(3.7.1.18)

Let suppose,

$$\mathbf{X_1} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \mathbf{X_2} = \begin{pmatrix} -x_2 \\ x_1 \end{pmatrix} \tag{3.7.1.19}$$

And,

$$\mathbf{U} = \begin{pmatrix} a & b \end{pmatrix} \tag{3.7.1.20}$$

$$\mathbf{T}(x_1, x_2) = \mathbf{AX_1}$$
 (3.7.1.21)

$$\implies \mathbf{A}\mathbf{X_1} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \mathbf{X_1} = \mathbf{X_2} \quad (3.7.1.22)$$

And (3.7.1.15) can be written as

$$f(x_1, x_2) = \mathbf{UX_1}$$
 (3.7.1.23)

Now, we have given,

$$g = \mathbf{T}^t f$$
 (3.7.1.24)

$$\implies g(x_1, x_2) = \mathbf{T}^t f(x_1, x_2)$$
 (3.7.1.25)

We know that, if V and W be vector spaces over the field F. For each linear transformation T from V into W, there is a unique linear transformation T^t from W^* into V^* such that,

$$(\mathbf{T}^t g)(\alpha) = g(\mathbf{T}\alpha) \tag{3.7.1.26}$$

Where for every g in W^* and α in V.

Now using (3.7.1.26) in (3.7.1.25) we can write.

$$\mathbf{T}^{t} f(x_{1}, x_{2}) = f(\mathbf{T}(x_{1}, x_{2}))$$

$$(3.7.1.27)$$

$$\implies f(\mathbf{T}(x_{1}, x_{2})) = \mathbf{UAX}_{1}$$

$$(3.7.1.28)$$

$$f(\mathbf{T}(x_{1}, x_{2})) = \begin{pmatrix} a & b \end{pmatrix} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} = \mathbf{UX}_{2}$$

$$(3.7.1.29)$$

$$\implies f(\mathbf{T}(x_{1}, x_{2})) = -ax_{2} + bx_{1}$$

$$(3.7.1.30)$$

Hence,

$$f(\mathbf{T}(x_1, x_2)) = -ax_2 + bx_1 = g(x_1, x_2)$$
$$= (a \ b) {\begin{pmatrix} -x_2 \\ x_1 \end{pmatrix}} = \mathbf{UX_2} \quad (3.7.1.31)$$

c) For the linear operator $T(x_1, x_2) = (x_1 - x_2)$ $x_2, x_1 + x_2$

Let, $g = T^t y$ and find $g(x_1, x_2)$

Solution: Given,
$$\alpha = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$
. Let,

$$\mathbf{A} = \begin{pmatrix} \alpha_1 & \alpha_2 & \alpha_3 \end{pmatrix} \qquad (3.7.1.32)$$
$$\mathbf{AX} = \alpha \qquad (3.7.1.33)$$

$$\mathbf{AX} = \alpha \qquad (3.7.1.33)$$

$$\begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \\ 1 & -2 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$
 (3.7.1.34)

 $\mathbf{X} = A^{-1}\alpha$ will give solution of the equation.

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 1 & -2 & 0 & 0 & 0 & 1 \end{pmatrix} \xrightarrow{R_3 \leftarrow R_3 - R_1}$$

$$(3.7.1.35)$$

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & -2 & 1 & -1 & 0 & 1 \end{pmatrix} \xrightarrow{R_3 \leftarrow R_3 + 2R_2} (3.7.1.36)$$

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & 0 & -1 & -1 & 2 & 1 \end{pmatrix} \xrightarrow{R_3 \leftarrow R_3/(-1)} (3.7.1.37)$$

$$\begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & -2 & -1 \end{pmatrix} \xrightarrow{R_1 \leftarrow R_1 + R_3}$$

$$(3.7.1.38)$$

$$\begin{pmatrix} 1 & 0 & 0 & 2 & -2 & -1 \\ 0 & 1 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & -2 & -1 \end{pmatrix} \xleftarrow{R_2 \leftarrow R_2 + R_3}$$

$$\begin{pmatrix} 1 & 0 & 0 & 2 & -2 & -1 \\ 0 & 1 & 0 & 1 & -1 & -1 \\ 0 & 0 & 1 & 1 & -2 & -1 \end{pmatrix}$$
(3.7.1.40)

Thus,

$$A^{-1} = \begin{pmatrix} 2 & -2 & -1 \\ 1 & -1 & -1 \\ 1 & -2 & -1 \end{pmatrix}$$
 (3.7.1.41)

$$\mathbf{X} = A^{-1}\alpha = \begin{pmatrix} 2 & -2 & -1 \\ 1 & -1 & -1 \\ 1 & -2 & -1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad (3.7.1.42)$$

Given, f is a linear functional on R^3 ,

$$\alpha = \alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 \tag{3.7.1.43}$$

$$\implies f(\alpha) = \mathbf{X}^T \begin{pmatrix} f(\alpha_1) \\ f(\alpha_2) \\ f(\alpha_3) \end{pmatrix}$$
 (3.7.1.44)

Given, $f(\alpha_1) = 0$, $f(\alpha_2) = 0$ and $f(\alpha_3) \neq 0$.

$$f(\alpha) = \mathbf{X}^T \begin{pmatrix} 0 \\ 0 \\ f(\alpha_3) \end{pmatrix}$$
 (3.7.1.45)

$$\implies f(\alpha) = \begin{pmatrix} a \\ b \\ c \end{pmatrix}^{T} \begin{pmatrix} 2 & 1 & 1 \\ -2 & -1 & -2 \\ -1 & -1 & -1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ f(\alpha_{3}) \end{pmatrix}$$
(3.7.1.46)

$$f(\alpha) = \begin{pmatrix} a \\ b \\ c \end{pmatrix}^{T} \begin{pmatrix} f(\alpha_3) \\ -2f(\alpha_3) \\ -f(\alpha_3) \end{pmatrix}$$
 (3.7.1.47)

Hence,

$$f(\alpha) = f(\alpha_3)(a - 2b - c) \tag{3.7.1.48}$$

Given $\alpha = \begin{pmatrix} 2 & 3 & -1 \end{pmatrix}$ Substituting value of α in equation (3.7.1.48) we have,

$$f(\alpha) = -3f(\alpha_3) \neq 0$$
 (3.7.1.49)

Hence proved, $f(\alpha) \neq 0$

(3.7.1.39) 3.7.2. Let V be the vector space of all polynomial function over the field of real numbers. Let a and b be fixed real numbers and let f be the linear functional on V defined by

$$f(p) = \int_{a}^{b} p(x) dx$$
 (3.7.2.1)

If D is the differentiation operator on V , what is $D^t f$?

Solution: Let

$$p(x) = c_0 + c_1 x + c_2 x^2 + \dots + c_n x^n$$
(3.7.2.2)

$$D^{t} f(p) = f(D(p))$$
 (3.7.2.3)

$$D^{t} f(p) = f(c_1 + 2c_2x + 3c_3x^2 + \dots + nc_nx^{n-1})$$
(3.7.2.4)

$$D^{t}f(p) = \int_{a}^{b} \left(c_{1} + 2c_{2}x + 3c_{3}x^{2} + \dots + nc_{n}x^{n-1}\right) dx$$
(3.7.2.5)

$$D^{t}f(p) = c_{1}x + c_{2}x^{2} + c_{3}x^{3} + \dots + c_{n}x^{n}\Big|_{a}^{b}$$
(3.7.2.6)

$$D^{t}f(p) = p(b) - p(a)$$
 (3.7.2.7)

3.7.3. Let, **V** be the space of all $n \times n$ matrices over a field **F** and let B be a fixed $n \times n$ matrix. If T is the linear operator on **V** defined by T(A) = AB - BA, and if f is the trace function, what is $T^t f$?

Solution:

$$\mathbf{T}^t f(\mathbf{A}) = f(\mathbf{T}(\mathbf{A})) \tag{3.7.3.1}$$

$$= f(\mathbf{AB} - \mathbf{BA}) \tag{3.7.3.2}$$

$$= trace(\mathbf{AB} - \mathbf{BA}) \tag{3.7.3.3}$$

using $trace(\mathbf{AB}) = trace(\mathbf{BA})$ in (3.7.3.3),

$$\mathbf{T}^{t} f(\mathbf{A}) = trace(\mathbf{AB}) - trace(\mathbf{BA}) = 0$$
(3.7.3.4)

Hence $\mathbf{T}^t f = 0$

- 3.7.4. Let **A** be an $m \times n$ matrix with real entries. Prove that $\mathbf{A} = \mathbf{0}$ is and only if $tr(\mathbf{A}^T \mathbf{A}) = 0$. **Solution:** The proof is given in the table (3.7.4.2) and the properties used for the proof are listed in the table (3.7.4.1)
- 3.7.5. Let **V** be the vector space of all polynomial functions over the field of real numbers.Let *a* and *b* be fixed real numbers and let *f* be the linear functional on **V** defined by

$$f(p) = \int_{a}^{b} p(x) dx$$
 (3.7.5.1)

If D is the differentiator operator on V, what is $D^t f$?

Solution: See Tables 3.7.5.1 and 3.7.5.2

3.7.6. Let V be a finite-dimensional vector space over the field F. Show that $T \to T^t$ is an isomorphism of L(V, V) onto $L(V^*, V^*)$.

Solution: See Tables 3.7.6.1 and 3.7.6.1 *Example* Consider, $T \in L(\mathbb{R}^3, \mathbb{R}^3)$

$$T = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \text{ and } (3.7.6.1)$$

$$T' = \begin{pmatrix} 2 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \tag{3.7.6.2}$$

If $T \in L(\mathbb{R}^3, \mathbb{R}^3)$ we can show that $T' \in L(\mathbb{R}^{3^*}, \mathbb{R}^{3^*})$ as follows.

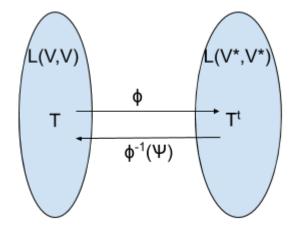


Fig. 3.7.6.1: $\phi: L(V, V) \to L(V^*, V^*)$

Consider linear functional $f: \mathbb{R}^{3^*} \to F$

$$f(x, y, z) = 2x + 3y + 4z (3.7.6.3)$$

Let $g: \mathbb{R}^3 \to F$. By definition of transpose,

$$g = T'f \tag{3.7.6.4}$$

$$= \begin{pmatrix} 2 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 7 \\ 5 \\ 6 \end{pmatrix}$$
 (3.7.6.5)

$$g(x, y, z) = 7x + 5y + 6z (3.7.6.6)$$

Consider vector in $v = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \in \mathbb{R}^3$,

$$Tv = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 7 \\ 3 \\ 3 \end{pmatrix}$$
 (3.7.6.7)

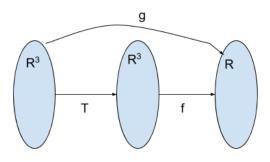
$$f(Tv) = f(7,3,3)$$
 (3.7.6.8)

$$= 2 * 7 + 3 * 3 + 4 * 3 = 35$$
 (3.7.6.9)

$$g(1,2,3) = 7 * 1 + 5 * 2 + 6 * 3 = 35$$
(3.7.6.10)

Hence verified g = T'f, and T' is transpose of T.

Now we will show that ϕ is invertible.



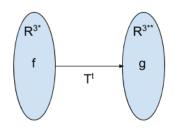


Fig. 3.7.6.2: $T: \mathbb{R}^3 \to \mathbb{R}^3$ and $T': \mathbb{R}^{3^*} \to \mathbb{R}^{3^{**}}$

Given,

$$\phi T = T' \tag{3.7.6.11}$$

$$\phi = T'T^{-1} \tag{3.7.6.12}$$

$$= \begin{pmatrix} 2 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 2 & 1 & 1 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}^{-1}$$
 (3.7.6.13)

$$= \begin{pmatrix} 2 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 & -1 \\ -1 & 2 & 1 \\ 0 & 0 & 1 \end{pmatrix}$$
 (3.7.6.14)

$$\phi = \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\phi^{-1} = \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}^{-1} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad (3.7.6.16)$$

Since ϕ^{-1} exists, ϕ is isomorphism of L(V, V) onto $L(V^*, V^*)$

- 3.7.7. Let *V* be the vector space of $n \times n$ matrices over the field *F*. If **B** is a fixed $n \times n$ matrix, define a function f_B on *V* by $f_B(\mathbf{A}) = tr(\mathbf{B}^t \mathbf{A})$.
 - a) Show that, f_B is a linear functional on V. Solution:

See Table 3.7.7.1

Hence, it can be said that f_B is a linear functional on V.

b) Let **V** be the vector space of $n \times n$ matrices over the field \mathbb{F} . If **B** is a fixed $n \times n$ matrix, define a function f_B on **V** by $f_B(\mathbf{A}) = Tr(\mathbf{B}^T\mathbf{A})$. Show that every linear functional on **V** is of the form f_B for some **B**.

