

# Solutions: Linear Algebra by **Hoffman and Kunze**



(1.1.1.6)

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e) 
$$ab = (x + y\sqrt{2})(w + z\sqrt{2}) = ba, ab \in S$$
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f) 
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1.1.2. Let 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,

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

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

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

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

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,

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

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

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

Now, performing elementary operations on the right side of  $\mathbf{A}$  we obtain matrix  $\mathbf{B}$  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,

$$\begin{aligned} \mathbf{P} &= (\mathbf{E_1} \mathbf{E_2} \mathbf{E_3} \mathbf{E_4} \mathbf{E_5}) \\ &= \begin{pmatrix} 1 & 0 \\ -2 & 1 \end{pmatrix} \begin{pmatrix} \frac{1}{3} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \\ &= \begin{pmatrix} \frac{4}{3} & \frac{2}{3} \\ \frac{-5}{3} & \frac{-1}{3} \end{pmatrix} \quad (1.1.2.10) \end{aligned}$$

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)  

$$\Rightarrow \begin{pmatrix} 1 & 1 \end{pmatrix} \mathbf{x} = \frac{-1}{3} \begin{pmatrix} 1 & -1 \end{pmatrix} \mathbf{x} + \frac{2}{3} \begin{pmatrix} 2 & 1 \end{pmatrix} \mathbf{x}$$
 (1.1.2.20)

(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  $\mathbf{C}^{-1}$  with  $\mathbf{B}$  as it takes the linear combination of each rows of matrix  $\mathbf{B}$  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 equa-

tions 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:** 

$$x_1 - x_3 = 0$$
  

$$x_2 + 3x_3 = 0$$
(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 echelon forms of the augumented matrices of the following systems of homogeneous equations respectively.

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

$$\mathbf{BX} = \mathbf{0}$$
 (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),

$$\mathbf{R_2} = \mathbf{DA} \tag{1.1.4.12}$$

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.

1.1.5. Let  $\mathbb{F}$  be a set which contains exactly two elements,0 and 1.Define an addition and multiplication by tables. Verify that the set  $\mathbb{F}$ ,

$$\begin{array}{c|cccc} \cdot & 0 & 1 \\ \hline 0 & 0 & 0 \\ 1 & 0 & 1 \\ \end{array}$$

together with these two operations, is a field. **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  $\in$  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

$$\mathbf{A}\mathbf{x} = \mathbf{0}$$
 (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

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

$$\mathbf{R_2} \mathbf{y} = 0 \tag{1.1.6.6}$$

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  $\mathbf{R_1}$ ,  $\mathbf{R_2}$  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

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

$$\implies \mathbf{CA} = \mathbf{DB} \tag{1.1.6.16}$$

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  $\mathbb{F}$  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

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

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

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

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

$$(1.1.7.16)$$

$$\Rightarrow \frac{p}{q} \in \mathbb{F}$$

$$(1.1.7.17)$$

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 de-

fined 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 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)

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.

## 1.2 Matrices and Elementary Row Operations

# 1.2.1. Find all solutions to the system of equations

$$(1-i)x_1 - ix_2 = 0$$
  
2x<sub>1</sub> + (1-i)x<sub>2</sub> = 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,

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

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

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

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

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

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

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

#### 1.2.2. If

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

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 Methodology of matrix *A* :

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

$$(1.2.2.3)$$

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

$$(1.2.2.4)$$

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

$$(1.2.2.5)$$

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

$$(1.2.2.6)$$

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

$$(1.2.2.7)$$

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

$$(1.2.2.8)$$

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

$$(1.2.2.9)$$

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

$$(1.2.2.10)$$

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

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

(1.2.2.11)

the solution is:

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

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),

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

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  $\lambda = 2$ 

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

Substituting values:

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

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

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

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

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  $\lambda = 3$ 

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

substituting we have:

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

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

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

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

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

$$(1.2.4.5)$$

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

$$(1.2.4.6)$$

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

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

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

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

$$\mathbf{E_{13}E_{23}A_4} = \begin{pmatrix} 1 & 0 & -i \\ 0 & 1 & -\left(\frac{-1+i}{2}\right) \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & i \\ 0 & 1 & \frac{-1+i}{2} \\ 0 & 0 & 1 \end{pmatrix}$$

$$\implies \mathbf{A_5} = \mathbf{E_{13}}\mathbf{E_{23}}\mathbf{A_4} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
(1.2.4.19)

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

$$\mathbf{A} = \begin{pmatrix} i & -1 - i & 0 \\ 1 & -2 & 1 \\ 1 & 2i & -1 \end{pmatrix} \xrightarrow{RREF} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \mathbf{I}$$
(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.

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

$$(1.2.5.3)$$

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

$$(1.2.5.4)$$

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

$$(1.2.5.5)$$

 $\mathbf{B}^T$  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  $2\times2$  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)

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

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

$$a + 0 + c + 1 = 0$$
 (1.2.6.8)  
 $\Rightarrow a = -(c + 1)$  (1.2.6.9)  
 $\Rightarrow c \neq -1$  (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  $3 \times 3$  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  $E_1$ 

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

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.

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

$$\begin{pmatrix}
1 & 0 & 0 \\
0 & -1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
\mathbf{a}_1 \\
\mathbf{a}_2 \\
\mathbf{a}_3
\end{pmatrix} = \times
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}$$

$$\begin{pmatrix}
0 & 1 & 0 \\
1 & 0 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
\mathbf{a}_1 \\
\mathbf{a}_2 \\
\mathbf{a}_3
\end{pmatrix}$$
(1.2.7.16)

where

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  $E_1$ .

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

Now, to prove that same matrix can be obtained by other two elementary operations. We will first perform elementary operation  $E_2$  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,

ence, we can say that,
$$\begin{pmatrix}
1 & 0 & 0 \\
0 & -1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}
\begin{pmatrix}
1 & 2 & 3 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{pmatrix}$$
(1.2.7.16)

where

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

is a  $2 \times 2$  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 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 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 AX=0 as follows,

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & b & 0 \\ c & d & 0 \end{pmatrix} = \begin{pmatrix} \frac{1}{c} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} c & d & 0 \\ 0 & b & 0 \end{pmatrix}$$
 (1.2.8.3)
$$= \begin{pmatrix} 1 & 0 \\ 0 & \frac{1}{b} \end{pmatrix} \begin{pmatrix} 1 & \frac{d}{c} & 0 \\ 0 & b & 0 \end{pmatrix}$$
 (1.2.8.4)
$$= \begin{pmatrix} 1 & -\frac{d}{c} \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{d}{c} & 0 \\ 0 & 1 & 0 \end{pmatrix}$$
 (1.2.8.5)
$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$
 (1.2.8.6)

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

$$\mathbf{AX} = \mathbf{u} \tag{1.2.8.7}$$

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

From (1.2.8.12) we get,

$$x_{1} = \frac{du_{1} - bu_{2}}{ad - bc}$$

$$x_{2} = \frac{au_{2} - cu_{1}}{ad - bc}$$
(1.2.8.13)
$$(1.2.8.14)$$

$$x_2 = \frac{au_2 - cu_1}{ad - bc} \tag{1.2.8.14}$$

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 AX = 0 has only one trivial solution i.e  $x_1 = x_2 = 0$  in all cases. Hence proved, the equation **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 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,

$$x_1 = yx_1^0 \tag{1.2.8.21}$$

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

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

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

$$x_2 = yx_2^0 (1.2.8.32) 1$$

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)

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

Hence using the result from (1.2.8.29) we can conclude for (1.2.8.34), 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,

$$x_1 = yx_1^0 (1.2.8.36)$$

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

which is a solution of the equation AX = 0.

1.3 Row Reduced Echelon Matrices

(1.2.8.32) 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:

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

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

$$(1.3.1.6)$$

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

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

(1.3.1.8)

(1.3.1.7)

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

(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)

$$\xrightarrow[R_1 \leftarrow R_1 + \frac{5R_3}{4}]{(1 \quad 0 \quad 0) \atop 0 \quad 1 \quad 0} \atop 0 \quad 0 \quad 1 \atop 0 \quad 0 \quad 0$$
(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 Ax = 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

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

$$= \begin{pmatrix} 1 & i & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -i & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{4}(1-i) & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

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

$$= \begin{pmatrix} \frac{1}{2}(1-i) & \frac{1}{4}(1+i) & 0 \\ \frac{1}{2}(-1+i) & \frac{1}{4}(1-i) & 0 \\ \frac{1}{2}(1-i) & \frac{1}{4}(-1-i) & 1 \end{pmatrix} (1.3.2.4)$$

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

:. Row-reduced matrix of A is,

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

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,

$$\mathbf{I_2x} = 0 \tag{1.3.2.8}$$

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

As  $I_2$  is invertible.

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,

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

$$\implies$$
 **Ax** = **b** (1.3.4.5)

Where,

(1.3.4.6)

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

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

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

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

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

$$\xrightarrow{(1.3.4.13)}$$

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)

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

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

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 1.3.6. Find all solutions of space V

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  $(5 \ 2)x = 7$  and  $(10 \ 4)x = -3$

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

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

$$\stackrel{R_1 = \frac{R_1}{5}}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{2}{5} & \frac{7}{5} \\ 0 & 0 & -17 \end{pmatrix} \tag{1.3.5.4}$$

$$\stackrel{R_2 = \frac{R_2}{-17}}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{2}{5} & \frac{7}{5} \\ 0 & 0 & 1 \end{pmatrix} \tag{1.3.5.5}$$

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

(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 and only if rank A < rank A | B.

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

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

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

**Solution:** The given equations can be written as,

$$\mathbf{A}\mathbf{x} = B \tag{1.3.6.1}$$

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

Now, we form the augmented matrix and per-

form Row reduction,

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

$$(1.3.6.3)$$

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

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

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

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

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

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

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

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

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

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

## 1.3.7. Find all solutions of

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

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

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

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

**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,

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

$$(1.3.7.6)$$

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

$$(1.3.7.7)$$

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

$$(1.3.7.8)$$

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

$$(1.3.7.9)$$

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

$$(1.3.7.10)$$

$$\stackrel{R_3=R_3+6R_2.R_4=R_4-7R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -2 & 1 & 1 & | & 2 \\ 0 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} & 0 & | & -1 \\ 0 & 0 & 0 & 0 & -1 & | & -1 \\ 0 & 0 & 0 & 0 & 1 & | & 1 \end{pmatrix}$$

$$(1.3.7.11)$$

$$\stackrel{R_2=-2R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -2 & 1 & 1 & | & 2 \\ 0 & 1 & 1 & -1 & 0 & | & 2 \\ 0 & 0 & 0 & 0 & -1 & | & -1 \\ 0 & 0 & 0 & 0 & 1 & | & 1 \end{pmatrix}$$

$$(1.3.7.12)$$

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

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)

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.

#### 1.3.8. Let

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

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

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

$$\mathbf{B_1A} = \begin{pmatrix} 3 & -6 & 2 & -1 \\ 0 & 0 & \frac{7}{3} & \frac{7}{3} \end{pmatrix}$$
 (1.3.9.11)1.3.10

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

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,

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

(1.3.9.11)1.3.10. Suppose  $\mathbf{R}$  and  $\mathbf{R}'$  are 2 × 3 row-reduced echelon matrices and that the system  $\mathbf{R}\mathbf{X}$ =0 and  $\mathbf{R}'\mathbf{X}$ =0 have exactly the same solutions. Prove that  $\mathbf{R} = \mathbf{R}'$ .

#### **Solution:**

Since **R** and **R**' are  $2 \times 3$  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)

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

Since y is a  $2 \times 1$  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  $y_1 \neq 0$ . Now, Substituting (1.3.10.15) in (1.3.10.6)

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

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

$$\mathbf{RX} = 0 \tag{1.3.10.23}$$

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

Here,

$$\mathbf{R} = \begin{pmatrix} \mathbf{I} & \mathbf{a} \end{pmatrix} \tag{1.3.10.25}$$

$$\mathbf{a} = \begin{pmatrix} b \\ c \end{pmatrix} \tag{1.3.10.26}$$

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

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)

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

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)

$$(\mathbf{y}^T \quad z) \begin{pmatrix} \mathbf{I} \\ \mathbf{b}^T \end{pmatrix} = 0$$
 (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 **RX**=0 will be satisfied for all values of 1.4.2. Let **X**. We know that  $\mathbf{R}'\mathbf{X}=0$  will have the exact

same solution as **RX**=0 then we can say that for all values of X=0 equation R'X=0 will

be satisfied. Hence,  $\mathbf{R}' = \mathbf{R} = 0$ .

1.4 Matrix Multiplication

1.4.1. Let

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

CAB = 
$$\begin{pmatrix} 1 & -1 \end{pmatrix} \begin{pmatrix} 4 \\ 4 \end{pmatrix}$$
 Hence verified.  
Find two different 2×2 matrices **A** such that  $\mathbf{A}^2 = 0$  but  $\mathbf{A} \neq 0$ 

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

therefore,

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

$$\mathbf{CAB} = 0 \tag{1.4.1.13}$$

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

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

**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 null space of A contains columns of matrix A. Also atleast A contains columns of matrix A. Also already one of the columns must be non-zero since given  $\mathbf{A} \neq 0$ . Thus, the null space of A contains 1.4.4. For the matrix  $\mathbf{A} = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 0 & 1 \\ 3 & 0 & 1 \end{pmatrix}$ , find elementary  $\mathbf{A} \neq 0$ . non zero vectors,  $rank(\mathbf{A}) < 2$ . Hence,  $\mathbf{A}$  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,  $n_2 = -cn_1$ 

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

From (1.4.3.13), (1.4.3.19) two different  $2\times 2$ 

matrices A can be given as,

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

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

tary matrices  $E_1, E_2, ..., E_k$  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,

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

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

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

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

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

$$\mathbf{E_6} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \frac{2}{7} \end{pmatrix} \tag{1.4.4.8}$$

$$\mathbf{E_7} = \begin{pmatrix} 1 & 0 & \frac{-1}{2} \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{1.4.4.9}$$

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

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,

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

$$(1.4.4.12)$$

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  $2 \times 2$  matrix and A is a  $3 \times 2$  matrix. so matrix C must be a  $2 \times 3$  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,

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

$$(1.4.5.4)$$

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

$$(1.4.5.5)$$

$$\Rightarrow \begin{pmatrix} 1 & 2 & 1 \\ -1 & 2 & 0 \end{pmatrix} \begin{pmatrix} \mathbf{c}_{1} & \mathbf{c}_{2} \end{pmatrix} = \begin{pmatrix} 3 & -4 \\ 1 & 4 \end{pmatrix}$$

$$(1.4.5.6)$$

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:

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

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

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

$$(1.4.5.12)$$

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,

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

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

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

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

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 1.4.7. Let **A** and **B** be  $n \times n$  matrices such that AB = I.

Brown that AB = I. Solution: Let AB = I.

1.4.6. Let **A** be an  $m \times n$  matrix and **B** be an  $n \times k$  matrix. Show that the columns of **C** = **AB** are linear combinations of columns of **A**. If  $\alpha_1, \alpha_2, \ldots, \alpha_n$  are the columns of **A** and  $\gamma_1, \gamma_2, \ldots, \gamma_k$  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  $\gamma_1$ 

$$\gamma_1 = \mathbf{A}\beta_1 \qquad (1.4.6.8)$$

$$= \begin{pmatrix} \alpha_1 & \alpha_2 & \dots & \alpha_n \end{pmatrix} \begin{pmatrix} B_{11} \\ B_{21} \\ \vdots \\ B_{n1} \end{pmatrix}$$
 (1.4.6.9)

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

Similarly, considering  $j^{th}$  column of  $\mathbb{C}$ 

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

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,

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

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

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

$$\implies$$
 **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 (1.4.7.8)

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

$$\implies$$
 **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.

#### 1.4.8. 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)

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

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

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

$$\implies tr(\mathbf{AB}) = tr(\mathbf{BA})$$
 (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  $\mathbf{R}$  which is row-equivalent to  $\mathbf{A}$  and an invertible 3x3 matrix  $\mathbf{P}$  such that  $\mathbf{R} = \mathbf{P} \mathbf{A}$ . 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

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

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

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

$$(1.5.1.5)$$

$$\stackrel{R_1=R_1-R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -3 & -5 & 0 & -1 & 0 \\
0 & 2 & 4 & 5 & 1 & 1 & 0 \\
0 & -4 & 0 & 1 & -1 & 0 & 1
\end{pmatrix}$$
(1.5.1.6)

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

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

$$\stackrel{R_1=R_1+3R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & -\frac{7}{8} \\
0 & 1 & 2 & \frac{5}{2} \\
0 & 0 & 1 & \frac{11}{8}
\end{pmatrix} \begin{vmatrix}
\frac{3}{8} & -\frac{1}{4} & \frac{3}{8} \\
\frac{1}{2} & \frac{1}{2} & 0 \\
\frac{1}{8} & \frac{1}{4} & \frac{1}{8}
\end{pmatrix} (1.5.1.10)$$

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

$$(1.5.1.11)$$

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

$$P = E$$
 (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)

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

echelon matrix  $\mathbf{R}$  which is row-equivalent to  $\mathbf{A}$  and an invertible 3x3 matrix  $\mathbf{P}$  such that  $\mathbf{R} = \mathbf{P} \mathbf{A}$ . 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

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

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

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

$$(1.5.2.4)$$

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

$$(1.5.2.5)$$

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

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

$$(1.5.2.7)$$

$$\stackrel{R_3 \leftarrow R_3/(3-i)/2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & \frac{i}{2} & | & \frac{1}{2} & 0 & 0 \\
0 & 1 & \frac{i}{2} & | & \frac{1}{6} & -\frac{1}{3} & 0 \\
0 & 0 & 1 & | & -\frac{(i)}{3} & \frac{3+i}{15} & \frac{3+i}{5}
\end{pmatrix} (1.5.2.8)$$

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

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

$$= [I E]$$

Hence, the row reduced matrix that is row equivalent to  $\mathbf{A}$  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 A i.e

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

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

$$\begin{pmatrix} 2 & 5 & -1 \\ 4 & -1 & 2 \\ 6 & 4 & 1 \end{pmatrix} \longleftrightarrow \mathbf{A} = \begin{pmatrix} 2 & 5 & -1 \\ 0 & -11 & 4 \\ 6 & 4 & 1 \end{pmatrix}$$

$$(1.5.3.2)$$

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

$$\stackrel{R_1 = R_1 - \frac{5}{2}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{9}{22} \\ 0 & 1 & \frac{-4}{11} \\ 0 & 0 & 0 \end{pmatrix}$$
(1.5.3.7)

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  $\mathbf{B}|\mathbf{I}$ , By applying row reductions on  $\mathbf{B}|\mathbf{I}$ 

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

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

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

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

$$(1.5.3.14)$$

For a matrix to be invertible, it has to be a matrix of full rank. Here, the matrix **B** is of full

rank ( $Rank(\mathbf{B}) = 3$ ). Therefore **B** is invertible 1.5.5. Discover whether 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)

1.5.4. Let

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

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.

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

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  $\mathbf{E}[\mathbf{A}\ \mathbf{I}] = [\mathbf{I}\ \mathbf{E}]$  then **E** is the inverse of **A** i.e  $\mathbf{E} = \mathbf{A}^{-1}$ .

$$[\mathbf{A} \ \mathbf{I}] = \begin{pmatrix} 1 & 2 & 3 & 4 & | & 1 & 0 & 0 & 0 \\ 0 & 2 & 3 & 4 & | & 0 & 1 & 0 & 0 \\ 0 & 0 & 3 & 4 & | & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 4 & | & 0 & 0 & 0 & 1 \end{pmatrix}$$
(1.5.5.2)

$$\stackrel{R_1 \leftarrow R_1 - R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & 0 & | & 1 & -1 & 0 & 0 \\
0 & 2 & 3 & 4 & | & 0 & 1 & 0 & 0 \\
0 & 0 & 3 & 4 & | & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 4 & | & 0 & 0 & 0 & 1
\end{pmatrix} (1.5.5.3)$$

$$\stackrel{R_2 \leftarrow R_2 - R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & 0 & | & 1 & -1 & 0 & 0 \\
0 & 2 & 0 & 0 & | & 0 & 1 & -1 & 0 \\
0 & 0 & 3 & 4 & | & 0 & 0 & 1 & 0 \\
0 & 0 & 0 & 4 & | & 0 & 0 & 0 & 1
\end{pmatrix} (1.5.5.4)$$

$$\stackrel{R_3 \leftarrow R_3 - R_4}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & 0 & | & 1 & -1 & 0 & 0 \\
0 & 2 & 0 & 0 & | & 0 & 1 & -1 & 0 \\
0 & 0 & 3 & 0 & | & 0 & 0 & 1 & -1 \\
0 & 0 & 0 & 4 & | & 0 & 0 & 0 & 1
\end{pmatrix}$$
(1.5.5.5)