Solution: For $A = (a_{ij}) \in V$, a linear functional $f_B : V \longrightarrow \mathbb{F}$ is defined as:

$$f_B(\mathbf{A}) = \mathbf{A}^T \mathbf{c} \tag{3.7.7.1}$$

(3.7.7.1) can be written as,

$$f_B(\mathbf{A}) = \sum_{i=1}^n \sum_{j=1}^n c_{ij} a_{ij}$$
 (3.7.7.2)

Now,

$$Tr(\mathbf{B}^{t}\mathbf{A}) = \sum_{i=1}^{n} (\mathbf{B}^{t}\mathbf{A})_{ii}$$
 (3.7.7.3)

$$= \sum_{j=1}^{n} \sum_{i=1}^{n} (b_{ji})^{T} a_{ij} \qquad (3.7.7.4)$$

$$= \sum_{j=1}^{n} \sum_{j=1}^{n} b_{ij} a_{ij}$$
 (3.7.7.5)

Let $c_{ij} = b_{ij}$, from (3.7.7.2) and (3.7.7.5),

$$Tr(\mathbf{B}^T \mathbf{A}) = f_B \tag{3.7.7.6}$$

Hence, Proved.

4 Polynomials

4.1 The Algebra of Polynomials

(3.7.6.15) 4.1.1. Let F be a sub field of the complex numbers , and let A be the following 2×2 matrix over F.

$$\mathbf{A} = \begin{pmatrix} 2 & 1 \\ -1 & 3 \end{pmatrix} \tag{4.1.1.1}$$

Compute f(A) for the polynomial

a)

$$f(x) = x^2 - x + 2 (4.1.1.2)$$

Solution: consider

$$|\mathbf{A} - \lambda \mathbf{I}| = 0 \tag{4.1.1.3}$$

$$\begin{pmatrix} 2 - \lambda & 1 \\ -1 & 3 - \lambda \end{pmatrix} = 0 \tag{4.1.1.4}$$

The characteristics equation of the matrix

$$\lambda^2 - 5\lambda + 7 = 0 \tag{4.1.1.5}$$

The characteristics equation will satisfy its own matrix

$$\mathbf{A}^2 - 5\mathbf{A} + 7 = 0 \tag{4.1.1.6}$$

$$\mathbf{A}^2 = 5\mathbf{A} - 7\mathbf{I} \tag{4.1.1.7}$$

The given polynomial

$$f(x) = x^2 - x + 2 (4.1.1.8)$$

$$f(\mathbf{A}) = \mathbf{A}^2 - \mathbf{A} + 2\mathbf{I} \tag{4.1.1.9}$$

substituting (4.1.1.7) in (4.1.1.9) we get

$$f(\mathbf{A}) = 5\mathbf{A} - 7\mathbf{I} - \mathbf{A} + 2\mathbf{I}$$
 (4.1.1.10)

$$f(\mathbf{A}) = 4\mathbf{A} - 5\mathbf{I}$$
 (4.1.1.11)

$$f(\mathbf{A}) = 4 \begin{pmatrix} 2 & 1 \\ -1 & 3 \end{pmatrix} - 5 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \quad (4.1.1.12)$$

$$f(\mathbf{A}) = \begin{pmatrix} 3 & 4 \\ -4 & 7 \end{pmatrix} \tag{4.1.1.13}$$

b)

$$f = x^3 - 1 \tag{4.1.1.14}$$

Solution: We first find the eigen values of the **A**. We get the characteristic equation of **A** as follows,

$$\det\left(\mathbf{A} - \lambda \mathbf{I}\right) = 0 \tag{4.1.1.15}$$

$$\implies \lambda^2 - 5\lambda + 7 = 0$$
 (4.1.1.16)

Where **I** is 2×2 Identity matrix. Now using Cayley Hamilton Theorem we get from (4.1.1.16) the following,

$$\mathbf{A}^2 - 5\mathbf{A} + 7\mathbf{I} = 0 \tag{4.1.1.17}$$

Hence the polynomial f(A) can be written using the characteristic function of A as

follows,

$$f(\mathbf{A}) = \mathbf{A}^{3} - \mathbf{I}$$
 (4.1.1.18)

$$= (\mathbf{A} - \mathbf{I})(\mathbf{A}^{2} + \mathbf{A} + \mathbf{I})$$
 (4.1.1.19)

$$= (\mathbf{A} - \mathbf{I})(\mathbf{A}^{2} - 5\mathbf{A} + 7\mathbf{I} + 6\mathbf{A} - 6\mathbf{I})$$
 (4.1.1.20)

$$= 6(\mathbf{A} - \mathbf{I})^{2}$$
 [From (4.1.1.17)]

$$= 6\left[\begin{pmatrix} 2 & 1 \\ -1 & 3 \end{pmatrix} - \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}\right]^{2}$$
 (4.1.1.22)

$$= \begin{pmatrix} 0 & 18 \\ -18 & 18 \end{pmatrix}$$
 (4.1.1.23)

Here, (4.1.1.23) is the required answer.

c) $f(x) = x^2 - 5x + 7$

Solution: Characteristic equation of **A** can be written as:

$$|\mathbf{A} - \lambda \mathbf{I}| = 0 \tag{4.1.1.24}$$

$$\begin{pmatrix} 2 - \lambda & 1 \\ -1 & 3 - \lambda \end{pmatrix} = 0 \tag{4.1.1.25}$$

The characteristics equation of the matrix will be,

$$\lambda^2 - 5\lambda + 7 = 0 \tag{4.1.1.26}$$

The characteristics equation will satisfy its own matrix

$$\mathbf{A}^2 - 5\mathbf{A} + 7\mathbf{I} = \mathbf{0} \tag{4.1.1.27}$$

The given polynomial

$$f(x) = x^2 - 5x + 7$$
 (4.1.1.28)

$$f(\mathbf{A}) = \mathbf{A}^2 - 5\mathbf{A} + 7\mathbf{I}$$
 (4.1.1.29)

substituting (4.1.1.27) in (4.1.1.29) we get

$$f(\mathbf{A}) = \mathbf{0} \tag{4.1.1.30}$$

4.1.2. Let T be the linear operator on \mathbb{R}^3 defined by

$$T(x_1, x_2, x_3) = (x_1, x_3, -2x_2 - x_3)$$
 (4.1.2.1)

Let f be the polynomial over \mathbb{R} defined by $f = -x^3 + 2$

Solution: The given transformation can be

written as,

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{4.1.2.2}$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix} \mathbf{x} \tag{4.1.2.3}$$

Hence,

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix} \tag{4.1.2.4}$$

Now the characteristic equation of A is given by,

$$\det (\mathbf{A} - \lambda \mathbf{I}) = 0 \qquad (4.1.2.5)$$

$$= \begin{pmatrix} 1 - \lambda & 0 & 0 \\ 0 & -\lambda & 1 \\ 0 & -2 & -1 - \lambda \end{pmatrix} (4.1.2.6)$$

$$\implies (1 - \lambda)(\lambda^2 + \lambda + 2) = 0$$

$$(4.1.2.7)$$

Now, simplifying the above equation,

$$(1 - \lambda)(\lambda^2 + \lambda + 2) = 0$$
 (4.1.2.8)

$$\lambda^2 + \lambda + 2 - \lambda^3 - \lambda^2 - 2\lambda = 0$$
 (4.1.2.9)

$$\lambda^3 = 2 - \lambda \qquad (4.1.2.10)$$

Now using Cayley Hamilton Theorem we get,

$$\mathbf{A}^3 = 2\mathbf{I} - \mathbf{A} \tag{4.1.2.11}$$

Hence the polynomial $f(\mathbf{A})$ can be written using the characteristic function of A as follows,

$$f(\mathbf{A}) = -\mathbf{A}^3 + 2\mathbf{I}$$
 (4.1.2.12)
= $2\mathbf{I} - \mathbf{A} + 2\mathbf{I}$ (4.1.2.13)

$$= 2\mathbf{I} - \mathbf{A} + 2\mathbf{I} \tag{4.1.2.13}$$

$$= \mathbf{A} \tag{4.1.2.14}$$

Hence,

$$f(T)(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{4.1.2.15}$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix} \mathbf{x}$$
 (4.1.2.16)

4.1.3. Let A be an $n \times n$ diagonal matrix over the 4.1.4. If f and g are independent polynomials over a field \mathbf{F} , i.e., a matrix satisfying $\mathbf{A}_{ij} = 0$ for $i \neq j$. Let f be the polynomial over **F** defined by $f = (x - A_{11})...(x - A_{nn})$. What is the matrix

Solution: The given transformation can be

written as,

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{4.1.3.1}$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix} \mathbf{x}$$
 (4.1.3.2)

Hence,

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix} \tag{4.1.3.3}$$

Now the characteristic equation of A is given by,

$$\det\left(\mathbf{A} - \lambda \mathbf{I}\right) = 0 \tag{4.1.3.4}$$

$$= \begin{pmatrix} 1 - \lambda & 0 & 0 \\ 0 & -\lambda & 1 \\ 0 & -2 & -1 - \lambda \end{pmatrix} (4.1.3.5)$$

$$\implies (1 - \lambda)(\lambda^2 + \lambda + 2) = 0$$
(4.1.3.6)

Now, simplifying the above equation,

$$(1 - \lambda)(\lambda^2 + \lambda + 2) = 0$$
 (4.1.3.7)

$$\lambda^{2} + \lambda + 2 - \lambda^{3} - \lambda^{2} - 2\lambda = 0 \qquad (4.1.3.8)$$

$$\lambda^3 = 2 - \lambda \tag{4.1.3.9}$$

Now using Cayley Hamilton Theorem we get,

$$\mathbf{A}^3 = 2\mathbf{I} - \mathbf{A} \tag{4.1.3.10}$$

Hence the polynomial $f(\mathbf{A})$ can be written using the characteristic function of A as follows,

$$f(\mathbf{A}) = -\mathbf{A}^3 + 2\mathbf{I} \tag{4.1.3.11}$$

$$= 2\mathbf{I} - \mathbf{A} + 2\mathbf{I}$$
 (4.1.3.11)
= $2\mathbf{I} - \mathbf{A} + 2\mathbf{I}$ (4.1.3.12)

$$= \mathbf{A} \tag{4.1.3.13}$$

Hence,

$$f(T)(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{4.1.3.14}$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix} \mathbf{x} \tag{4.1.3.15}$$

field Fand h is a non-zero polynomial over F, show that fh and gh are independent.

Solution: Polynomials (f_1, f_2, \dots, f_m) $k(x_1,\ldots,x_n)$ are called algebraically independent over a field F, if there is no nonzero m-variate polynomial A ∈ $k[y_1,\ldots,y_m]$ such that $A(f_1,\ldots,f_m)=0$ Example: The smallest degree independent polynomials are: 1+x and 1-x.

$$a(1+x) + b(1-x) = 0 (4.1.4.1)$$

Simplifying,

$$a + b = 0 (4.1.4.2)$$

$$a - b = 0 (4.1.4.3)$$

solving,we get a=0 and b=0.So, polynomials are linearly independent. Given f and g are independent polynomials over a field F. Consider scalars a and $b \in F$ Hence,

$$af + bg = 0$$
 (4.1.4.4)

Since f and g are independent Hence f and g $\neq 0$

$$\implies a, b = 0.$$
 (4.1.4.5)

Given h a non zero polynomial over F. Substituting in equation (4.1.4.4) we have,

$$a(fh) + b(gh) = 0$$
 (4.1.4.6)

$$(af)h + (bg)h = 0$$
 (4.1.4.7)

$$(af + bg)h = 0$$
 (4.1.4.8)

$$af + bg = 0$$
 (4.1.4.9)

f and g are independent polynomial. Also from equation (4.1.4.5) a=0 and b=0.

Hence proved fh and gh are independent.