$$\xrightarrow{R_{4} \leftarrow \frac{R_{4}}{4}}
\xrightarrow{R_{2} \leftarrow \frac{R_{2}}{2} R_{3} \leftarrow \frac{R_{3}}{3}}
\begin{pmatrix}
1 & 0 & 0 & 0 & | & 1 & -1 & 0 & 0 \\
0 & 1 & 0 & 0 & | & 0 & \frac{1}{2} & -\frac{1}{2} & 0 \\
0 & 0 & 1 & 0 & | & 0 & 0 & \frac{1}{3} & -\frac{1}{3} \\
0 & 0 & 0 & 1 & | & 0 & 0 & 0 & \frac{1}{4}
\end{pmatrix}$$

$$= [\mathbf{I} \ \mathbf{E}]$$
(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)

$$\mathbf{E}_{2}\mathbf{E}_{1}\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & -1 & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & 1 \end{pmatrix} \begin{pmatrix} a_{1} & 0 & 0 & \dots & 0 \\ 0 & a_{2} & a_{3} & \dots & a_{N} \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & 1 \end{pmatrix} \begin{pmatrix} a_{1} & 0 & 0 & \dots & 0 \\ 0 & a_{2} & a_{3} & \dots & a_{N} \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & a_{N} \end{pmatrix} \text{matrix.Prove that } \mathbf{C} = \mathbf{A}\mathbf{B} \text{ is non invertible.}$$
Solution: Let's take  $\mathbf{A}$  and  $\mathbf{B}$  to be non zero vectors. Now,we know that for  $\mathbf{C}$  to be non invertible  $\mathbf{C}\mathbf{x} = 0$  should have a non trivial solution.So,

$$= \begin{pmatrix} a_1 & 0 & 0 & \dots & 0 \\ 0 & a_2 & 0 & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & a_N \end{pmatrix}$$
 (1.5.5.12)

Proceeding in similar manner, we get

$$\mathbf{E}_{N}\mathbf{E}_{N-1}\dots\mathbf{E}_{2}\mathbf{E}_{1}\mathbf{A} = \mathbf{U} = \begin{pmatrix} a_{1} & 0 & 0 & \dots & 0 \\ 0 & a_{2} & 0 & \dots & 0 \\ 0 & 0 & a_{3} & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & a_{N} \end{pmatrix}$$

$$(1.5.5.13)$$

$$= \operatorname{diag} \begin{pmatrix} a_{1} & a_{2} & \dots & a_{N} \end{pmatrix}$$

$$(1.5.5.14)$$

$$\implies \mathbf{A} = \mathbf{L}\mathbf{U} \tag{1.5.5.15}$$

where 
$$\mathbf{L} = \mathbf{E}_1^{-1} \mathbf{E}_2^{-1} \dots \mathbf{E}_N^{-1}$$
  
 $\implies \mathbf{A}^{-1} = \mathbf{U}^{-1} \mathbf{L}^{-1}$  (1.5.5.16)

$$\Rightarrow \mathbf{A}^{-1} = \begin{pmatrix} \frac{1}{a_1} & 0 & 0 & \dots & 0 \\ 0 & \frac{1}{a_2} & 0 & \dots & 0 \\ 0 & 0 & \frac{1}{a_3} & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & \frac{1}{a_N} \end{pmatrix} \begin{pmatrix} 1 & -1 & 0 & \dots & 0 \\ 0 & 1 & -1 & \dots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \dots & \dots & \dots & \dots \end{pmatrix}$$

$$(1.5.5.17)$$

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  $2\times1$  matrix and **B** is  $1\times2$ matrix. Prove that **C=AB** is non invertible.

 $\ldots a_N$ ) 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  $1 \times 2$  matrix and **x** is  $2 \times 1$  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  $1 \times 2$  matrix i.e. rows are less than column hence,

$$\mathbf{Bx} = 0$$
 (1.5.6.4)

will have a non trivial solution. Hence, using (1.5.6.3) and (1.5.6.4) we can say,

$$ABx = 0$$
 (1.5.6.5)

will have a non trivial solution so, C is non

invertible.

- 1.5.7. Let **A** be an  $n \times n$  (square) matrix, Prove the following two statements:
  - a) If **A** is invertible and  $\mathbf{AB} = 0$  for some  $n \times n$  matrix **B**, then  $\mathbf{B} = 0$ .
  - b) If **A** is not invertible, then there exists an  $n \times n$  matrix **B** such that AB = 0 but  $B \neq 0$ .

#### **Solution:**

a) Given **A** is an invertible matrix and  $\mathbf{AB} = 0$  then.

$$\mathbf{AB} = 0 \tag{1.5.7.1}$$

$$\implies \mathbf{A}^{-1}(\mathbf{AB}) = 0 \tag{1.5.7.2}$$

$$\implies (\mathbf{A}^{-1}\mathbf{A})\mathbf{B} = 0 \qquad (1.5.7.3)$$
$$\implies \mathbf{IB} = 0 \quad [\because \mathbf{A}^{-1}\mathbf{A} = \mathbf{I}]$$

$$\implies \mathbf{B} = 0 \tag{1.5.7.5}$$

b) If **A** is not invertible, then there exists an  $n \times n$  matrix **B** such that  $\mathbf{AB} = 0$  but  $\mathbf{B} \neq 0$ . Since **A** is not invertible,  $\mathbf{AX} = 0$  must have a non-trivial solution. Let the non-trivial 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 **y**.

$$\mathbf{B} = \begin{pmatrix} \mathbf{y} & \mathbf{y} & \cdots & \mathbf{y} \end{pmatrix} \tag{1.5.7.7}$$

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  $(A|I) \rightarrow (I|A^{-1})$ . 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  $\bf A$  to  $\bf I$  cannot even begin. So, one necessary condition for  $\bf A$  to be invertible is that the entries  $\bf a$  and  $\bf c$  are not both 0.

a) Assume that  $a \neq 0$ , Then:

$$\begin{pmatrix} a & b & 1 & 0 \\ c & d & 0 & 1 \end{pmatrix} \xrightarrow{R_1 = R_1/a} \begin{pmatrix} 1 & \frac{b}{a} & \frac{1}{a} & 0 \\ c & d & 0 & 1 \end{pmatrix}$$

$$(1.5.8.3)$$

$$\xrightarrow{R_2 = R_2 - cR_1} \begin{pmatrix} 1 & \frac{b}{a} & \frac{1}{a} & 0 \\ 0 & \frac{ad-bc}{a} & \frac{-c}{a} & 1 \end{pmatrix}$$

Next, assuming that  $ad - bc \neq 0$ , we get:

$$\stackrel{R_1=R_1-\frac{b}{ad-bc}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{d}{ad-bc} & \frac{-b}{ad-bc} \\ 0 & \frac{ad-bc}{a} & \frac{-c}{a} & 1 \end{pmatrix}$$

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

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  $c \neq 0$ :

$$\begin{pmatrix}
0 & b & 1 & 0 \\
c & d & 0 & 1
\end{pmatrix}
\xrightarrow{R_1 \leftrightarrow R_2} \begin{pmatrix}
c & d & 0 & 1 \\
0 & b & 1 & 0
\end{pmatrix}$$

$$(1.5.8.5)$$

$$\xrightarrow{R_1 = R_1/c} \begin{pmatrix}
1 & \frac{d}{c} & 0 & \frac{1}{c} \\
0 & b & 1 & 0
\end{pmatrix}$$

$$\xrightarrow{R_1 = R_1 - R_2 \times \frac{d}{bc}} \begin{pmatrix}
1 & 0 & -\frac{d}{bc} & \frac{1}{c} \\
0 & b & 1 & 0
\end{pmatrix}$$

$$\xrightarrow{R_2 = R_2/b} \begin{pmatrix}
1 & 0 & -\frac{d}{bc} & \frac{1}{c} \\
0 & 1 & \frac{1}{b} & 0
\end{pmatrix}$$

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  $a \neq 0$ :

$$\begin{pmatrix} a & b & 1 & 0 \\ 0 & d & 0 & 1 \end{pmatrix} \xrightarrow{R_1 = R_1/a} \begin{pmatrix} 1 & \frac{b}{a} & \frac{1}{a} & 0 \\ 0 & d & 0 & 1 \end{pmatrix}$$

$$(1.5.8.6)$$

$$\xrightarrow{R_1 = R_1 - R_2 \times \frac{b}{ad}} \begin{pmatrix} 1 & 0 & \frac{1}{a} & -\frac{b}{ad} \\ 0 & d & 0 & 1 \end{pmatrix}$$

$$\xrightarrow{R_2 = R_2/d} \begin{pmatrix} 1 & 0 & \frac{1}{a} & -\frac{b}{ad} \\ 0 & 1 & 0 & \frac{1}{d} \end{pmatrix}$$

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  $\mathbf{A}$  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,

$$\mathbf{R}' = \begin{pmatrix} \mathbf{I} & \mathbf{F} \\ \mathbf{0} & \mathbf{0} \end{pmatrix}$$

I is an identity matrix, **F** is Free variables matrix and **0** represents a block of zeroes

 $\mathbf{R}'$  is in row-reduced echelon form. To perform column operations, elementary matrices should be multiplied on the right side in order to convert the  $\mathbf{R}'$  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** 

$$\mathbf{P} = \begin{pmatrix} -5 & 3 & 0 \\ 2 & -1 & 0 \\ -1 & 0 & 1 \end{pmatrix} \tag{1.5.10.7}$$

$$\mathbf{R'} = \mathbf{PA} = \begin{pmatrix} 1 & 2 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix} \tag{1.5.10.8}$$

 $\mathbf{R}'$  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)

#### 2 Vector Spaces

#### 2.1 Vector Spaces

2.1.1. If  $\mathbf{F}$  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  $\mathbf{F}$  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  $F^n$ :

Let 
$$\alpha = (a_i)$$
,  $\beta = (b_i)$ ,  $\gamma = (g_i) \ \forall \ i = 1, 2, \dots, n \in \mathbf{F}^n$  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  $F^n$ :

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

$$\left(-a_i\right) + \left(a_i\right) = \mathbf{0} \tag{2.1.1.15}$$

Therefore  $-\alpha = (-a_i)$  is the additive inverse of  $\alpha$ . 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)

$$= (a[a_i + b_i]) (2.1.1.17)$$

$$= \left(aa_i + ab_i\right) \tag{2.1.1.18}$$

$$(aa_i) + (ab_i) \tag{2.1.1.19}$$

$$= a(a_i) + a(b_i)$$
 (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)

$$= \left(aa_i + ba_i\right) \tag{2.1.1.23}$$

$$= (aa_i) + (ba_i) (2.1.1.24)$$

$$= a\left(a_i\right) + b\left(a_i\right) \tag{2.1.1.25}$$

$$= a\alpha + b\alpha \tag{2.1.1.26}$$

c) If  $a,b \in \mathbf{F}$  and  $\alpha = (a_i) \ \forall \ i = 1, 2, \dots, n \in \mathbf{F}^n$ 

then

$$(ab)\alpha = ([ab]a_i) \tag{2.1.1.27}$$

$$= \left(a[ba_i]\right) \tag{2.1.1.28}$$

$$= a\left(ba_i\right) \tag{2.1.1.29}$$

$$= a(b\alpha) \tag{2.1.1.30}$$

d) If 1 is the unity element of  $\mathbf{F}$  and  $\alpha$  =  $(a_i) \ \forall \ i=1,2,\cdots,n \in \mathbf{F}^n \text{ then}$ 

$$1\alpha = (1a_i) \tag{2.1.1.31}$$

$$= (a_i) \tag{2.1.1.32}$$

$$= \alpha \tag{2.1.1.33}$$

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)

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)

=  $(aa_i) + (ba_i)$  (2.1.1.24) 2.1.3. If  $\mathbb{C}$  is the field of complex numbers, which vectors in  $\mathbb{C}^3$  are linear combinations of  $\begin{bmatrix} 0 \\ \end{bmatrix}$ ,

$$\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$
 and  $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ?

**Solution:** Expressing the given vectors as the 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 
$$\begin{pmatrix} 1\\0\\-1 \end{pmatrix}$$
,  $\begin{pmatrix} 0\\1\\1\\1 \end{pmatrix}$  and  $\begin{pmatrix} 1\\1\\1 \end{pmatrix}$ .

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) Hence **V** is not a vector space.  
 $c(x,y) = (cx,y)$  (2.1.4.2) Let  $\mathbb{V}$  be the set of all complex-valued functions from the real line such that

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)

$$= ((a+b)x_1, y_1) \neq au + bu \qquad (2.1.4.4)$$

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.

2.1.5. On  $\mathbb{R}^n$  define two 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 **0** in **V** called the zero vector, such that  $\alpha + \mathbf{0} = \alpha$  for all  $\alpha$  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

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 $\beta$
$\alpha \oplus (\beta \oplus \gamma) \neq (\alpha \oplus \beta) \oplus \gamma$	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 $\alpha$
$\alpha \oplus \beta \neq \beta \oplus \alpha$	
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  $(\mathbb{R}^n, \oplus, \cdot)$ 

cf(t)+g(t).

Hence, f(x) is not real valued. Now,

$$(cf+g)(t)$$
 (2.1.7.4)  $f(x) = a + ix$  (2.1.7.13)  
=  $(cf)(t) + g(t)$  (2.1.7.5)  $f(-x) = a - ix$  (2.1.7.14)  
=  $cf(t) + g(t)$  (2.1.7.6)  $f(-x) = \overline{f(x)}$  (2.1.7.15)

Now, we know that  $f(-t) = \overline{f(-t)}$  and so (cf+g)(t) should also satisfy the property,

Since a and  $x \in \mathbb{R}$ , so  $f \in \mathbb{V}$ 

$$(cf + g)(-t)$$
 (2.1.7.7) 2.2 Subspaces  
=  $cf(-t) + g(-t)$  (2.1.7.8) 2.2.1. Which of the following set of vectors  
=  $cf(t) + g(t)$  (2.1.7.9)  $\alpha = (a_1, a_2, \dots, a_n)$   
=  $cf(t) + g(t)$  (2.1.7.10) in  $\mathbf{R}^n$  are subspace of  $\mathbf{R}^n$  ( $n \ge 3$ )?

**Example** Let's take f(x)=a+ix

a) All  $\alpha$  such that  $a_1 \ge 0$ 

$$f(1) = a + i$$
 (2.1.7.12) b) All  $\alpha$  such that  $a_1 + 3a_2 = a_3$ 

c) All  $\alpha$  such that  $a_2 = a_1^2$ 

$\alpha = (a_1, a_2, \dots, a_n)$			
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 rational	Not a subspace. Scalar multiplication is not satisfied. $a_2 \neq \sqrt{2}a_1$		

TABLE 2.2.1: Summary

- d) All  $\alpha$  such that  $a_1a_2 = 0$
- e) All  $\alpha$  such that  $a_2$  is rational **Solution:** Table 2.2.1 lists the summary of which set of vectors in  $\mathbf{R}^n$  are subspace of  $\mathbf{R}^n$  $(n \ge 3)$ .
- 2.2.2. Is the vector  $\begin{pmatrix} 3 \\ -1 \\ 0 \\ 1 \end{pmatrix}$  in the subspace of  $\mathbb{R}^4$

spanned by the vectors  $\begin{pmatrix} 2 \\ -1 \\ 3 \\ 2 \end{pmatrix}$ ,  $\begin{pmatrix} -1 \\ 1 \\ 1 \\ -3 \end{pmatrix}$  and  $\begin{pmatrix} 1 \\ 1 \\ 9 \\ 2 \end{pmatrix}$ 

? **Solution:** Expressing the given three vectors as columns of a matrix,

$$\mathbf{A} = \begin{pmatrix} 2 & -1 & 1 \\ -1 & 1 & 1 \\ 3 & 1 & 9 \\ 2 & -3 & -5 \end{pmatrix} \tag{2.2.2.1}$$

and

$$\mathbf{b} = \begin{pmatrix} 3 \\ -1 \\ 0 \\ -1 \end{pmatrix} \tag{2.2.2.2}$$

spanned by the three vectors.

$$\mathbf{A}\mathbf{x} = \mathbf{b} \tag{2.2.2.3}$$

must have a solution.

$$\begin{pmatrix} 2 & -1 & 1 \\ -1 & 1 & 1 \\ 3 & 1 & 9 \\ 2 & -3 & -5 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 3 \\ -1 \\ 0 \\ -1 \end{pmatrix}$$
 (2.2.2.4)

Forming the augmented matrix and row reducing it by elementary row operations,

$$\begin{pmatrix}
2 & -1 & 1 & 3 \\
-1 & 1 & 1 & -1 \\
3 & 1 & 9 & 0 \\
2 & -3 & -5 & -1
\end{pmatrix}
\xrightarrow{R_2 \leftarrow 2R_2 + R_1, R_3 \leftarrow R_3 - \frac{3}{2}R_1}
\xrightarrow{R_4 \leftarrow R_4 - R_1}$$

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

$$\begin{pmatrix} 2 & -1 & 1 & 3 \\ 0 & 1 & 3 & 1 \\ 0 & \frac{5}{2} & \frac{15}{2} & \frac{-9}{2} \\ 0 & -2 & -6 & -4 \end{pmatrix} \xrightarrow{R_3 \leftarrow 2R_3 - 5R_2} \begin{pmatrix} 2 & -1 & 1 & 3 \\ 0 & 1 & 3 & 1 \\ 0 & 0 & 0 & -14 \\ 0 & 0 & 0 & -2 \end{pmatrix}$$

$$(2.2.2.6)$$

From (2.2.2.6), it is clear that the system does

not have a solution. Hence the vector  $\begin{pmatrix} 3 \\ -1 \\ 0 \end{pmatrix}$  does

not lie in the subspace of  $\mathbf{R}^4$  spanned by the given three vectors.

For the vector **b** to be in the subspace of  $\mathbf{R}^4$  2.2.3. Let **W** be the set of all  $(x_1, x_2, x_3, x_4, x_5)$  in  $\mathbb{R}^5$ 

$$2x_1 - x_2 + \frac{4}{3}x_3 - x_4 = 0 (2.2.3.1)$$
$$x_1 + \frac{2}{3}x_3 - x_5 = 0 (2.2.3.2)$$
$$9x_1 - 3x_2 + 6x_3 - 3x_4 - 3x_5 = 0 (2.2.3.3)$$

Find a finite set of vectors which spans W. **Solution:** The given equations are,

$$2x_1 - x_2 + \frac{4}{3}x_3 - x_4 = 0 (2.2.3.4)$$
$$x_1 + \frac{2}{3}x_3 - x_5 = 0 (2.2.3.5)$$

$$9x_1 - 3x_2 + 6x_3 - 3x_4 - 3x_5 = 0 (2.2.3.6)$$

which can be written as,

$$\begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 \\ 1 & 0 & \frac{2}{3} & 0 & -1 \\ 9 & -3 & 6 & -3 & -3 \end{pmatrix} \mathbf{x} = 0$$
 (2.2.3.7)

Now, the augmented matrix,

$$\begin{pmatrix}
2 & -1 & \frac{4}{3} & -1 & 0 & 0 \\
1 & 0 & \frac{2}{3} & 0 & -1 & 0 \\
9 & -3 & 6 & -3 & -3 & 0
\end{pmatrix}$$

$$\stackrel{R_3=R_3-3R_1-3R_2}{\longleftrightarrow} \begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 & 0 \\ 1 & 0 & \frac{2}{3} & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$
(2.2.3.9)

$$\stackrel{R_2=R_2-\frac{1}{2}R_1}{\longleftrightarrow} \begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 & 0\\ 0 & \frac{1}{2} & 0 & \frac{1}{2} & -1 & 0\\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$
(2.2.3.10)

$$\stackrel{R_2=2R_2}{\longleftrightarrow} \begin{pmatrix} 2 & -1 & \frac{4}{3} & -1 & 0 & 0\\ 0 & 1 & 0 & 1 & -2 & 0\\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(2.2.3.11)$$

$$\stackrel{R_1=R_1+R_2}{\longleftrightarrow} \begin{pmatrix} 2 & 0 & \frac{4}{3} & 0 & -2 & 0\\ 0 & 1 & 0 & 1 & -2 & 0\\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$(2.2.3.12)$$

So,

$$2x_1 + \frac{4}{3}x_3 - 2x_5 = 0 (2.2.3.13)$$

$$x_2 + x_4 - 2x_5 = 0 (2.2.3.14)$$

Solving the equations we get,

$$x_1 = -\frac{2}{3}x_3 + x_5 \tag{2.2.3.15}$$

$$x_2 = -x_4 + 2x_5 \tag{2.2.3.16}$$

which can be written as,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} \tag{2.2.3.17}$$

$$= \begin{pmatrix} -\frac{2}{3}x_3 + x_5 \\ -x_4 + 2x_5 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 (2.2.3.18)

$$= x_3 \begin{pmatrix} -\frac{2}{3} \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} 0 \\ -1 \\ 0 \\ 1 \\ 0 \end{pmatrix} + x_5 \begin{pmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$
 (2.2.3.19)

where  $x_3, x_4$  and  $x_5 \in \mathbb{R}$ . Hence, the vectors

$$\begin{pmatrix} -\frac{2}{3} \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 \\ 0 \\ 1 \\ 0 \end{pmatrix} \text{ and } \begin{pmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 1 \end{pmatrix} \text{ will span } \mathbf{W}$$