4.1.5. Let S be a set of non-zero polynomials over a field F. If no two elements of S have the same 4.2.1. Use the Lagrange interpolation formula to find degree, show that S is an independent set in $\mathbf{F}[\mathbf{x}]$.

Solution: See Tables 4.1.5.1, 4.1.5.2 and 4.1.5.3

Hence, it is proved that **S** is an independent set in F[x].

4.1.6. Let ${\mathbb F}$ be a subfield of complex numbers and let T be the transformation on $\mathbb{F}(x)$ defined by

$$T\left(\sum_{i=0}^{n} c_i x^i\right) = \sum_{i=0}^{n} \frac{c_i}{i+1} x^{i+1}$$
 (4.1.6.1)

a) Show that T is a non-singular linear operator on $\mathbb{F}[x]$. Also show that T is not invertible. transformation T **Solution:** The does integral of a polynomial. Table

refeq:solutions/4/1/9/a/table:2 provides proof that the transformation T is a linear operator and non-singular. Table refeq:solutions/4/1/9/a/table:3 provides proof that T is not invertible, however there exists a left inverse. The parameters used in the proof are listed in the table refeq:solutions/4/1/9/a/table:1.

b) Let **F** be a sub-field of the complex numbers and let **D** be the transformation on $\mathbf{F}[x]$ defined by

$$\mathbf{D}\left(\sum_{i=0}^{n} c_{i} x^{i}\right) = \sum_{i=0}^{n} i c_{i} x^{i-1}$$
 (4.1.6.14)

$$\mathbf{D}(\mathbf{x}^{\mathrm{T}}\mathbf{c}) = \mathbf{y}^{\mathrm{T}}\mathbf{b} \tag{4.1.6.15}$$

where

$$\mathbf{x} = \begin{pmatrix} 1 \\ x \\ \vdots \\ x^n \end{pmatrix} \quad and \quad \mathbf{c} = \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_n \end{pmatrix}$$
 (4.1.6.16)

$$\mathbf{y} = \begin{pmatrix} 1 \\ x \\ \vdots \\ x^{n-1} \end{pmatrix} \quad and \quad \mathbf{b} = \begin{pmatrix} c_1 \\ 2c_2 \\ \vdots \\ nc_n \end{pmatrix} \quad (4.1.6.17)$$

Show that **D** is a linear operator on $\mathbf{F}[x]$ and find its null space.

Solution: See Tables 4.1.6.4 and 4.1.6.5

4.2 Lagrange Interpolation

a polynomial f with real coefficients such that f has degree ≤ 3 and f(-1)=-6, f(0)=2, f(1)=-2f(2)=6.

Solution: Let the required polynomial be

$$f(x) = a + bx + cx^{2} + dx^{3}$$
 (4.2.1.1)

Testing points are

$$t_0 = -1, t_1 = 0, t_2 = 1, t_3 = 2$$
 (4.2.1.2)

Substituting the values of testing points in (4.2.1.1) we form matrix equation

$$\begin{pmatrix} 1 & -1 & 1 & -1 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = \begin{pmatrix} -6 \\ 2 \\ -2 \\ 6 \end{pmatrix}$$
(4.2.1.3)

By row reducing augmented matrix we get

$$\begin{pmatrix} 1 & -1 & 1 & -1 & -6 \\ 1 & 0 & 0 & 0 & 2 \\ 1 & 1 & 1 & 1 & -2 \\ 1 & 2 & 4 & 8 & 6 \end{pmatrix}$$

$$f(x) = 2 - 2x - 6x^2 + 4x^3 (4.2.1.13)$$

 $\begin{cases} 1 & -1 & 1 & -1 & -6 \\ 1 & 0 & 0 & 0 & 2 \\ 1 & 1 & 1 & 1 & -2 \\ 1 & 2 & 4 & 9 & 6 \end{cases}$ (4.2.1.13) $(4.2.1.4) \ 4.2.2. \text{ Let } T \text{ be the linear operator on } R^3 \text{ defined by } T(x_1, x_2, x_3) = (x_1, x_3, -2x_2 - x_3)$ $T(x_1, x_2, x_3) = (x_1, x_3, -2x_2 - x_3)$

> Let f be the polynomial over R defined by f = $-x^{3} + 2$. Find f(T).

Solution: The transformation is given as:

$$\xrightarrow{R_2 \leftarrow R_2 - R_1} \left(\begin{array}{ccccc} 1 & -1 & 1 & -1 & -6 \\ 0 & 1 & -1 & 1 & 8 \\ 0 & 2 & 0 & 2 & 4 \\ 1 & 2 & 4 & 8 & 6 \end{array} \right)$$
(4.2.1.5)

Solution: The transformation is g .

$$T(\mathbf{x}) = \begin{pmatrix} x_1 \\ x_3 \\ 2 & y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \\ y_6 \\ y_6 \\ y_7 \\ y$$

$$\xrightarrow[R_3 \leftarrow R_3/2]{R_4 \leftarrow R_4 - R_1} \begin{pmatrix}
1 & -1 & 1 & -1 & -6 \\
0 & 1 & -1 & 1 & 8 \\
0 & 1 & 0 & 1 & 2 \\
0 & 3 & 3 & 9 & 12
\end{pmatrix} (4.2.1.6)$$

$$\stackrel{R_4 \leftarrow R_4/3}{\underset{R_3 \leftarrow R_3 - R_2}{\longleftrightarrow}} \begin{pmatrix}
1 & -1 & 1 & -1 & -6 \\
0 & 1 & -1 & 1 & 8 \\
0 & 0 & 1 & 0 & -6 \\
0 & 1 & 1 & 3 & 4
\end{pmatrix} (4.2.1.7)$$

$$\stackrel{R_1 \leftarrow R_1 + R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & 0 & 2 \\
0 & 1 & -1 & 1 & 8 \\
0 & 0 & 1 & 0 & -6 \\
0 & 0 & 2 & 2 & -4
\end{pmatrix}$$
(4.2.1.8)

$$\xrightarrow[R_4 \leftarrow R_4 \rightarrow R_2]{R_1 \leftarrow R_1 + R_2} \begin{pmatrix}
1 & 0 & 0 & 0 & 2 \\
0 & 1 & -1 & 1 & 8 \\
0 & 0 & 1 & 0 & -6 \\
0 & 0 & 2 & 2 & -4
\end{pmatrix}$$
(4.2.1.9)

$$\stackrel{R_4 \leftarrow R_4/2}{\underset{R_2 \leftarrow R_2 + R_3}{\longleftrightarrow}} \begin{pmatrix}
1 & 0 & 0 & 0 & 2 \\
0 & 1 & 0 & 1 & 2 \\
0 & 0 & 1 & 0 & -6 \\
0 & 0 & 1 & 1 & -2
\end{pmatrix}$$
(4.2.1.10)

$$\stackrel{R_4 \leftarrow R_4 - R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & 0 & 2 \\
0 & 1 & 0 & 0 & -2 \\
0 & 0 & 1 & 0 & -6 \\
0 & 0 & 0 & 1 & 4
\end{pmatrix}$$
(4.2.1.11)

Therefore,

$$T(\mathbf{x}) = \begin{pmatrix} x_1 \\ x_3 \\ -2x_2 - x_3 \end{pmatrix}$$
 (4.2.2.1)
$$T(\mathbf{x}) = \mathbf{TX}$$
 (4.2.2.2)

$$T(\mathbf{x}) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix} \mathbf{X}$$
 (4.2.2.3)

Characteristic equation of **T** can be written as:

$$\left|\mathbf{T} - \lambda \mathbf{I}\right| = 0 \tag{4.2.2.4}$$

$$\begin{pmatrix} 1 - \lambda & 0 & 0 \\ 0 & -\lambda & 1 \\ 0 & -2 & -1 - \lambda \end{pmatrix} = 0 \tag{4.2.2.5}$$

The characteristics equation of the matrix will be.

$$-\lambda^3 - \lambda + 2 = 0 (4.2.2.6)$$

$$-\lambda^3 + 2 = \lambda \tag{4.2.2.7}$$

Given, Polynomial

$$f(x) = -x^3 + 2 (4.2.2.8)$$

$$f(\lambda) = -\lambda^3 + 2 \tag{4.2.2.9}$$

$$f(\lambda) = \lambda \tag{4.2.2.10}$$

Substituting **T** for λ in equation 4.2.2.10, We get

$$f(\mathbf{T}) = \mathbf{T} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -2 & -1 \end{pmatrix}$$
(4.2.2.11)

4.2.3. Let **F** be the field of real numbers,

$$\begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = \begin{pmatrix} 2 \\ -2 \\ -6 \\ 4 \end{pmatrix}$$
 (4.2.1.12)
$$\mathbf{A} = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$
 (4.2.3.1)

$$p = (x-2)(x-3)(x-1)$$
 (4.2.3.2)

a) Let \mathbb{F} be the field of real numbers,

$$\mathbf{A} = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \tag{4.2.3.3}$$

$$p = (x-2)(x-3)(x-1)$$
 (4.2.3.4)

Solution:

- i) Show that $p(\mathbf{A}) = 0$
- ii) Let P_1, P_2, P_3 be the Lagrange polynomials for $t_1 = 2, t_2 = 3, t_3 = 1$. Compute $E_i = P_i(\mathbf{A}), i=1,2,3$
- i) Since A is a diagonal matrix, It's characteristic polynomial is,

$$\det (\mathbf{A} - x\mathbf{I}) = 0$$
(4.2.3.5)

$$f(x) = (x-2)^{2}(x-3)(x-1) = 0$$
(4.2.3.6)

From, (4.2.3.6) and using Cayley Hamilton Theorem,

$$(\mathbf{A} - 2)^2(\mathbf{A} - 3)(\mathbf{A} - 1) = 0$$
 (4.2.3.7)

We can also see that (x-2)(x-3)(x-1)is a minimal polynomial for A, Hence $p(\mathbf{A})=0$.

ii) Using Lagrange Interpolation,

$$P_1(x) = \frac{(x-3)(x-1)}{(2-3)(2-1)}$$
 (4.2.3.8)

$$= -(x-3)(x-1)$$
 (4.2.3.9)

$$P_{2}(x) = \frac{(x-2)(x-1)}{(3-2)(3-1)}$$

$$= \frac{(x-2)(x-1)}{2}$$

$$(4.2.3.10)$$

$$= \frac{(x-2)(x-1)}{2}$$

$$(4.2.3.11)$$

$$P_{3}(x) = \frac{(x-2)(x-3)}{(1-2)(1-3)}$$

$$(4.2.3.12)$$

$$=\frac{(x-2)(x-1)}{2} \qquad (4.2.3.11)$$

$$P_3(x) = \frac{(x-2)(x-3)}{(1-2)(1-3)}$$
 (4.2.3.12)

$$=\frac{(x-2)(x-3)}{2} \qquad (4.2.3.13)$$

Now, Substituting the value of A,

b) Let P_1 , P_2 , P_3 be the Lagrange polynomials for $t_1 = 2$, $t_2 = 3$, $t_3 = 1$. Compute $E_i =$ $P_i(\mathbf{A})$ i = 1, 2, 3.