- 2.2.4. 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,

$$\mathbf{A} = \mathbf{I} \tag{2.2.4.1}$$

$$\mathbf{B} = -\mathbf{I} \tag{2.2.4.2}$$

It could be easily proven that both matrices

A and B are invertible as,

$$rank(\mathbf{I}_{nxn}) = rank \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{pmatrix}_{nxn}$$

$$(2.2.4.3)$$

$$\implies rank(-\mathbf{I}_{nxn}) = rank(\mathbf{I}_{nxn}) = n$$

$$(2.2.4.4)$$

or it is a full rank matrix as there are n pivots.

$$\therefore \mathbf{A} + \mathbf{B} = \mathbf{0}. \tag{2.2.4.5}$$

But the zero matrix **0** is non-invertible as,

$$rank(\mathbf{0}_{nxn}) = rank \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}_{nxn}$$

$$(2.2.4.6)$$

$$\implies rank(\mathbf{0}_{nxn}) = 0$$

$$(2.2.4.7)$$

# .. the set of invertible matrices are not closed under addition. Hence not a subspace of V.

b) Let the matrices  $A_1, A_2, \cdots, A_n \in V$ , be set of non-invertible matrix. For them to be a subspace they need to be closed under addition. Let,

$$\mathbf{A_{1}} = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}_{nxn}$$

$$\mathbf{A_{2}} = \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}_{nxn}$$

$$\mathbf{A_{n}} = \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}_{nxn}$$

$$(2.2.4.9)$$

 $\begin{pmatrix}
\vdots & \vdots & \vdots \\
0 & 0 & \cdots & 1
\end{pmatrix}_{nxn}$ (2.2.4.11)

It could be proven that matrices  $A_1$ ,

 $A_2, \cdots, A_n$  are non-invertible as,

$$rank(\mathbf{A}_{1}) = rank \begin{pmatrix} \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix} \end{pmatrix}$$

$$(2.2.4.12)$$

$$\implies rank(\mathbf{A}_{1}) = 1$$

$$(2.2.4.13)$$

or there is only one pivot hence rank is 1.

$$\implies \mathbf{A_1} + \mathbf{A_2} + \mathbf{A_3} + \cdots + \mathbf{A_n} = \mathbf{I}_{nxn}$$
(2.2.4.14)

Now the identity matrix **I** is invertible as shown in equation (2.2.4.4). ∴ **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  $\alpha$ ,  $\beta$  in W and each scalar  $c \in F$ , the vector  $c\alpha + \beta \in W$ .

Let the matrices  $A_1$  and  $A_2$  satisfy,

$$\mathbf{A_1B} = \mathbf{BA_1} \tag{2.2.4.15}$$

$$\mathbf{A_2B} = \mathbf{BA_2} \tag{2.2.4.16}$$

Let,  $c \in \mathbf{F}$  be any constant.

$$(cA_1 + A_2)B = cA_1B + A_2B$$
 (2.2.4.17)

Substituting from equations (2.2.4.15) and (2.2.4.16) to (2.2.4.17),

$$\Rightarrow (c\mathbf{A}_1 + \mathbf{A}_2) \mathbf{B} = c\mathbf{B}\mathbf{A}_1 + \mathbf{B}\mathbf{A}_2$$

$$(2.2.4.18)$$

$$\Rightarrow \mathbf{B}c\mathbf{A}_1 + \mathbf{B}\mathbf{A}_2$$

$$(2.2.4.19)$$

$$\Rightarrow \mathbf{B}(c\mathbf{A}_1 + \mathbf{A}_2)$$

$$(2.2.4.20)$$

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,

$$\mathbf{A}^2 = \mathbf{A} \tag{2.2.4.21}$$

$$\mathbf{B}^2 = \mathbf{B} \tag{2.2.4.22}$$

Now for them to be closed under addition,

$$(\mathbf{A} + \mathbf{B})^2 = \mathbf{A} + \mathbf{B} \tag{2.2.4.23}$$

Which is not always same. Example let,

$$\mathbf{A} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \tag{2.2.4.24}$$

$$\mathbf{B} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \tag{2.2.4.25}$$

Clearly,

$$\mathbf{A}^2 = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} = \mathbf{A}$$
(2.2.4.26)

$$\mathbf{B}^2 = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \mathbf{B}$$
(2.2.4.27)

Now,

$$\mathbf{A} + \mathbf{B} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

$$(2.2.4.28)$$

$$\Rightarrow (\mathbf{A} + \mathbf{B})^2 = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix}$$

$$\implies (\mathbf{A} + \mathbf{B})^2 = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix}$$
(2.2.4.29)

Hence, clearly from equations (2.2.4.28) and (2.2.4.29),

$$(\mathbf{A} + \mathbf{B})^2 \neq \mathbf{A} + \mathbf{B}$$
 (2.2.4.30)

 $\therefore$  the set of all A such that  $A^2 = A$  is not closed under addition. Hence, not a subspace of V.

- 2.2.5. 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  $c\mathbf{u} + d\mathbf{v} = \mathbf{x}$ 

$$\begin{pmatrix} cu_1 \\ cu_2 \end{pmatrix} + \begin{pmatrix} dv_1 \\ dv_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$
 (2.2.5.1)

$$\begin{pmatrix} u_1 & v_1 \\ u_2 & v_2 \end{pmatrix} \begin{pmatrix} c \\ d \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$
 (2.2.5.2)

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  $v_2 \neq 0$ . Then

$$A = \begin{pmatrix} u_1 & v_1 \\ u_2 & v_2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$
 (2.2.5.3)

Hence A is row equivalent to  $I_2$  and so A is invertible and so (2.2.5.2) 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  $\alpha$  and  $\beta$  be vectors in W and let c be a scalar. By definition of W, both  $\alpha$  and  $\beta$  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.

2.2.6. 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

$$\mathbf{W}_1 \subseteq \mathbf{W}_2 \quad or \quad \mathbf{W}_2 \subseteq \mathbf{W}_1$$
 (2.2.6.1)

Let us assume that

$$\mathbf{W}_1 \not\subseteq \mathbf{W}_2$$
 (2.2.6.2)

We need to show that

$$\mathbf{W}_2 \subseteq \mathbf{W}_1 \tag{2.2.6.3}$$

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,

$$\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_1 \cup \mathbf{W}_2 \tag{2.2.6.4}$$

$$\implies$$
  $\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_1 \quad or$  (2.2.6.5)

$$\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_2 \tag{2.2.6.6}$$

But,  $\mathbf{w}_1 + \mathbf{w}_2 \notin \mathbf{W}_2$  because for some vector  $-\mathbf{w}_2 \in \mathbf{W}_2$ ,

$$(\mathbf{w}_1 + \mathbf{w}_2) - \mathbf{w}_2 = \mathbf{w}_1 \notin \mathbf{W}_2$$
 (2.2.6.7)

Hence it must be that,  $\mathbf{w}_1 + \mathbf{w}_2 \in \mathbf{W}_1$  because for some vector  $-\mathbf{w}_1 \in \mathbf{W}_1$ ,

$$(\mathbf{w}_1 + \mathbf{w}_2) - \mathbf{w}_1 = w_2 \in \mathbf{W}_1$$
 (2.2.6.8)

Thus, we have shown that every vector  $\mathbf{w}_2$  in  $\mathbf{W}_2$  is also in  $\mathbf{W}_1$ . Hence,  $\mathbf{W}_2 \subseteq \mathbf{W}_1$ 

- 2.2.7. 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  $\alpha$ ,  $\beta$ in W and each scalar c in F the vector  $c\alpha + \beta$ is again in W.

Then,

$$\mathbf{h}(-x) = c\mathbf{u}(-x) + \mathbf{v}(-x)$$

$$= c\mathbf{u}(x) + \mathbf{v}(x) \qquad (2.2.7.1)$$

$$= \mathbf{h}(x)$$

From (2.2.7.1)

$$\implies \mathbf{h}(-x) = \mathbf{h}(x) \tag{2.2.7.2}$$

$$\implies$$
 **h**  $\in$  **V**<sub>e</sub> (2.2.7.3)

Let  $\mathbf{u}, \mathbf{v} \in \mathbf{V_o}$  and  $c \in \mathbf{R}$  and let  $\mathbf{h} = c\mathbf{u} + \mathbf{v}$ . Then,

$$\mathbf{h}(-x) = c\mathbf{u}(-x) + \mathbf{v}(-x)$$

$$= -c\mathbf{u}(x) - \mathbf{v}(x)$$

$$= -\mathbf{h}(x)$$
(2.2.7.4)

From (2.2.7.4)

$$\implies \mathbf{h}(-x) = -\mathbf{h}(x) \tag{2.2.7.5}$$

$$\implies$$
 **h**  $\in$  **V**<sub>0</sub> (2.2.7.6)

From (2.2.7.3) and (2.2.7.6),  $V_e$  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** (2.2.7.3), (2.2.7.6),(2.2.1.9),(2.2.2.14) proves 1, 2 and 3.

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 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  $\alpha$  in **V** there are unique vectors  $\alpha_1$  in  $W_1$  and  $\alpha_2$  in  $W_2$  such that

$$\alpha = \alpha_1 + \alpha_2 \tag{2.2.3.3}$$

**Solution:** Suppose, vectors  $\alpha_1$  and  $\alpha_2$  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  $\alpha_1$  and  $\alpha'_1$  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

$$\alpha_{1} - \alpha'_{1} = \alpha'_{2} - \alpha_{2} = 0$$
 (2.2.3.13)  
 $\implies \alpha_{1} = \alpha'_{1}$  (2.2.3.14)  
 $\alpha_{2} = \alpha'_{2}$  (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  $\alpha \in \mathbf{V}$ 

#### 2.3 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

$$\begin{pmatrix} 1 & 1 & 2 & 4 \\ 2 & -1 & -5 & 2 \\ 1 & -1 & -4 & 0 \\ 2 & 1 & 1 & 6 \end{pmatrix} \xrightarrow{R_2 \to R_2 - 2R_1} \begin{pmatrix} 1 & 1 & 2 & 4 \\ 0 & -3 & -9 & -6 \\ 0 & -2 & -6 & -4 \\ 2 & 1 & 1 & 6 \end{pmatrix}$$

$$(2.3.1.1)$$

vectors are not linearly independent.

Vectors are not linearly independent.

$$R_4 \leftarrow R_4 - 2R_1 \rightarrow R_2 \leftarrow R_4 \rightarrow R_4 \rightarrow R_2 \leftarrow R_4 \rightarrow R_4$$

$$\stackrel{R_4 \leftarrow R_2}{\longleftarrow} \begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & -3 & -9 & -6 \\
0 & -2 & -6 & -4
\end{pmatrix}$$
(2.3.1.3)

$$\stackrel{R_3 \leftarrow R_3 + 3R_2}{\underset{R_4 \leftarrow R_4 + 2R_2}{\longleftarrow}} \begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$
(2.3.1.4)

Therefore the rank = no. of pivot columns = 2 (less than no. of columns). Thus the fours vectors are not linearly independent.

2.3.2. Are the vectors

$$\alpha_1 = (1, 1, 2, 4), \alpha_2 = (2, -1, -5, 2)$$
 (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

$$\alpha_{1} - \alpha'_{1} = \alpha'_{2} - \alpha_{2} = \mathbf{0} \qquad (2.2.3.13)$$

$$\Rightarrow \alpha_{1} = \alpha'_{1} \qquad (2.2.3.14)$$

$$\alpha_{2} = \alpha'_{2} \qquad (2.2.3.15)$$

$$\alpha_{3} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{1} = \alpha'_{1} \qquad (2.2.3.14)$$

$$\alpha_{2} = \alpha'_{2} \qquad (2.2.3.15)$$

$$\alpha_{1} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{2} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{3} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{1} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{2} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{3} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{1} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{2} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\alpha_{3} = \alpha_{1} + \alpha_{2} \qquad (2.2.3.16)$$

$$\begin{array}{c|cccc}
R_4 \leftarrow R_4 - 2R_1 \\
\hline
R_2 \leftarrow R_4
\end{array}
\begin{array}{c|cccc}
0 & -1 & -3 & -2 \\
0 & -2 & -6 & -4 \\
0 & -3 & -9 & -6
\end{array}$$
(2.3.2.4)

$$\stackrel{R_4 \leftarrow R_2}{\leftarrow} \stackrel{1}{\leftarrow} \begin{pmatrix} 1 & 1 & 2 & 4 \\ 0 & 1 & 3 & 2 \\ 0 & -3 & -9 & -6 \\ 0 & -2 & -6 & -4 \end{pmatrix}$$
(2.3.2.5)

$$\xrightarrow{R_3 \leftarrow R_3 + 3R_2} \begin{pmatrix}
1 & 1 & 2 & 4 \\
0 & 1 & 3 & 2 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix}$$
(2.3.2.6)

Therefore the rank = no. of pivot columns = 2 (less than no. of columns). Thus the fours vectors are not linearly independent.

$$\alpha_1 = \begin{pmatrix} 1 & 1 & 2 & 4 \end{pmatrix} \tag{2.3.3.1}$$

$$\alpha_2 = \begin{pmatrix} 2 & -1 & -5 & 2 \end{pmatrix} \tag{2.3.3.2}$$

$$\alpha_1 = \begin{pmatrix} 1 & 1 & 2 & 4 \end{pmatrix}$$
 (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)

$$\alpha_4 = \begin{pmatrix} 2 & 1 & 1 & 6 \end{pmatrix} \tag{2.3.3.4}$$

**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,

$$\begin{pmatrix}
1 & 2 & 1 & 2 \\
1 & -1 & -1 & 1 \\
2 & -5 & -4 & 1 \\
4 & 2 & 0 & 6
\end{pmatrix}
\xrightarrow{R_2 = R_2 - R_1}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & -3 & -2 & -1 \\
0 & -9 & -6 & -3 \\
4 & 2 & 0 & 6
\end{pmatrix}$$

$$(2.3.3.6)$$

$$\xrightarrow{R_4 = R_4 - R_1}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & -3 & -2 & -1 \\
0 & -9 & -6 & -3 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$(2.3.3.7)$$

$$\xrightarrow{R_2 = -\frac{1}{3}R_2}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & -9 & -6 & -3 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$(2.3.3.8)$$

$$\xrightarrow{R_3 = R_3 - 9R_2}
\begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & 0 & 0 & 0 \\
0 & -6 & -4 & -2
\end{pmatrix}$$

$$\stackrel{R_4=R_4+6R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 2 & 1 & 2 \\
0 & 1 & -\frac{2}{3} & \frac{1}{3} \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{pmatrix} (2.3.3.10)$$

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  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$ ,  $\alpha_4$ 

Hence the required basis vectors are,

$$\mathbf{a_1} = \begin{pmatrix} 1 & 1 & 2 & 4 \end{pmatrix} \tag{2.3.3.11}$$

$$\mathbf{a_2} = \begin{pmatrix} 2 & -1 & -5 & 2 \end{pmatrix} \tag{2.3.3.12}$$

2.3.4. Let V be the vector space of all  $2\times 2$  matrices over the field  $\mathbb{F}$ . Let  $W_1$  be the set of matrices of the form

$$\begin{pmatrix} x & -x \\ y & z \end{pmatrix} \tag{2.3.4.1}$$

and let  $W_2$  be the set of matrices of the form

$$\begin{pmatrix} a & b \\ -a & c \end{pmatrix} \tag{2.3.4.2}$$

- 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  $\alpha$ ,  $\beta$  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,

$$A_1 = \begin{pmatrix} x_1 & -x_1 \\ y_1 & z_1 \end{pmatrix}, A_2 = \begin{pmatrix} x_2 & -x_2 \\ y_2 & z_2 \end{pmatrix}$$
 (2.3.4.3)

Let  $c \in F$  then,

$$cA_1 + A_2 = \begin{pmatrix} cx_1 + x_2 & -cx_1 - x_2 \\ cy_1 + y_2 & cz_1 + z_2 \end{pmatrix} = \begin{pmatrix} u & -u \\ v & w \end{pmatrix}$$
(2.3.4.4)

Thus  $cA_1 + A_2 \in W_1$ . Hence  $W_1$  is a subspace. Similarly, let  $A_1, A_2 \in W_2$  where,

$$A_1 = \begin{pmatrix} a_1 & b_1 \\ -a_1 & c_1 \end{pmatrix}, A_2 = \begin{pmatrix} a_2 & b_2 \\ -a_2 & c_2 \end{pmatrix}$$
 (2.3.4.5)

Let  $c \in F$  then,

$$cA_1 + A_2 = \begin{pmatrix} ca_1 + a_2 & cb_1 + b_2 \\ -ca_1 - a_2 & cc_1 + c_2 \end{pmatrix} = \begin{pmatrix} u & v \\ -u & w \end{pmatrix}$$
(2.3.4.6)

Thus  $cA_1 + A_2 \in W_2$ . Hence  $W_2$  is a subspace.

b) The subspace  $W_1$  can be given as,

$$\begin{pmatrix} x & -x \\ y & z \end{pmatrix} = x \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + z \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

$$= xA_1 + yA_2 + zA_2$$

$$(2.3.4.8)$$

Now.

$$x \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + z \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$(2.3.4.9)$$

$$\implies x = y = z = 0$$

$$(2.3.4.10)$$

 $A_1, A_2, A_3$  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,

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

$$= aA_1 + bA_2 + cA_2 \qquad (2.3.4.12)$$

Now,

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

$$(2.3.4.13)$$

$$\Rightarrow a = b = c = 0$$

$$(2.3.4.14)$$

 $A_1, A_2, A_3$  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,

$$\begin{pmatrix} x+a & -x+b \\ y-a & z+c \end{pmatrix}$$
 (2.3.4.15)

For  $x + a \neq -x + b \neq y - a \neq z + c$ ,

$$\begin{pmatrix} x+a & -x+b \\ y-a & z+c \end{pmatrix} = \begin{pmatrix} j & k \\ l & m \end{pmatrix}$$
 (2.3.4.16)  
=  $j \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + k \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + l \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + m \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$  (2.3.4.17)

$$= jA_1 + kA_2 + lA_3 + mA_4 (2.3.4.18)$$

Now,

$$jA_1 + kA_2 + lA_3 + mA_4 = 0$$
 (2.3.4.19)  
 $\implies j = k = l = m = 0$  (2.3.4.20)

 $A_1, A_2, A_3, A_4$  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,

$$\begin{pmatrix} x & -x \\ -x & y \end{pmatrix} = x \begin{pmatrix} 1 & -1 \\ -1 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

$$= xA_1 + yA_2 \qquad (2.3.4.21)$$

Now,

$$x \begin{pmatrix} 1 & -1 \\ -1 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \quad (2.3.4.23)$$

$$\implies x = y = 0 \qquad (2.3.4.24)$$

 $A_1, A_2$  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.

2.3.5. Let **V** be the space of  $2 \times 2$  matrices over **F**. Find a basis  $\{\mathbf{A}_1, \mathbf{A}_2, \mathbf{A}_3, \mathbf{A}_4\}$  for **V** such that  $\mathbf{A}_j^2 = \mathbf{A}_j$  for each j

**Solution:** Every 2×2 matrix may be written as

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

$$(2.3.5.1)$$

This shows that

$$\{\mathbf{A}_{1}, \mathbf{A}_{2}, \mathbf{A}_{3}, \mathbf{A}_{4}\} = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$
(2.3.5.2)

can be the basis for the space V of all  $2 \times 2$  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

$$\begin{pmatrix} a & 0 \\ 0 & d \end{pmatrix} \tag{2.3.5.3}$$

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,

$$\mathbf{A}_2 = \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} \tag{2.3.5.4}$$

$$\mathbf{A}_3 = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} \tag{2.3.5.5}$$

Here,  $\mathbf{A}_2^2 = \mathbf{A}_2$  and  $\mathbf{A}_3^2 = \mathbf{A}_3$ . Therefore the basis can be

$$\{\mathbf{A}_{1}, \mathbf{A}_{2}, \mathbf{A}_{3}, \mathbf{A}_{4}\} = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$
(2.3.5.6)

 $\{A_1, A_2, A_3, A_4\}$  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.