Solution: Lagrange polynomials are given

by,

$$P_{i} = \prod_{i \neq j} \frac{x - t_{j}}{t_{i} - t_{j}}$$

$$(4.2.3.20)$$

$$P_{1} = \frac{(x - 3)(x - 1)}{(2 - 3)(2 - 1)} = -(x - 3)(x - 1)$$

$$(4.2.3.21)$$

$$P_{2} = \frac{(x - 2)(x - 1)}{(3 - 2)(3 - 1)} = \frac{1}{2}(x - 2)(x - 1)$$

$$(4.2.3.22)$$

$$P_{3} = \frac{(x - 2)(x - 3)}{(1 - 2)(1 - 3)} = \frac{1}{2}(x - 2)(x - 3)$$

$$(4.2.3.23)$$

Now calculating E_i ,

$$E_1 = P_1(A) = -(A - 3I)(A - I)$$

4.2.4. Let n be a positive integer and \mathbb{F} be a field. Suppose \mathbf{A} is an $n \times n$ matrix over field \mathbb{F} and \mathbf{P} is an invertible $n \times n$ matrix over field \mathbb{F} . If f is any polynomial over \mathbb{F} , prove that,

$$f(\mathbf{P}^{-1}\mathbf{A}\mathbf{P}) = \mathbf{P}^{-1}f(\mathbf{A})\mathbf{P}$$
 (4.2.4.1)

Solution: First we observe the following,

$$(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})^2 = (\mathbf{P}^{-1}\mathbf{A}\mathbf{P})(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})$$
 (4.2.4.2)
= $\mathbf{P}^{-1}\mathbf{A}^2\mathbf{P}$ (4.2.4.3)

Let the (4.2.4.3) be true for a positive integer m i.e,

$$(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})^m = \mathbf{P}^{-1}\mathbf{A}^m\mathbf{P}$$
 (4.2.4.4)

Now for the integer m+1 we get from (4.2.4.4),

$$(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})^{m+1} = (\mathbf{P}^{-1}\mathbf{A}^{m}\mathbf{P})(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})$$
 (4.2.4.5)
= $\mathbf{P}^{-1}\mathbf{A}^{m+1}\mathbf{P}$ (4.2.4.6)

From (4.2.4.4) and (4.2.4.6) we get for any positive integer n,

$$(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})^n = \mathbf{P}^{-1}\mathbf{A}^n\mathbf{P}$$
 (4.2.4.7)

Again we have,

$$\mathbf{P}^{-1}\mathbf{P} = \mathbf{I} \tag{4.2.4.8}$$

The general form of polynomial $f(\mathbf{A})$ is defined as,

$$f(\mathbf{A}) = a_0 + a_1 \mathbf{A} + a_2 \mathbf{A}^2 + \dots + a_n \mathbf{A}^n$$
(4.2.4.9)

Now we have,

$$f(\mathbf{P}^{-1}\mathbf{A}\mathbf{P}) = a_0 + a_1(\mathbf{P}^{-1}\mathbf{A}\mathbf{P}) + \dots + a_n(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})^{n-1} = 2.0. \text{ if } F \text{ is a possible of degree}$$

$$(4.2.4.10) \qquad \qquad \text{of degree}$$

$$f(h) \text{ is a of degree}$$

$$f(h) \text{ is a of degree}$$

$$f(a_0 + a_1\mathbf{A}\mathbf{P}) + \dots + (\mathbf{P}^{-1}a_n\mathbf{A}\mathbf{P})^{n} \text{ into } F[x].$$

$$(4.2.4.11) \qquad \text{isomorphis}$$

$$= (\mathbf{P}^{-1}a_0\mathbf{P}) + (\mathbf{P}^{-1}a_1\mathbf{A}\mathbf{P}) + \dots + (\mathbf{P}^{-1}a_n\mathbf{A}^n\mathbf{P}) \text{deg } h = 1.$$

$$(4.2.4.12) \qquad \text{Solution:}$$

$$= \mathbf{P}^{-1}(a_0 + a_1\mathbf{A} + a_2\mathbf{A}^2 + \dots + a_n\mathbf{A}^n)\mathbf{P}$$

$$(4.2.4.13) \qquad \text{over field}$$

$$= \mathbf{P}^{-1}f(\mathbf{A})\mathbf{P} \quad [\text{From } (4.2.4.9)]$$

$$(4.2.4.14)$$

Hence proved (4.2.4.14)

4.2.5. Let F be a field. We have considered certain special linear functionals on $\mathbf{F}[x]$ obtained via 'evaluation at t': L(f) = f(t). Such functionals are not only linear but also have the property that L(fg) = L(f)L(g). Prove that if L is any linear functional on F[x] such that L(fg) =L(f)L(g) for all f and g, then either L=0 or there is a t in F such that L(f) = f(t) for all f. **Solution:** Let L be a non zero linear transformation.

$$f(x) = a_0 + a_1(x) + \dots + a_n(x^n)$$
 (4.2.5.1)

$$L(f) = L(a_0 + a_1(x) + \dots + a_n(x^n)).$$
 (4.2.5.2)

Given,L is any linear functional on f(x). Hence,

$$L(f) = L(f.1) = L(f)L(1)$$
 (4.2.5.3)

 $\implies L(1) \neq 0.$

Similarly,

$$L(1) = L(1)L(1)$$
 (4.2.5.4)

$$\implies L(1) = 1$$
 (4.2.5.5)

Similarly, $\forall a \in \mathbf{F}$.

$$L(a) = L(a.1) \implies aL(1) = a$$
 (4.2.5.6)

Simplifying (4.2.5.2) we have,

$$L(f) = L(a_0) + L(a_1)L(x) + \dots + L(a_n)L(x^n)$$
(4.2.5.7)

From (4.2.5.6) we have, L(x) = t.

$$L(f) = a_0 + a_1 L(x) + \dots + a_n L(x)^n$$
 (4.2.5.8)

$$\implies L(f) = a_0 + a_1(t) + \cdots + a_n(t)^n \quad (4.2.5.9)$$

Hence proved, L(f) = f(t).

 $f(\mathbf{P}^{-1}\mathbf{A}\mathbf{P}) = a_0 + a_1(\mathbf{P}^{-1}\mathbf{A}\mathbf{P}) + \dots + a_n(\mathbf{P}^{-1}\mathbf{A}\mathbf{P})^n$ 4.2.6. If F is a field and h is a polynomial over Ff(h) is a one-one linear transformation of F[x] $= (\mathbf{P}^{-1}a_0\mathbf{P}) + (\mathbf{P}^{-1}a_1\mathbf{A}\mathbf{P}) + \dots + (\mathbf{P}^{-1}a_n\mathbf{A}\mathbf{P})^n \text{ into } F[x].$ Show that this transformation is an isomorphism of F[x] onto F[x] if and only if

> **Solution:** Here, F[x] is a set of polynomials over field F, written as:

$$F[x] = \left\{ \sum_{i=0}^{\infty} a_i x^i \mid a_i \in F \right\}$$
 (4.2.6.1)

Let,

$$G(f) = f(h)$$
 (4.2.6.2)

Thus, G(f) is clearly a function from F[x] into F[x].

Now, we need to show that the function G is one-one linear transformation. Let us first show that G is a linear transformation:

Let,
$$f, g \in F[x]$$
 and $\alpha \in F$

$$G(\alpha f + g) = (\alpha f + g)(h)$$

$$= (\alpha f)(h) + g(h)$$

$$= \alpha f(h) + g(h)$$

$$= \alpha G(f) + G(g) \qquad (4.2.6.3)$$

From (4.2.6.3), G is a linear transformation. For G to be one-one linear transformation, it should map a set of linearly independent polynomials in F(x) to another set of linearly independent polynomials in F(x). let us consider the following basis set for F(x):

$$S = \{f_0, f_1, f_2, f_3, f_4, \ldots\}$$
 (4.2.6.4)

Where.

$$f_i = x^i (4.2.6.5)$$

Since, the set S forms the basis for F(x), the set S is a set of linearly independent polynomials. Let us apply the transformation G to set S, then we obtain another set S' as:

$$S' = \{ f_0(h), f_1(h), f_2(h), f_3(h), f_4(h), \dots \}$$
(4.2.6.6)

Where,

$$f_i = x^i \tag{4.2.6.7}$$

Here, The degree of each polynomial in set S' is distinct and given by $i \cdot \deg(h)$. Thus, set S' is also a set of linearly independent polynomials.

Conclusion: G will maps any arbitrary set S_a of linearly independent polynomials in F(x) to another set S'_a of linearly independent polynomials in F(x). (Since any arbitrary set S_a can be written in terms of basis set S). Hence, G is one-one linear transformation.

Now, Let us prove that G is an isomorphism of F(x) onto F(x) if and only if deg(h) = 1. Let deg(h) = 1, then h can be written as:

$$h = a + bx$$
, Where, $b \neq 0$ (4.2.6.8)

Let us define h' such that:

$$h' = \frac{1}{b}x - \frac{a}{b}$$
 (4.2.6.9) 4.3 Polynomial Ideals

to F(x) given by:

$$G'(f) = f\left(\frac{1}{b}x - \frac{a}{b}\right)$$
 (4.2.6.10)

It can be shown that G' is inverse of G as follow:

$$G(G'(f)) = G\left(f\left(\frac{1}{b}x - \frac{a}{b}\right)\right)$$
(4.2.6.11)
$$= f\left(a\left(\frac{1}{a}x - \frac{b}{a}\right) + b\right)$$
(4.2.6.12)
$$= f(x)$$
(4.2.6.13)

Similarly,

$$G'(G(f)) = G'(f(ax+b))$$
 (4.2.6.14)
= $f\left(\frac{1}{a}(ax+b) - \frac{b}{a}\right)$ (4.2.6.15)
= $f(x)$ (4.2.6.16)

Thus, G' is inverse of G. Therefore, G is isomorphism and we can say:

$$deg(h) = 1 \implies G \text{ is isomorphism.}$$

$$(4.2.6.17)$$

Let deg(h) > 1, then

$$\deg f(h) = \deg f \cdot \deg h$$
 (4.2.6.18)

$$\implies \deg f(h) \ge 1 \tag{4.2.6.19}$$

$$\implies G(f) = f(h) \neq x$$
 (4.2.6.20)

This means the image of G does not contain polynomials of degree one. Hence G is not onto and therefore G can not be an isomorphism. Thus we can write:

$$deg(h) > 1 \implies G \text{ is not isomorphism.}$$

$$(4.2.6.21)$$

From (4.2.6.17) and (4.2.6.21), We can conclude:

G is isomorphism.
$$\iff$$
 deg(h) = 1 (4.2.6.22)

- Let G' be the linear transformation from F(x) 4.3.1. Let $\mathbb Q$ be the field of rational numbers. Determine if the following subset of $\mathbb{Q}[x]$ is ideal or
 - a) The subset is defined by all f with degree ≥ 5

Solution: See Table 4.3.1.1

4.3.2. a) Find the g.c.d of each of the following pairs of polynomials.

$$2x^5 - x^3 - 3x^2 - 6x + 4, x^4 + x^3 - x^2 - 2x - 2$$
(4.3.2.1)

Solution: Refer Table 4.3.2.1.

b) Let Q be the field of rational numbers. Determine which of the following subsets of Q[x] are ideals. When the set is an ideal, find its monic generator.

All f such that f(2) = f(4) = 0.

Solution: *Definition*: Let *F* be a field. An ideal in F[x] is a subspace M of F[x] such that fg belongs to M whenever f is in F[x]and g is in M.

See Table 4.3.2.2

4.3.3. Let A be an $n \times n$ matrix over a field **F**. Show that the set of all polynomials f in F[x] such that f(A) = 0 is an ideal.

> **Solution:** Given a square matrix of order n and f(A) = 0. Let **I** be an ideal of $C[z_1, \dots, z_n]$.

$$\mathbf{I} = \{ \mathbf{F} \in \mathbf{F}(x) | f(A) = 0 \}$$
 (4.3.3.1)

Now, Consider the polynomials,

$$f = f_0 + f_1 z_n + \dots + f_d z_n^{\ d}$$
 (4.3.3.2)

$$g = g_0 + g_1 z_n + \cdots g_e z_n^e$$
 (4.3.3.3)

$$f,g \in \mathbf{I} \& k \in \mathbf{F}$$
.

$$(kf + g)(A) = kf(A) + g(A)$$
 (4.3.3.4)

$$\implies c.0 + 0 = 0$$
 (4.3.3.5)

From equation(4.3.3.5) **I** is a subspace of $\mathbf{F}(x)$ Now, $f \in \mathbf{I}$ and $g \in F(x)$ then,

$$(gf)A = g(A)f(A)$$
 (4.3.3.6)

$$\implies g(A).0 = 0$$
 (4.3.3.7)

Hence, proved **I** is ideal

4.3.4. Let **F** be a subfield of complex numbers, and let

$$\mathbf{A} = \begin{pmatrix} 1 & -2 \\ 0 & 3 \end{pmatrix} \tag{4.3.4.1}$$

5.1 Characteristic Values Find the monic generator of the ideal of all polynomials f in F[z] such that $f(\mathbf{A}) = 0$.

 $f(\mathbf{A}) = 0$. where,

$$f(x) = \sum_{i=0}^{n} a_i x^i \quad a_n = 1$$
 (4.3.4.2)

Computing $f(\mathbf{A})$ for $deg(f) \leq 1$ we get,

$$\mathbf{A}^0 = \mathbf{I} \qquad (4.3.4.3)$$

$$\mathbf{A}^1 = \begin{pmatrix} 1 & -2 \\ 0 & 3 \end{pmatrix} \tag{4.3.4.4}$$

$$f(\mathbf{A}) = a_0 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + a_1 \begin{pmatrix} 1 & -2 \\ 0 & 3 \end{pmatrix}$$
 (4.3.4.5)

As I and A are linearly independent so from (4.3.4.5) we see f(A) = 0 only if $a_1 = 0$ and $a_0 = 0$ but this can't happen from (4.3.4.2). Hence for $deg(f) \le 1$, $f(\mathbf{A}) \ne 0$. Hence for any $f(x) \in I$ such that deg(f) = 2 then f is Monic generator or minimal polymial. We can write.

$$f(x) = x^2 + a_1 x^1 + a_0 (4.3.4.6)$$

Minimal polynomial or monic generator can be

found using Characteristic equation,

$$\left|\mathbf{A} - \mathbf{I}\lambda\right| = 0 \tag{4.3.4.7}$$

$$\begin{vmatrix} 1 - \lambda & -2 \\ 0 & 3 - \lambda \end{vmatrix} = 0 \tag{4.3.4.8}$$

$$(1 - \lambda)(3 - \lambda) = 0 (4.3.4.9)$$

$$\implies \lambda^2 - 4\lambda + 3 = 0 \tag{4.3.4.10}$$

comparing (4.3.4.6) and (4.3.4.10) we get,

$$f(x) = x^2 - 4x + 3 = 0 (4.3.4.11)$$

using Cayley-Hamilton equation,

$$f(\mathbf{A}) = \mathbf{A}^2 - 4\mathbf{A} + 3 = 0 \tag{4.3.4.12}$$

Hence, $f = x^2 - 4x + 3$ is the monic generator.

5 ELEMENTARY CANONICAL FORMS

Solution: let, ideal $I = f(x) \in F[x]$ such that 5.1.1. Let T be the linear operator on \mathbb{R}^3 which is represented in the standard ordered basis by the matrix

$$\begin{pmatrix} -9 & 4 & 4 \\ -8 & 3 & 4 \\ -16 & 8 & 7 \end{pmatrix} \tag{5.1.1.1}$$

Prove that T is diagonalizable by exhibiting a basis for \mathbb{R}^3 , each vector of which is a characteristic vector of T.

Solution: Let **T** be a linear operator on a finite-dimensional space V. Let c_1, \dots, c_k be the distinct characteristic values of T and let W_i be the null space of $\mathbf{T} - c_i \mathbf{I}$. The following are equivalent.

- (i) **T** is diagonalizable.
- (ii) The characteristic polynomial for **T** is,

$$f = (x - c_1)_1^d \cdots (x - c_k)^{d_k}$$
 (5.1.1.2)

and dim $W_i = d_i, i = 1, \dots, k$

(iii) $\dim W_1 + \cdots + \dim W_k = \dim V$

Now let,

$$\mathbf{A} = \begin{pmatrix} -9 & 4 & 4 \\ -8 & 3 & 4 \\ -16 & 8 & 7 \end{pmatrix} \tag{5.1.1.3}$$

Solving $|\lambda \mathbf{I} - \mathbf{A}| = 0$

$$\begin{vmatrix} \lambda \mathbf{I} - \mathbf{A} \end{vmatrix} = \begin{vmatrix} \lambda + 9 & -4 & -4 \\ 8 & \lambda - 3 & -4 \\ 16 & -8 & \lambda - 7 \end{vmatrix}$$
 (5.1.1.4)

$$|\lambda \mathbf{I} - \mathbf{A}| = (\lambda + 1) \begin{vmatrix} \lambda + 9 & 0 & -4 \\ 8 & 1 & -4 \\ 16 & -1 & \lambda - 7 \end{vmatrix}$$

$$(5.1.1.6)$$

$$\stackrel{R_3 \leftarrow R_3 + R_2}{\longleftrightarrow} (\lambda + 1) \begin{vmatrix} \lambda + 9 & 0 & -4 \\ 8 & 1 & -4 \\ 24 & 0 & \lambda - 11 \end{vmatrix}$$

$$(5.1.1.7)$$

$$\implies (\lambda + 1) \begin{vmatrix} \lambda + 9 & -4 \\ 24 & \lambda - 11 \end{vmatrix}$$
(5.1.1.8)

$$\implies |\lambda \mathbf{I} - \mathbf{A}| = (\lambda + 1)^2 (\lambda - 3) = 0$$

$$(5.1.1.9)$$

$$\implies \lambda_1 = -1, \lambda_2 = -1, \lambda_3 = 3$$

Now at λ_1 and λ_3 ,

$$\mathbf{A} + \mathbf{I} = \begin{pmatrix} -8 & 4 & 4 \\ -8 & 4 & 4 \\ -16 & 8 & 8 \end{pmatrix}$$
 (5.1.1.11)

$$\mathbf{A} - 3\mathbf{I} = \begin{pmatrix} -12 & 4 & 4 \\ -8 & 0 & 4 \\ -16 & 8 & 4 \end{pmatrix}$$
 (5.1.1.12)

Now we know that A - 3I is singular and rank $(A - 3I) \ge 2$. Therefore, rank(A - 3I) = 2. Hence from the theorem-1 (iii) it is evident that rank(A + I) = 1. Let X_1 and X_3 be the spaces of characteristic vectors associated with the characteristic values 1 and 3 respectively. We know from rank nullity theorem that dim $X_1 = 2$ and dim $X_3 = 1$. Hence by Theorem-2 (i) T is diagonalizable.

The null-space of T + I is spanned by the vectors,

$$\alpha_1 = \begin{pmatrix} 1 & 0 & 2 \end{pmatrix} \tag{5.1.1.13}$$

$$\alpha_2 = \begin{pmatrix} 1 & 2 & 0 \end{pmatrix} \tag{5.1.1.14}$$

As both α_1 and α_2 are independent, hence they form basis for X_1 . The null-space of $\mathbf{T} - 3\mathbf{I}$ is spanned by the vector,

$$\alpha_3 = \begin{pmatrix} 1 & 1 & 2 \end{pmatrix} \tag{5.1.1.15}$$

Here α_3 is a characteristic vector and a basis for \mathbf{X}_3 . Also the matrix P which enables us to change coordinates from the basis β to the standard basis is the matrix which has transposes of α_1, α_2 and α_3 as it's column vectors:

$$\mathbf{P} = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 2 & 1 \\ 2 & 0 & 2 \end{pmatrix} \tag{5.1.1.16}$$

6 THE RATIONAL AND JORDAN FORMS

6.1 The Jordan Form

6.1.1. If A is a complex 5×5 matrix with characteristic polynomial $f = (x - 2)^3(x + 7)^2$ and minimal polynomial $p = (x - 2)^2(x + 7)$, what is the Jordan form for A?

Solution: *Theory:* Table 6.1.1.1 gives the overview of properties of a Jordan block based on characteristic polynomial, minimal polynomial, algebraic multiplicity and geometric multiplicity.

From the properties stated in table 6.1.1.1, the Jordan blocks for eigenvalues of **A** can be written as,

$$\mathbf{J_1} = \begin{pmatrix} 2 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix}; \quad \mathbf{J_2} = \begin{pmatrix} -7 & 0 \\ 0 & -7 \end{pmatrix} \quad (6.1.1.1)$$

Where J_1 and J_1 are the Jordan blocks corresponding to $\lambda_1 = 2$ and $\lambda_1 = -7$ respectively. The Jordan form for **A** can be written as,

$$\mathbf{J} = \begin{pmatrix} \mathbf{J_1} & 0 \\ 0 & \mathbf{J_2} \end{pmatrix} = \begin{pmatrix} 2 & 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & -7 & 0 \\ 0 & 0 & 0 & 0 & -7 \end{pmatrix}$$
 (6.1.1.2)

Inference An $n \times n$ matrix with with λ as diagonal elements, ones on the super diagonal and zeroes in all other entries is nilpotent with minimal polynomial $(A - \lambda I)^n$ Example Let,

$$\mathbf{A} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \tag{6.1.1.3}$$

- (6.1.1.3) is nilpotent for minimal polynomial A^3
- 6.1.2. The differentiation operator on the space of polynomials of degree less than or equal to 3 is represented in the natural ordered basis by the matrix,

$$\mathbf{A} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix} \tag{6.1.2.1}$$

What is the Jordan form of this matrix? (\mathbb{F} a subfield of the complex numbers.)

Solution: First, we find the characteristic polynomial of **A**,

$$\det\left(\mathbf{A} - \lambda \mathbf{I}\right) = 0 \qquad (6.1.2.2)$$

$$\Rightarrow \begin{vmatrix} -\lambda & 1 & 0 & 0 \\ 0 & -\lambda & 2 & 0 \\ 0 & 0 & -\lambda & 3 \\ 0 & 0 & 0 & -\lambda \end{vmatrix} = 0 \qquad (6.1.2.3)$$
$$\Rightarrow \lambda^4 = 0 \qquad (6.1.2.4)$$

(6.1.2.4) is the required characteristic equation and $\lambda_1 = 0$ is the only eigen value of **A**. Hence the characteristic polynomial of **A** is,

$$f(\lambda) = \lambda^4 \tag{6.1.2.5}$$

Again we observe that for k = 4 we have,

$$\implies f(\mathbf{A}) = 0 \tag{6.1.2.7}$$

And for k = 3 we also have,

$$\implies \mathbf{A}^3 \neq 0 \tag{6.1.2.9}$$

From (6.1.2.7) and (6.1.2.9) we conclude that the minimal polynomial of **A** is,

$$g(\lambda) = \lambda^4 \tag{6.1.2.10}$$

Hence, the Jordan form of **A** is a 4×4 matrix consisting of only one block with principal diagonal values as $\lambda_1 = 0$ and super diagonal

of the matrix (i.e the set of elements that lies directly above the elements comprising the principal diagonal) contains 1. Hence the required Jordan form of $\bf A$ is,

$$\mathbf{J} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix} \tag{6.1.2.11}$$

(6.1.2.11) is the required Jordan form of A.

Given	$x_1 + \ldots + x_n = 0$ $(x_1, \ldots, x_n) \in \mathbf{W}$ F is a field \mathbf{W}^* is dual space of \mathbf{W}
To prove	$\mathbf{W} \rightarrow \mathbf{W}^*$ is a natural isomorphism $f(x_1, \dots, x_n) = c_1 x_1 + \dots c_n x_n$ which satisfy $c_1 + \dots + c_n = 0$
Proof	Let $\alpha_i = \epsilon_1 - \epsilon_{i+1}$ $i \in \{1, \dots, n-1\}$ $\sum_{i=1}^{n-1} c_i \alpha_i = 0$ $\Longrightarrow \left(\sum_{i=1}^{n-1} c_i\right) \epsilon_1 - \sum_{i=1}^{n-1} c_i \epsilon_{i+1} = 0$ $(\alpha_1, \dots, \alpha_{n-1})$ are linearly independent and form a basis for W
	$\mathbf{W} \xrightarrow{P} (\mathbf{F}^n)^* \xrightarrow{Q} \mathbf{W}^*$ The function P is defined as $P(c_1, \dots, c_n) = f_{c_1, \dots, c_n}; \text{ where,}$ $f_{c_1, \dots, c_n}(x_1, \dots, x_n) = c_1 x_1 + \dots c_n x_n$
	Let $Q \circ P(c_1,, c_n) = 0$; $(c_1,, c_n) \in \mathbf{W}$ $Q(f_{c_1,, c_n}) = 0 \implies f_{c_1,, c_n \mid \mathbf{W}} = 0$ $\implies f_{c_1,, c_n}(x_1,, x_n) = 0$
	$f_{c_1,\dots,c_n}(\alpha_i) = 0; i = 1,\dots,n-1$ $\implies c_1 = c_i; i = 2,\dots,n$ $\implies \sum_{i=2}^n c_i = (n-1)c_1$ since $(c_1,\dots,c_n) \in \mathbf{W}$
	$\sum_{i=1}^{n} c_i = 0$ $\implies c_1 = 0$ $\implies c_i = 0 ; i = 1, \dots, n$
	Hence, $f_{c_1,,c_n}$ is a zero function. Thus the mapping $\mathbf{W} \to \mathbf{W}^*$ is a natural isomorphism