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

$$(2.3.5.7)$$

$$\implies a + b = 0$$

$$(2.3.5.8)$$

$$b = 0$$

$$(2.3.5.9)$$

$$c = 0$$

$$(2.3.5.10)$$

$$c + d = 0$$

$$(2.3.5.11)$$

The corresponding matrix form is Ax = 0

$$\begin{pmatrix} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$
 (2.3.5.12)

Row reducing the augmented matrix,

$$\begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \end{pmatrix} \xrightarrow{R_2 \longleftrightarrow R_3} \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ R_4 \leftarrow R_4 - R_3 \end{pmatrix} \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix} (2.3.5.13)$$

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

$$(2.3.5.14)$$

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,

$$\begin{pmatrix} w & x \\ y & z \end{pmatrix} \tag{2.3.5.15}$$

Compute a, b, c, d such that

$$\begin{pmatrix} w & x \\ y & z \end{pmatrix} = a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$
(2.3.5.16)

$$= \begin{pmatrix} a+b & c \\ b & c+d \end{pmatrix} \tag{2.3.5.17}$$

Equating the entries, this produces system of linear equations,

$$a + b = w, y = b, x = c, z = c + d$$
 (2.3.5.18)

$$\implies a = w - y$$
 (2.3.5.19) 2.3.7

$$b = y (2.3.5.20)$$

$$c = x$$
 (2.3.5.21)

$$d = z - x \tag{2.3.5.22}$$

In particular, there exists at least one solution regardless of the values of w, x, y, z. For example, consider the following matrix,

$$\begin{pmatrix} w & x \\ y & z \end{pmatrix} = \begin{pmatrix} 3 & 4 \\ -2 & 7 \end{pmatrix} \tag{2.3.5.23}$$

Here, a = 5, b = -2, c = 4, d = 3. Using

(2.3.5.16), we get

$$5\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} - 2\begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} + 4\begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} + 3\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 3 & 4 \\ -2 & 7 \end{pmatrix}$$
(2.3.5.24)

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  $V$ .

2.3.6. Let **V** be a vector space over a subfield **F** of complex numbers. Suppose  $\alpha$ ,  $\beta$  and  $\gamma$  are linearly independent vectors in **V**. Prove that  $(\alpha+\beta)$ , $(\beta+\gamma)$  and  $(\gamma+\alpha)$  are linearly independent.

**Solution:** Let  $\alpha$ ,  $\beta$  and  $\gamma$  be three n× 1 dimensional vectors. We need to prove that,

$$(\alpha + \beta \quad \beta + \gamma \quad \gamma + \alpha)\mathbf{x} = 0 \qquad (2.3.6.1)$$

will only have a trivial solution. The above equation can be written as

$$\begin{pmatrix} \alpha & \beta & \gamma \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \mathbf{x} = 0 \qquad (2.3.6.2)$$

$$\mathbf{x}^T \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} \alpha^T \\ \boldsymbol{\beta}^T \\ \boldsymbol{\gamma}^T \end{pmatrix} = 0 \qquad (2.3.6.3)$$

Since,  $\alpha$ ,  $\beta$  and  $\gamma$  are independent.

$$\mathbf{x}^T \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} = 0 \tag{2.3.6.4}$$

In the above equation we can see that the  $3 \times 3$  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.

(2.3.5.19) 2.3.7. Prove that the space of all  $m \times n$  matrices over (2.3.5.20) the field  $\mathbf{F}$  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,

$$\mathbf{M}_{ij} = \begin{cases} 0 & m \neq i, n \neq j \\ 1 & m = i, n = j \end{cases}$$
 (2.3.7.1)

For example,

$$\mathbf{M}_{12} = \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 \end{pmatrix}_{mxn}$$
 (2.3.7.2)

(2.3.7.3)

Let  $A \in M$  given as,

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \cdots & a_{mn} \end{pmatrix} \tag{2.3.7.4}$$

Now clearly,

$$\mathbf{a}_{11} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}$$

$$(2.3.7.5)$$

$$\implies$$
  $\mathbf{a}_{11} = \mathbf{A}\mathbf{M}_{11}$  (2.3.7.6)

$$\therefore \mathbf{A} = \sum_{i=1}^{m} \sum_{j=1}^{n} a_{ij} M_{ij}$$
 (2.3.7.7)

 $\implies$  **M**<sub>ij</sub> span **M**. Also from the above equation **A**= 0 if and only if all elements are zero, that is,

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} = \begin{pmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}$$
(2.3.7.8)

$$\implies a_{ij} = 0 \tag{2.3.7.9}$$

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.

2.3.8. Let V be a vector space over the field  $F = \{0, 1\}$ . Suppose  $\alpha$ ,  $\beta$  and  $\gamma$  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,

$$0 + 0 = 0$$
  
 $1 + 1 = 0$  (2.3.8.1)

A set are vectors  $\{v_1,v_2,v_3\}$  are linearly independent if

$$a\mathbf{v_1} + b\mathbf{v_2} + c\mathbf{v_3} = 0 \tag{2.3.8.2}$$

has only one trivial solution

$$a = b = c = 0 (2.3.8.3)$$

Now,

$$a(\alpha + \beta) + b(\beta + \gamma) + c(\gamma + \alpha) = 0 \quad (2.3.8.4)$$

$$\implies (a+c)\alpha + (a+b)\beta + (b+c)\gamma = 0$$
(2.3.8.5)

Writing (2.3.8.5) in matrix form,

$$(\alpha \quad \beta \quad \gamma) \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \mathbf{x} = 0$$
 (2.3.8.6)

where,

$$\mathbf{x} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$$

$$\mathbf{x}^T \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = 0 \tag{2.3.8.7}$$

Since  $\alpha$ ,  $\beta$  and  $\gamma$  are linearly independent vectors,

$$\mathbf{x}^{T} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \end{pmatrix} \tag{2.3.8.8}$$

Transposing on both sides,

$$\begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \mathbf{x} = 0 \tag{2.3.8.9}$$

By using the properties from (2.3.8.1) and

reducing (2.3.8.9) to row echelon form,

$$\begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \xrightarrow{R_2 \leftarrow R_1 + R_2} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

$$\xrightarrow{R_3 \leftarrow R_2 + R_3} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix} \qquad (2.3.8.10)$$

Expressing (2.3.8.10) as a linear combination of vectors,

$$a \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + b \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + c \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\implies \begin{pmatrix} a+c \\ b+c \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\implies a+c=0; \quad b+c=0 \qquad (2.3.8.11)$$

The solutions to (2.3.8.11) are,

$$a = b = c = 0;$$
  $a = b = c = 1$  (2.3.8.12)

Since there is no trivial solution,  $(\alpha + \beta)$ ,  $(\beta + \gamma)$  and  $(\gamma + \alpha)$  are linearly dependent

2.3.9. 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

$$\mathbf{v} = \sum_{i=0}^{n-1} c_i \alpha_i$$
 (2.3.9.1)

where 
$$c_i \in \mathbb{Q}$$
 (2.3.9.2)

and 
$$\mathbf{B} = \{\alpha_0, \alpha_1, \dots, \alpha_{n-1}\}\$$
 (2.3.9.3)

is the basis with linearly independent  $\alpha_i$  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

$$\beta_0 + \beta_1 x = 0 \tag{2.3.9.4}$$

$$\implies x = -\frac{\beta_0}{\beta_1} \tag{2.3.9.5}$$

But x is irrational and  $-\frac{\beta_0}{\beta_1}$  is rational so (2.3.9.5) can't be possible so  $\beta_0, \beta_1 = 0$ Hence  $\{1, x\}$  are independent. Similarly for the set  $\{1, x, x^2\}$  for  $\beta_0, \beta_1, \beta_2 \in \mathbb{Q}$ 

$$\beta_0 + \beta_1 x + \beta_2 x^2 = 0 \tag{2.3.9.6}$$

 $\beta_1 x + \beta_2 x^2$  is irrational and  $\beta_0$  is rational. Therefore

$$\beta_0 = 0 \tag{2.3.9.7}$$

and 
$$\beta_1 x + \beta_2 x^2 = 0$$
,  $(x \neq 0)$  (2.3.9.8)

$$\implies \beta_1 + \beta_2 x = 0 \qquad (2.3.9.9)$$

$$\implies \beta_1, \beta_2 = 0$$
 (2.3.9.10)

$$\therefore \beta_0 + \beta_1 x + \beta_2 x^2 = 0 \qquad (2.3.9.11)$$

$$\iff \beta_0, \beta_1, \beta_2 = 0 \qquad (2.3.9.12)$$

Hence  $\{1, x, x^2\}$  are independent

By induction, let us say the set  $\{1, x, x^2, \dots, x^n\}$  is independent

for 
$$\beta_0, \beta_1, \beta_2, \dots, \beta_n \in \mathbb{Q}$$
 (2.3.9.13)

$$\beta_0 + \beta_1 x + \beta_2 x^2 + \dots + \beta_n x^n = 0$$
 (2.3.9.14)

$$\iff \beta_0, \beta_1, \beta_2, \dots, \beta_n = 0 \quad (2.3.9.15)$$

To prove this for the set  $A = \{1, x, x^2, \dots, x^{n+1}\}$ 

for 
$$\beta_0, \beta_1, \beta_2, \dots, \beta_n, \beta_{n+1} \in \mathbb{Q}$$
 (2.3.9.16)

$$\beta_0 + \beta_1 x + \dots + \beta_n x^n + \beta_{n+1} x^{n+1} = 0$$
(2.3.9.17)

Comparing to (2.3.9.7) and (2.3.9.8)

$$\beta_0 = 0 \qquad (2.3.9.18)$$

$$\beta_1 + \beta_2 x + \ldots + \beta_{n+1} x^n = 0$$
 (2.3.9.19)

Comparing with (2.3.9.14),we have  $\beta_1, \beta_2, \dots, \beta_{n+1} = 0$ 

$$\therefore \beta_0 + \beta_1 x + \dots + \beta_n x^n + \beta_{n+1} x^{n+1} = 0$$
(2.3.9.20)

$$\iff \beta_0, \beta_1, \beta_2, \dots, \beta_n, \beta_{n+1} = 0$$

$$(2.3.9.21)$$

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  $\mathbb Q$ 

#### 2.4 Coordinates

#### 2.4.1. Show that the vectors

$$\alpha_1 = \begin{pmatrix} 1 & 1 & 0 & 0 \end{pmatrix}$$
  $\alpha_2 = \begin{pmatrix} 0 & 0 & 1 & 1 \end{pmatrix}$  (2.4.1.1)  
 $\alpha_3 = \begin{pmatrix} 1 & 0 & 0 & 4 \end{pmatrix}$   $\alpha_4 = \begin{pmatrix} 0 & 0 & 0 & 2 \end{pmatrix}$  (2.4.1.2)