PARAMETERS	DESCRIPTION
V	Finite dimensional vector
	space with $dim(\mathbf{V}) = n$
W	Subspace of V
\mathbf{W}^0	Annihilator of W
$g \in \mathbf{W}^0$	$g(\mathbf{w}) = 0 \forall \mathbf{w} \in \mathbf{W}$
$\{g_1,g_2,\ldots,g_r\}$	basis for \mathbf{W}^0
\mathbf{N}_{g_i}	Null space of g_i

TABLE 3.6.2.1: Input Parameters

TABLE 3.6.1.1

STATEMENT	DESCRIPTION	
$g \in \mathbf{W}^0$	$g = \sum_{i=1}^{r} c_i g_i$	(3.6.2.2)
	From the theorem,	
If N is nullspace of g	$\bigcap_{i=1}^r \mathbf{N}_{g_i} \subseteq \mathbf{N}$	(3.6.2.3)
$\forall \mathbf{w} \in \mathbf{W}$	$g(\mathbf{w}) = 0$	(3.6.2.4)
$(3.6.2.3), (3.6.2.4) \implies$	$\mathbf{W} \subseteq \bigcap_{i=1}^r \mathbf{N}_{g_i}$	(3.6.2.5)
Starting with method of contradiction	$\mathbf{W} \neq \bigcap_{i=1}^{r} \mathbf{N}_{g_i}$	(3.6.2.6)
From (3.6.2.5),(3.6.2.6) there exists		
a vector e such that	$\mathbf{e} \in \bigcap_{i=1}^{r} \mathbf{N}_{g_i}, \mathbf{e} \notin \mathbf{W}$	(3.6.2.7)
	So $\forall g \in \mathbf{W}^0, g(\mathbf{e}) = 0$	(3.6.2.8)
Since g is a linear functional on \mathbf{V}	$(3.6.2.8) \implies \mathbf{e} \in \mathbf{V}$	(3.6.2.9)
	A linear functional on V defined as	
f	$f(\alpha) \begin{cases} = 0 & \text{for } \alpha \in \mathbf{W} \\ \neq 0 & \text{for } \alpha \in \mathbf{V}, \alpha \notin \mathbf{W} \end{cases}$ $\therefore f \in \mathbf{W}^0 \text{ and } f(e) \neq 0$	(3.6.2.10) (3.6.2.11)
(3.6.2.8),(3.6.2.11) contradict each other	$\mathbf{W} = \bigcap_{i=1}^{r} \mathbf{N}_{g_i}$	(3.6.2.12)

TABLE 3.6.2.2: Proof

Given	$S \text{ is a set}$ $F \text{ is a field}$ $V(S, F) \text{ is a linear functional}$ such that $W \text{ be } n\text{-dim subspace of } V(S, F).$ $Also, \delta_{ij} = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$
To prove	$f_i(x_j) = \delta_{ij}$
	where $x_1, x_2, \dots, x_n \in S$ and $f_1, f_2, \dots, f_n \in W$
Proof	Let $\phi_x:W\to F$
	Suppose $\phi_x(f) = 0 \ \forall x \in S \ \& \ f \in W$ $\implies f(x) = 0$
	If $\forall x, \ \phi_x(f) \neq 0$ for some $f \in W$ If $n > 0 \ \exists \in S$ such that $\phi_x(f) \neq 0$ for some $f \in W$ $\implies f_1(x_1) \neq 0$
	By scaling we can have $f_1(x_1) = 1$
	Hence $f_i(x_j) = \delta_{ij}$

Properties Used		
SVD of matrix A	$\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$	
U is unitary	$\mathbf{U}^T\mathbf{U} = \mathbf{I}$	
V is unitary	$\mathbf{V}^T\mathbf{V} = \mathbf{I}$	
Σ is diagonal	$\mathbf{\Sigma}^T \mathbf{\Sigma} = \mathbf{\Sigma}^2$	
Cyclic property	$tr\left(\mathbf{ABC}\right) = tr\left(\mathbf{CAB}\right)$	
Rank (A)	$Rank(\mathbf{A}) = #non-zero singular values$	

TABLE 3.7.4.1: Properties

Statement	Proof	
	$tr\left(\mathbf{A}^{T}\mathbf{A}\right) = tr\left(\mathbf{V}\mathbf{\Sigma}^{T}\mathbf{U}^{T}\mathbf{U}\mathbf{\Sigma}\mathbf{V}^{T}\right)$	(3.7.4.1)
	$= tr(\mathbf{V}\mathbf{\Sigma}^T\mathbf{\Sigma}\mathbf{V}^T)$	(3.7.4.2)
	$= tr\left(\mathbf{V}\mathbf{\Sigma}^2\mathbf{V}^T\right)$	(3.7.4.3)
	$= tr\left(\mathbf{V}^T\mathbf{V}\mathbf{\Sigma}^2\right)$	(3.7.4.4)
	$=tr(\Sigma^2)$	(3.7.4.5)
$tr(\mathbf{A}^T\mathbf{A}) = 0 \implies \mathbf{A} = 0$	$=\sum_{i=1}^{r}\sigma_{i}^{2}$	(3.7.4.6)
	$tr\left(\mathbf{A}^{T}\mathbf{A}\right) = 0$	(3.7.4.7)
	$\implies \sum_{i=1}^{r} \sigma_i^2 = 0 \ \forall i = 1, 2, \dots r$	(3.7.4.8)
	$\implies \sigma_i = 0 \ \forall i = 1, 2, \dots r$	(3.7.4.9)
	$\therefore Rank(\mathbf{A}) = #non-zero singular values = 0$	(3.7.4.10)
	$\implies \mathbf{A} = 0$	(3.7.4.11)
	A = 0	(3.7.4.12)
$\mathbf{A} = 0 \implies tr(\mathbf{A}^T \mathbf{A}) = 0$	$\implies \mathbf{A}^T \mathbf{A} = 0$	(3.7.4.13)
	$\implies tr(\mathbf{A}^T\mathbf{A}) = 0$	(3.7.4.14)

TABLE 3.7.4.2: Proofs

PARAMETERS	DESCRIPTION
\mathbb{R}	Field of real numbers
V	Vector space of all polynomi-
	als over \mathbb{R}
a, b	Fixed real numbers
f	Linear functional on V
D	Differentiator operator on V
D^t	Transpose of D

TABLE 3.7.5.1: Input Parameters

STATEMENTS	DERIVATIONS	
D^t is transpose of D	$(D^t f)(p) = f[D(p)]$	(3.7.5.2)
A polynomial of degree n		
$p(x) = c_0 + c_1 x + c_2 x^2 + \ldots + c_n x^n$	$\mathbf{c} = \begin{pmatrix} c_0 \\ c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix}, \mathbf{x} = \begin{pmatrix} 1 \\ x \\ x^2 \\ \vdots \\ x^n \end{pmatrix}$	(3.7.5.4)
	$f(p) = \mathbf{c}^T \mathbf{F}$	(3.7.5.5)
$f(p) = \int_{a}^{b} p(x) dx$	$\mathbf{F} = \begin{pmatrix} b - a \\ \frac{b^2 - a^2}{2} \\ \frac{b^3 - a^3}{3} \\ \vdots \\ \frac{b^{n+1} - a^{n+1}}{n+1} \end{pmatrix}$	(3.7.5.6)
	$D(p) = \mathbf{c}^T \mathbf{D} \mathbf{x}$	(3.7.5.7)
$D(p) = c_1 + 2c_2x + 3c_3x^2 + \dots + nc_nx^{n-1}$	$\mathbf{D} = \begin{pmatrix} 0 & 0 & 0 & \dots & 0 & 0 \\ 1 & 0 & 0 & \dots & 0 & 0 \\ 0 & 2 & 0 & \dots & 0 & 0 \\ 0 & 0 & 3 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \dots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & n & 0 \end{pmatrix}$	(3.7.5.8)
	Let $D(p) = p'$	(3.7.5.9)
	$\implies p' = \mathbf{c}'^T \mathbf{x}$	(3.7.5.10)
$f[D(p)] = c_1(b-a) + c_2(b^2 - a^2) + \dots + c_n(b^n - a^n)$	where $\mathbf{c}'^T = \begin{pmatrix} c_1 \\ 2c_2 \\ 3c_3 \\ \vdots \\ nc_n \\ 0 \end{pmatrix}$	(3.7.5.11)
	$f[D(p)] = \mathbf{c}'^T \mathbf{F}$	(3.7.5.12)
$(D^t f)(p) = p(b) - p(a)$	$(D^{t}f)(p) = c_{1}(b-a) + c_{2}(b^{2} - a^{2}) + \dots + c_{n}(b^{n} - a^{n}) + c_{0} - c_{0}$ $= (c_{0} + c_{1}b + c_{2}b^{2} + \dots + c_{n}b^{n}) - (c_{0} + c_{1}a + c_{2}a^{2} + \dots + c_{n}a^{n})$ $= p(b) - p(a)$	(3.7.5.13) (3.7.5.14) (3.7.5.15) (3.7.5.16) (3.7.5.17)

TABLE 3.7.5.2: Proof

iso- morphism	A linear transformation $T: V \to W$ is isomorphism of V onto W if T is one – one
one-one	A linear transfrmation $T: \mathbb{R}^n \to \mathbb{R}^m$ is <i>one-one</i> if $\forall \mathbf{b} \in \mathbb{R}^m$, $\mathbf{AX} = \mathbf{b}$ has at most one solution in \mathbb{R}^n By definition, all <i>invertible</i> linear transformations are <i>one-one</i>
invertible	A linear transformation $T: V \to W$ is <i>invertible</i> if there exists another linear transformation $U: W \to V$ such that UT is identity transformation on V and TU is identity transformation on W

Given	$\phi(T) = T' \text{ where}$ $\phi: L(V, V) \to L(V^*, V^*)$ $T \in L(V, V) \text{ and } T' \in L(V^*, V^*)$
To prove	ϕ is an <i>isomorphism</i> of $L(V, V)$ onto $L(V^*, V^*)$ i.e ϕ is one – one $\implies \phi$ is invertible
Proof	Consider a linear transformation $\psi: L(V^*, V^*) \to L((V^*)^*, (V^*)^*)$
	We know, $L((V^*)^*, (V^*)^*) = L(V, V)$ Hence $\psi : L(V^*, V^*) \rightarrow L(V, V)$
	We have $\phi \circ \psi = I$ and $\psi \circ \phi = I$
	$\therefore \phi$ is invertible with ψ its inverse.
	Hence ϕ is isomorphic.

TABLE 3.7.6.1: Definitions

TABLE 3.7.6.2: Solution

Given	V is a vector space of $n \times n$ matrices over the field F \mathbf{B} is a fixed $n \times n$ matrix A function f_B on V is defined such that $f_B(\mathbf{A}) = tr(\mathbf{B}^t \mathbf{A})$
To prove	f_B is a linear functional on V
Properties of Trace of Matrix	$1.tr(\mathbf{M} + \mathbf{N}) = tr(\mathbf{M}) + tr(\mathbf{N})$ $2. tr(c\mathbf{M}) = c(tr(\mathbf{M}))$ Where M and N are two square matrices
Proof	As $f_B(\mathbf{A}) = tr(\mathbf{B}^t \mathbf{A})$ $\Rightarrow f_B(c\mathbf{A}_1 + \mathbf{A}_2)$ $= tr(\mathbf{B}^t(c\mathbf{A}_1 + \mathbf{A}_2))$ $\Rightarrow f_B(c\mathbf{A}_1 + \mathbf{A}_2) = tr(c\mathbf{B}^t \mathbf{A}_1)$ $+tr(\mathbf{B}^t \mathbf{A}_2)$ $\Rightarrow f_B(c\mathbf{A}_1 + \mathbf{A}_2) = c(tr(\mathbf{B}^t \mathbf{A}_1))$ $+tr(\mathbf{B}^t \mathbf{A}_2)$ $\Rightarrow f_B(c\mathbf{A}_1 + \mathbf{A}_2)$ $= cf_B(\mathbf{A}_1) + f_B(\mathbf{A}_2)$

Given	S be a set of non-zero polynomial over a field F No two elements of S have the same degree
To prove	S is an independent set in F[x]
Linear Independency	Let $f_1, f_2,, f_n$ are the polynomials and they will be linearly independent if $a_1f_1 + a_2f_2 + + a_nf_n = \theta$ for $a_1 = a_2 = = a_n = 0$ where $a_1, a_2,, a_n$ are scalars from field F

TABLE 4.1.5.1

TABLE 3.7.7.1

Proof	Let the degrees of $f_1, f_2,, f_n$ are
11001	d_1, d_2, \dots, d_n respectively such that
	the degree of $f_i = d_i \neq d_i$
	for $j=1,2,,n$ and $i \neq j$
	and $d_1 < d_2 < < d_n$
	so d_n is the largest degree
	Now, let $a_1 f_1 + a_2 f_2 + + a_n f_n = \theta$
	, 131 232 131
	where $f_1 = \sum_{i=0}^{d_1} k_{1i} x^i$
	$J_{i=0}$
	$f_2 = \sum_{i=0}^{d_2} k_{2i} x^i$
	$\mathbf{or}, f_2 = \sum_{i=0}^{d_1} k_{2i} x^i + \sum_{i=d_1+1}^{d_2} k_{2i} x^i$
	Similarly $f_{n-1} = \sum_{i=1}^{d_{n-1}} k_{i-1} \cdot x^i$
	Similarly, $f_{n-1} = \sum_{i=0}^{d_{n-1}} k_{(n-1)i} x^i$ $\implies f_{n-1} = \sum_{i=0}^{d_{n-2}} k_{(n-1)i} x^i + \sum_{i=d_{n-2}+1}^{d_{n-1}} k_{(n-1)i} x^i$
	$\int \int $
	and $f = \sum_{n=1}^{d_{n-1}} k \cdot x^i + \sum_{n=1}^{d_n} k \cdot x^i$
	and $f_n = \sum_{i=0}^{d_{n-1}} k_{ni} x^i + \sum_{i=d_{n-1}+1}^{d_n} k_{ni} x^i$ Now
	$a_1 f_1 + a_2 f_2 + + a_n f_n$
	$= a_1 \sum_{i=0}^{d_1} k_{1i} x^i + a_2 \left(\sum_{i=0}^{d_1} k_{2i} x^i + \sum_{i=d_1+1}^{d_2} k_{2i} x^i \right)$
	$+ \dots + a_n \left(\sum_{i=0}^{d_1} k_{ni} x^i + \sum_{i=d_1+1}^{d_2} k_{ni} x^i + \dots \right)$
	$\dots + \sum_{i=d_{n-1}+1}^{d_n} k_{ni} x^i$
	∇^{d_1} (- 1 - 1 - 1 - 1 - 1 - 1 - 1
	$= \sum_{i=0}^{d_1} (a_1 k_{1i} + a_2 k_{2i} + + a_n k_{ni}) x^i$
	$+\sum_{i=d_1+1}^{d_2} (a_2 k_{2i} + + a_n k_{ni}) x^i$
	$++\sum_{i=d_{n-1}+1}^{d_n}a_nk_{ni}x^i$
	Now, as $a_1 f_1 + a_2 f_2 + + a_n f_n = \theta$
	for $d_{n-1} + 1 \le i \le d_n$, $k_{ni} \ne 0$
	so a_n must be 0
	Now, discarding a_n associated term, we get
	$\sum_{i=0}^{d_1} (a_1 k_{1i} + a_2 k_{2i} + + a_n k_{ni}) x^i$
	$+\sum_{i=d_1+1}^{d_2}(a_2k_{2i}++a_nk_{ni})x^i$
	$++\sum_{i=d_{n-2}+1}^{d_{n-1}}a_{n-1}k_{(n-1)i}x^{i}=0$
	so, for $d_{n-2} + 1 \le i \le d_{n-1}$, $k_{(n-1)i} \ne 0$
	$\implies a_{n-1} = 0$

Proof

Similarly, for $d_1 + 1 \le i \le d_2$ $\sum_{i=0}^{d_1} (a_1 k_{1i} + a_2 k_{2i}) x^i + \sum_{i=d_1+1}^{d_2} a_2 k_{2i} x^i$ $\implies a_2 = 0 \text{ as } k_{2i} \ne 0$ $\implies \sum_{i=0}^{d_1} a_1 k_{1i} x^i = 0$ $\implies a_1 = 0 \text{ as } k_{1i} \ne 0$ In this way, it can be proved that $a_1 = a_2 = \dots = a_{n-1} = a_n = 0 \text{ for } a_1 f_1 + a_2 f_2 + \dots + a_n f_n = \theta$ $\implies f_1, f_2, \dots, f_n \text{ are linearly independent}$

TABLE 4.1.5.3

TABLE 4.1.5.2

PARAMETER	DESCRIPTION
F	Field of complex numbers
\mathbb{F}^{∞}	Vector space defined on the field \mathbb{F}
$\mathbb{F}[x]$	Subspace of \mathbb{F}^{∞} spanned by $\{1, x, x^2, x^3, \ldots\}$
$T\colon \mathbb{F}[x]\to \mathbb{F}[x]$	Transformation T
$f = \sum_{i=0}^{n} c_i x^i$	Polynomial $f \in \mathbb{F}[x]$
	Polynomial $f' \in \mathbb{F}[x]$
	Scalars in \mathbb{F} and coefficients of polynomials f and f'
$T(f) = g = \sum_{i=0}^{n} \frac{c_i}{i+1} x^{i+1}$ $T(f') = g' = \sum_{i=0}^{n} \frac{c'_i}{i+1} x^{i+1}$	Transformed polynomial $g \in \mathbb{F}[x]$
$T(f') = g' = \sum_{i=0}^{n} \frac{c'_i}{i+1} x^{i+1}$	Transformed polynomial $g' \in \mathbb{F}[x]$
\mathbf{M}_T	Transformation matrix for T
N(T)	Null Space of T

TABLE 4.1.6.1: Parameters

Statement	Derivation	
$f = \sum_{i=0}^{n} c_i x^i$	$f = \begin{pmatrix} c_0 & c_1 & c_2 & \cdots & c_n \end{pmatrix}_{(n+1)\times 1}^T$	
$T\left[f\right] = \mathbf{M}_T f$	$T[f] = \begin{pmatrix} c_0 & c_1 & c_2 & \cdots & c_n \end{pmatrix}_{(n+1)\times 1}$ $T[f] = \begin{pmatrix} 0 & 0 & 0 & \cdots & 0 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & \frac{1}{2} & 0 & \cdots & 0 \\ 0 & 0 & \frac{1}{3} & \cdots & 0 \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & 0 & \cdots & \frac{1}{n+1} \end{pmatrix}_{(n+2)\times(n+1)} \begin{pmatrix} c_0 \\ c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix}_{(n+1)\times 1} = \begin{pmatrix} 0 \\ c_0 \\ \frac{c_1}{2} \\ \frac{c_2}{3} \\ \vdots \\ \frac{c_n}{n+1} \end{pmatrix}_{(n+2)\times 1} = g \in \mathbb{F}[x]$	
	$T\left[\alpha f + f'\right] \tag{4.1.6.2}$	
	$= \mathbf{M}_T \left(\alpha f + f' \right) \tag{4.1.6.3}$	
	$= \begin{pmatrix} 0 & 0 & 0 & \cdots & 0 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & \frac{1}{2} & 0 & \cdots & 0 \\ 0 & 0 & \frac{1}{3} & \cdots & 0 \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & 0 & \cdots & \frac{1}{n+1} \end{pmatrix} \begin{pmatrix} \alpha c_0 + c'_0 \\ \alpha c_1 + c'_1 \\ \alpha c_2 + c'_2 \\ \vdots \\ \alpha c_n + c'_n \end{pmatrix} $ (4.1.6.4)	
T is a linear operator	$= \begin{pmatrix} 0 \\ \alpha c_0 + c'_0 \\ \frac{\alpha c_1 + c'_1}{2} \\ \frac{\alpha c_2 + c'_2}{3} \\ \vdots \\ \frac{\alpha c_n + c'_n}{n+1} \end{pmatrix} = \alpha \begin{pmatrix} 0 \\ c_0 \\ \frac{c_1}{2} \\ \frac{c_2}{3} \\ \vdots \\ \frac{c_n}{n+1} \end{pmatrix} + \begin{pmatrix} 0 \\ c'_0 \\ \frac{c'_1}{2} \\ \frac{c'_2}{3} \\ \vdots \\ \frac{c'_n}{n+1} \end{pmatrix} $ (4.1.6.5)	
	$= \alpha T[f] + T[f'] $ (4.1.6.6)	
	$= \alpha g + g' \tag{4.1.6.7}$	
	$\therefore T\left[\alpha f + f'\right] = \alpha T\left[f\right] + T\left[f'\right] \tag{4.1.6.8}$	
	T[f] = 0 (4.1.6.9)	
T is non-singular	$\implies \mathbf{M}_T f = 0 \implies f = 0 :: \mathbf{M}_T \neq 0$ (4.1.6.10)	
	$\implies N(T) = \{0\}$ (4.1.6.11)	

TABLE 4.1.6.2: Proof for Non-Singular and linear transformation T

Statement	Derivation	
T is not invertible	As \mathbf{M}_T is a non-square matrix with dimensions $(n+2) \times (n+1)$, the transformation T is not invertible	
\mathbf{M}_D is left inverse of \mathbf{M}_T	$(n+1), \text{ the transformation T is not invertible}$ $\mathbf{M}_{D}\mathbf{M}_{T} = \mathbf{I}_{n+1} \qquad (4.1.6.12)$ $\Rightarrow \mathbf{M}_{D} = \begin{pmatrix} 0 & 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 2 & 0 & \cdots & 0 \\ 0 & 0 & 0 & 3 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & n+1 \end{pmatrix} \qquad (4.1.6.13)$	

TABLE 4.1.6.3: Non-Invertibility of transformation T

Linear Transformation	A linear transformation from V into W is a function T from V into W such that $\mathbf{T}(c\alpha + \beta) = c\mathbf{T}(\alpha) + \mathbf{T}(\beta)$ $\forall \alpha \text{ and } \beta \text{ in } \mathbf{V} \text{ and } \forall \text{ scalars } c \text{ in } \mathbf{F}.$
$\mathbf{F}[x]$	Let $\mathbf{F}[x]$ be the subspace of \mathbf{F}^{∞} spanned by the vectors $1, x, x^2,$ An element of $\mathbf{F}[x]$ is called a polynomial over \mathbf{F} .
Differentiation Transformation	Let F be a field and let V be the space of polynomial functions f from F into F , given by $f(x) = c_0 + c_1 x + + c_k x^k$ Then, $\mathbf{D} f(x) = c_1 + 2c_2 x + + kc_k x^{k-1}$ is called Differentiation Transformation. The Differentiation Transformation is a Linear map because $\mathbf{D} (cf + g)(x) = \left(c.c_1 + c_1'\right) + 2\left(c.c_2 + c_2'\right)x + + k\left(c.c_k + c_k'\right)x^{k-1}$ $= c.c_1 + 2c.c_2 x + + kc.c_k x^{k-1} + c_1' + 2c_2' x + + kc_k' x^{k-1}$ $= c\mathbf{D} f(x) + \mathbf{D} g(x)$

TABLE 4.1.6.4

	Proving D is Linear	From (4.1.6.14), clearly \mathbf{D} is a function from $\mathbf{F}[\mathbf{x}]$ to $\mathbf{F}[\mathbf{x}]$. We must show that \mathbf{D} is linear. Clearly \mathbf{D} is a Differentiation Transformation, and hence is linear. In other words	
		Let $\mathbf{m} = \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_n \end{pmatrix}$, $\mathbf{n} = \begin{pmatrix} c'_0 \\ c'_1 \\ \vdots \\ c'_n \end{pmatrix}$ and α be a scalar. Then	
		$\mathbf{D}\left(\mathbf{x}^{T}(\alpha\mathbf{m}+\mathbf{n})\right) = \mathbf{y}^{T}\mathbf{p}, \text{ where } \mathbf{p} = \begin{pmatrix} \alpha c_1 + c_1' \\ 2\left(\alpha c_2 + c_2'\right) \\ \vdots \\ n\left(\alpha c_n + c_n'\right) \end{pmatrix} \text{ and } \alpha\mathbf{m} + \mathbf{n} = \begin{pmatrix} \alpha c_0 + c_0' \\ \alpha c_1 + c_1' \\ \vdots \\ \alpha c_n + c_n' \end{pmatrix}$ $\mathbf{D}\left(\mathbf{x}^{T}(\alpha\mathbf{m}+\mathbf{n})\right) = \mathbf{y}^{T}(\mathbf{m}'+\mathbf{n}'), \text{ where } \mathbf{m}' = \alpha \begin{pmatrix} c_1 \\ 2c_2 \\ \vdots \\ nc_n \end{pmatrix} \text{ and } \mathbf{n}' = \begin{pmatrix} c_1' \\ 2c_2' \\ \vdots \\ nc_n' \end{pmatrix}$	
		$\mathbf{D}(\mathbf{x}^{\mathbf{T}}(\alpha\mathbf{m} + \mathbf{n})) = \mathbf{y}^{\mathbf{T}}(\mathbf{m}' + \mathbf{n}'), \text{ where } \mathbf{m}' = \alpha \begin{pmatrix} c_1 \\ 2c_2 \\ \vdots \\ nc_n \end{pmatrix} \text{ and } \mathbf{n}' = \begin{pmatrix} c'_1 \\ 2c'_2 \\ \vdots \\ nc'_n \end{pmatrix}$	
		Thus,	
		Now, $\mathbf{D}(\mathbf{x}^{T}(\alpha \mathbf{m} + \mathbf{n})) = \mathbf{y}^{T}\mathbf{m}' + \mathbf{y}^{T}\mathbf{n}'$ $\mathbf{D}(\mathbf{x}^{T}(\alpha \mathbf{m} + \mathbf{n})) = \alpha \mathbf{D}(\mathbf{x}^{T}\mathbf{m}) + \mathbf{D}(\mathbf{x}^{T}\mathbf{n})$	
		Hence, D is a linear transformation.	
A p		Let $\mathbf{N}(\mathbf{D})$ denotes the nullspace of \mathbf{D} . Then $\mathbf{N}(\mathbf{D}) = \{f \in \mathbf{F}[\mathbf{x}] : \mathbf{D}f(x) = 0\}$ A polynomial is zero if and only if its every coeficient is zero. Thus, it must be such that each $c_1 = c_2 = \dots = 0$. Since, \mathbf{D} is a Differentiation Transformation and we know that derivative of a constant polynomial is zero. Thus, the nullspace of \mathbf{D} contains the constant polynomials. Hence,	

TABLE 4.1.6.5

 $\mathbf{N}(\mathbf{D}) = \{ f \in \mathbf{F}[\mathbf{x}] : f(x) = c \}$

Example The defined subset of $\mathbb{Q}[x]$ be \mathbb{U} , $f(x) = c_1 x^5 + c_2 x^4 \in \mathbb{U}$ $g(x) = -c_1 x^5 + c_3 x^4 \in \mathbb{U}$ If \mathbb{U} is an ideal then, \mathbb{U} must be a subset. \mathbb{U} must be closed under addition. $f \in \mathbb{U}$ $g \in \mathbb{U}$ Proof $\Rightarrow f + g \in \mathbb{U}$ But here, $f + g = c_1 x^5 + c_2 x^4 - c_1 x^5 + c_3 x^4$ $\Rightarrow f + g = (c_2 + c_3) x^4 \notin \mathbb{U}$ Observation $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ $g \in \mathbb{U}$		
$g(x) = -c_1 x^5 + c_3 x^4 \in \mathbb{U}$ If \mathbb{U} is an ideal then, $\mathbb{U} \text{ must be a subset.}$ $\mathbb{U} \text{ must be closed under addition.}$ $f \in \mathbb{U}$ $g \in \mathbb{U}$ But here, $f + g = c_1 x^5 + c_2 x^4 - c_1 x^5 + c_3 x^4$ $\implies f + g = (c_2 + c_3) x^4 \notin \mathbb{U}$ Observation $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ $\text{But, } f + g \notin \mathbb{U}$ $\mathbb{U} \text{ is not closed under addition}$ $\implies \mathbb{U} \text{ is not a subset of } \mathbb{Q}[x]$,
If \mathbb{U} is an ideal then, \mathbb{U} must be a subset. \mathbb{U} must be closed under addition. $f \in \mathbb{U}$ $g \in \mathbb{U}$ Proof	Example	$f(x) = c_1 x^5 + c_2 x^4 \in \mathbb{U}$
		$g(x) = -c_1 x^5 + c_3 x^4 \in \mathbb{U}$
$f \in \mathbb{U}$ $g \in \mathbb{U}$		If U is an ideal then,
Proof $f \in \mathbb{U}$ $g \in \mathbb{U}$ $\Rightarrow f + g \in \mathbb{U}$ But here, $f + g = c_1 x^5 + c_2 x^4 - c_1 x^5 + c_3 x^4$ $\Rightarrow f + g = (c_2 + c_3) x^4 \notin \mathbb{U}$ Observation $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ $\text{But, } f + g \notin \mathbb{U}$ $\text{U is not closed under addition}$ $\Rightarrow \mathbb{U} \text{ is not a subset of } \mathbb{Q}[x]$		U must be a subset.
Proof $g \in \mathbb{U}$ $\Rightarrow f + g \in \mathbb{U}$ But here, $f + g = c_1 x^5 + c_2 x^4 - c_1 x^5 + c_3 x^4$ $\Rightarrow f + g = (c_2 + c_3) x^4 \notin \mathbb{U}$ Observation $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ U is not closed under addition $\Rightarrow \mathbb{U} \text{ is not a subset of } \mathbb{Q}[x]$		U must be closed under addition.
Proof		$f \in \mathbb{U}$
But here, $f + g = c_1 x^5 + c_2 x^4 - c_1 x^5 + c_3 x^4$ $\implies f + g = (c_2 + c_3) x^4 \notin \mathbb{U}$ Observation $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ U is not closed under addition $\implies \mathbb{U} \text{ is not a subset of } \mathbb{Q}[x]$		$g \in \mathbb{U}$
$f + g = c_1 x^5 + c_2 x^4 - c_1 x^5 + c_3 x^4$ $\implies f + g = (c_2 + c_3) x^4 \notin \mathbb{U}$ $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ $\mathbb{U} \text{ is not closed under addition}$ $\mathbb{U} \text{ is not a subset of } \mathbb{Q}[x]$	Proof	$\implies f + g \in \mathbb{U}$
$\Rightarrow f + g = (c_2 + c_3)x^4 \notin \mathbb{U}$ $f \in \mathbb{U}$ Observation $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ $\mathbb{U} \text{ is not closed under addition}$ Conclusion $\Rightarrow \mathbb{U} \text{ is not a subset of } \mathbb{Q}[x]$		But here,
$\begin{array}{ccc} & & f \in \mathbb{U} \\ & & g \in \mathbb{U} \\ & & \text{But, } f + g \notin \mathbb{U} \\ & & \mathbb{U} \text{ is not closed under addition} \\ & & \text{Conclusion} & \implies \mathbb{U} \text{ is not a subset of } \mathbb{Q}[x] \end{array}$		$f + g = c_1 x^5 + c_2 x^4 - c_1 x^5 + c_3 x^4$
Observation $g \in \mathbb{U}$ But, $f + g \notin \mathbb{U}$ U is not closed under addition Conclusion $\Longrightarrow \mathbb{U}$ is not a subset of $\mathbb{Q}[x]$		$\implies f + g = (c_2 + c_3)x^4 \notin \mathbb{U}$
But, $f + g \notin \mathbb{U}$ U is not closed under addition Conclusion $\Longrightarrow \mathbb{U}$ is not a subset of $\mathbb{Q}[x]$		$f \in \mathbb{U}$
$\begin{array}{c c} & \mathbb{U} \text{ is not closed under addition} \\ \text{Conclusion} & \Longrightarrow \mathbb{U} \text{ is not a subset of } \mathbb{Q}[x] \end{array}$	Observation	$g \in \mathbb{U}$
Conclusion $\implies \mathbb{U}$ is not a subset of $\mathbb{Q}[x]$		But, $f + g \notin \mathbb{U}$
		U is not closed under addition
$\implies \mathbb{U}$ is not an ideal of $\mathbb{Q}[x]$	Conclusion	$\implies \mathbb{U}$ is not a subset of $\mathbb{Q}[x]$
		$\implies \mathbb{U}$ is not an ideal of $\mathbb{Q}[x]$

TABLE 4.3.1.1

Let the field be rational numbers		
Steps	Explanation	
Say $f(x)$ and $g(x)$	$f(x) = 2x^5 - x^3 - 3x^2 - 6x + 4 $ (4.3.2.2) $g(x) = x^4 + x^3 - x^2 - 2x - 2 $ (4.3.2.3)	
Expanding $f(x)$ in term of $g(x)$	$2x^{5} - x^{3} - 3x^{2} - 6x + 4 = (x^{4} + x^{3} - x^{2} - 2x - 2)(2x - 2) + (3x^{3} - x^{2} - 6x)$ (4.3.2.4)	
Expanding degree four polynomial	$x^{4} + x^{3} - x^{2} - 2x - 2 = \left(3x^{3} - x^{2} - 6x\right)\left(\frac{1}{3}x + \frac{4}{9}\right) + \left(\frac{13}{9}x^{2} + \frac{2}{3}x - 2\right) $ (4.3.2.5)	
Expanding degree three polynomial	$3x^{3} - x^{2} - 6x = \left(\frac{13}{9}x^{2} + \frac{2}{3}x - 2\right)\left(\frac{27}{13}x - \frac{279}{169}\right) + \left(-\frac{126}{169}x - \frac{558}{169}\right) $ (4.3.2.6)	
Expanding degree two polynomial	$\frac{13}{9}x^2 + \frac{2}{3}x - 2 = \left(-\frac{126}{169}x - \frac{558}{169}\right)\left(-\frac{2197}{1134}x + \frac{61009}{7938}\right) + \left(\frac{10309}{441}\right)$ (4.3.2.7)	
	Since it contains scalar polynomial hence the g.c.d of $f(x)$, $g(x)$ is 1.	

TABLE 4.3.2.1: Solution

Given	Q be the field of rational numbers. Subset: All f such that $f(2) = f(4) = 0$	
To prove	Given subset is an Ideal.	
	If Ideal then find its monotic generator.	
Proof	Let, $M = \{ f \in F[x] f(2) = f(4) = 0 \}$. g(2) = g(4) = 0 and $f(2) = f(4) = 0$	
	(df + g)(2) = df(2) + g(2) = 0 $(df + g)(4) = df(4) + g(4) = 0$	
	$\implies M$ is a subspace of $F[x]$	
	Let suppose $f \in M$ and $g \in F[x]$. $(fg)(2) = f(2)g(2) = 0$ $(fg)(4) = f(4)g(4) = 0$ $\therefore fg \in M$	
	Hence, M is an ideal.	
Representation of polynomial	f(x) = (x - c)q(x) + r When $f(c) = 0$ then $r = 0$.	
	$\implies f(x) = (x-2)(x-4)q(x)$ $\therefore f(x) = g(x)q(x)$	
	Hence, $f(x)$ is generated by $g(x)$.	

TABLE 4.3.2.2

Feature	Effect on Jordan block	Example
characteristic	The multiplicity of λ	Let, $f = (x - 2)^4$ be characteristic polynomial $\begin{pmatrix} 2 & * & 0 & 0 \end{pmatrix}$
polynomial or	in the characteristic polynomial	$\mathbf{J} = \begin{pmatrix} 2 & * & 0 & 0 \\ 0 & 2 & * & 0 \\ 0 & 0 & 2 & * \\ 0 & 0 & 0 & 2 \end{pmatrix}$
Algebraic multiplicity	determines the size of the Jordan block for that eigenvalue. A_M = Size of Jordan block for λ	where * can be either 1 or 0
Geometric multiplicity	The geometric multiplicity determines the total number of Jordan sub blocks for λ	If $A_M = 4$; $G_M = 2$; $\lambda = 2$ There should be 2 Jordan sub blocks for $\lambda = 2$. So, J has 2 possibilities $ \mathbf{J} = \begin{pmatrix} 2 & 1 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 2 \end{pmatrix}; \text{ or } \mathbf{J} = \begin{pmatrix} 2 & 1 & 0 & 0 \\ 0 & 2 & 1 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{pmatrix} $
minimal polynomial	The multiplicity of λ in the minimal polynomial determines the size of the largest sub-block (Elementary Jordan block).	Let $p = (x - 2)^3$ be minimal polynomial \implies Size of largest sub-block is 3 Hence, one sub-block of size 3 and one sub-block of size 1 $\mathbf{J} = \begin{pmatrix} 2 & 1 & 0 & 0 \\ 0 & 2 & 1 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{pmatrix}$

TABLE 6.1.1.1: Properties of Jordan blocks and Jordan canonical form