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 2.1.** Let V be an n-dimensional vector space over the field F, and let  $\beta$  and  $\beta'$  be two ordered basis of V. Then, there is a unique, necessarily invertible,  $n \times n$  matrix P with entries in F such that

$$\begin{array}{ll} a) & \left[\alpha\right]_{\beta} = \mathbf{P} \left[\alpha\right]_{\beta'} \\ b) & \left[\alpha\right]_{\beta'} = \mathbf{P}^{-1} \left[\alpha\right]_{\beta} \end{array}$$

for every vector  $\alpha$  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  $\alpha_4$  are basis for  $\Re^4$ . For, this we first show that  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$  and  $\alpha_4$  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,

$$\begin{pmatrix} 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 4 & 2 \end{pmatrix} \xrightarrow{r_2 = r_2 - r_1} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 4 & 2 \end{pmatrix} \xrightarrow{(2.4.1.6)} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 4 & 2 \end{pmatrix} \xrightarrow{(2.4.1.7)} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 4 & 2 \end{pmatrix} \xrightarrow{r_2 \leftrightarrow r_3} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 1 & 4 & 2 \end{pmatrix} \xrightarrow{(2.4.1.7)} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 4 & 2 \end{pmatrix} \xrightarrow{(2.4.1.8)} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 4 & 2 \end{pmatrix} \xrightarrow{(2.4.1.8)} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 4 & 2 \end{pmatrix} \xrightarrow{(2.4.1.9)} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 2 \end{pmatrix} \xrightarrow{(2.4.1.10)} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 2 \end{pmatrix} \xrightarrow{(2.4.1.11)} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 2 \end{pmatrix} \xrightarrow{(2.4.1.12)} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \xrightarrow{(2.4.1.12)} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

(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  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$  and  $\alpha_4$  are linearly independent and their column space is  $\Re^4$  which means vectors  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$  and  $\alpha_4$  span  $\Re^4$ . Hence, vectors  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$  and  $\alpha_4$  form a basis for  $\Re^4$ .

Now, we use theorem (2.1), 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)

$$\begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & 4 & 2 & 0 & 0 & 0 & 1 \end{pmatrix} \xrightarrow{r_4 = r_4 - r_2} \xrightarrow{r_4 = r_4 - r_2}$$

$$\begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 0 & 4 & 2 & 0 & 0 & -1 & 1 \end{pmatrix} (2.4.1.17)$$

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & -1 & 0 & -1 & 1 & 0 & 0 \\
0 & 0 & 4 & 2 & 0 & 0 & -1 & 1
\end{pmatrix}
\xrightarrow{r_3 = -r_3}$$

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

$$\begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 1 & -1 & 0 & 0 \\
0 & 0 & 0 & 2 & -4 & 4 & -1 & 1
\end{pmatrix}
\xrightarrow{r_1 = r_1 - r_3}$$

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

Now, let  $\mathbf{e_1} = \begin{pmatrix} 1 & 0 & 0 & 0 \end{pmatrix}$ ,  $\mathbf{e_2} = \begin{pmatrix} 0 & 1 & 0 & 0 \end{pmatrix}$ ,  $\mathbf{e_3} = \begin{pmatrix} 0 & 0 & 1 & 0 \end{pmatrix}$  and  $\mathbf{e_4} = \begin{pmatrix} 0 & 0 & 0 & 1 \end{pmatrix}$  be

the standard basis for  $\Re^4$ . Hence,

$$\mathbf{e}_1 = \alpha_3 - 2\alpha_4 \tag{2.4.1.23}$$

$$\mathbf{e_2} = \alpha_1 - \alpha_3 + 2\alpha_4 \tag{2.4.1.24}$$

$$\mathbf{e_3} = \alpha_2 - \frac{1}{2}\alpha_4 \tag{2.4.1.25}$$

$$\mathbf{e_4} = \frac{1}{2}\alpha_4 \tag{2.4.1.26}$$

2.4.2. Find the coordinate matrix of the vector  $\begin{pmatrix} 1 & 0 & 1 \end{pmatrix}$  in the basis of  $C^3$  consisting of the vectors  $\begin{pmatrix} 2i & 1 & 0 \end{pmatrix}$ ,  $\begin{pmatrix} 2 & -1 & 1 \end{pmatrix}$ ,  $\begin{pmatrix} 0 & 1+i & 1-i \end{pmatrix}$  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  $\alpha_i$  by row reducing augmented matrix.

$$\begin{pmatrix} 2i & 2 & 0 & 1 \\ 1 & -1 & 1+i & 0 \\ 0 & 1 & 1-i & 1 \end{pmatrix} \xrightarrow{R_1 \to R_2} \begin{pmatrix} 1 & -1 & 1+i & 0 \\ 0 & 2+2i & 2-2i & 1 \\ 0 & 1 & 1-i & 1 \end{pmatrix}$$

$$(2.4.2.3)$$

$$\stackrel{R_2 \leftarrow R_2/2}{\underset{R_3 \leftarrow R_3 - R_2}{\longleftrightarrow}} \begin{pmatrix} 1 & -1 & 1+i & 0 \\ 0 & 1+i & 1-i & \frac{1}{2} \\ 0 & -i & 0 & \frac{1}{2} \end{pmatrix}$$
(2.4.2.4)

Therefore the coordinate matrix of the vector is

$$\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = \begin{pmatrix} \frac{-1-i}{2} \\ \frac{1}{2} \\ \frac{3+i}{4} \end{pmatrix}$$
 (2.4.2.5)

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

$$\alpha_1 = \begin{pmatrix} 1 & 0 & -1 \end{pmatrix}, \alpha_2 = \begin{pmatrix} 1 & 1 & 1 \end{pmatrix}, \alpha_3 = \begin{pmatrix} 1 & 0 & 0 \end{pmatrix}.$$

What are the coordinates of vector  $\begin{pmatrix} a & b & c \end{pmatrix}$  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  $R^3$ , then the coordinates of vector,

$$\alpha = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{2.4.3.2}$$

in the ordered basis  $R^3$  is the vector,

$$[\alpha]_{\mathbf{B}} = \begin{pmatrix} x_1 \\ x_2 \\ x_2 \end{pmatrix} \tag{2.4.3.3}$$

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

$$\begin{pmatrix} 1 & 1 & 1 & a \\ 0 & 1 & 0 & b \\ -1 & 1 & 0 & c \end{pmatrix} \tag{2.4.3.6}$$

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)

$$\stackrel{R_1 \leftarrow R_1 - R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 1 & a - b \\ 0 & 1 & 0 & b \\ 0 & 0 & 1 & a - 2b + c \end{pmatrix}$$
(2.4.3.9)

$$\stackrel{R_1 \leftarrow R_1 - R_3}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & b - c \\ 0 & 1 & 0 & b \\ 0 & 0 & 1 & a - 2b + c \end{pmatrix}$$
(2.4.3.10)

 $\therefore$  The coordinates of  $\alpha$  w.r.t **B** is

$$[\alpha]_{\mathbf{B}} = \begin{pmatrix} b - c \\ b \\ a - 2b + c \end{pmatrix} \tag{2.4.3.11}$$

- 2.4.4. Let **W** be the subspace of  $\mathbb{C}^3$  spanned by  $\alpha_1 = \begin{pmatrix} 1 \\ 0 \\ i \end{pmatrix}$  and  $\alpha_2 = \begin{pmatrix} 1 \\ i \\ 1+i \end{pmatrix}$ .
  - a) Show that  $\alpha_1$  and  $\alpha_2$  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  $\alpha_1$  and  $\alpha_2$  in the ordered basis  $\{\beta_1, \beta_2\}$  for **W**.

#### **Solution:**

a) It is given that  $\alpha_1$  and  $\alpha_2$  span **W**. For  $\alpha_1$  and  $\alpha_2$  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  $\mathbf{A} = \begin{pmatrix} \alpha_1 & \alpha_2 \end{pmatrix}$ 

$$\begin{pmatrix}
1 & 1+i \\
0 & 1 \\
i & -1
\end{pmatrix}
\xrightarrow{R_3 \leftarrow R_3 - iR_1}
\begin{pmatrix}
1 & 1+i \\
0 & 1 \\
0 & -i
\end{pmatrix}
(2.4.4.2)$$

$$\xrightarrow{R_3 \leftarrow R_3 + iR_2}
\begin{pmatrix}
1 & 1+i \\
0 & 1 \\
0 & 0
\end{pmatrix}
(2.4.4.3)$$

$$\xrightarrow{R_1 \leftarrow R_1 - (i+1)R_2}
\begin{pmatrix}
1 & 0 \\
0 & 1 \\
0 & 0
\end{pmatrix}
(2.4.4.4)$$

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  $\mathbf{W}$  and if  $\beta_1$  and  $\beta_2 \in \mathbf{W}$  there exist a unique solution  $\mathbf{x}$  such that

$$(\alpha_1 \quad \alpha_2) \mathbf{x} = (\beta_1 \quad \beta_2) \tag{2.4.4.7}$$

Using row reduction on augmented matrix

$$\begin{pmatrix} 1 & 1+i & | & 1 & 1 \\ 0 & 1 & | & 1 & i \\ i & -1 & | & 0 & 1+i \end{pmatrix} (2.4.4.8)$$

$$\xrightarrow{R3 \leftarrow R_3 - iR - 1} \begin{pmatrix} 1 & 1 + i & | & 1 & 1 \\ 0 & 1 & | & 1 & i \\ 0 & -i & | & -i & 1 \end{pmatrix} (2.4.4.9)$$

$$\xrightarrow{R_3 \leftarrow R_3 + iR_2} \begin{pmatrix} 1 & 1 + i & | & 1 & 1 \\ 0 & 1 & | & 1 & i \\ 0 & 0 & | & 0 & 0 \end{pmatrix}$$

(2.4.4.10)

$$\xrightarrow{R_1 \leftarrow R_1 - (i+1)R_2} \begin{pmatrix} 1 & 0 & | & -i & 2-i \\ 0 & 1 & | & 1 & i \\ 0 & 0 & | & 0 & 0 \end{pmatrix}$$
(2.4.4.11)

$$\implies \mathbf{x} = \begin{pmatrix} -i & 2-i \\ 1 & i \end{pmatrix}$$
 (2.4.4.12)

Therefore  $\beta_1$  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

$$\mathbf{B} = \begin{pmatrix} 1 & 1 \\ 1 & i \\ 0 & 1+i \end{pmatrix} \tag{2.4.4.14}$$

Using row reduction on matrix **B** 

$$\begin{pmatrix}
1 & 1 \\
1 & i \\
0 & 1+i
\end{pmatrix}
\xrightarrow{R_2 \leftarrow R_2 - R_1}
\begin{pmatrix}
1 & 1 \\
0 & i-1 \\
0 & 1+i
\end{pmatrix}
(2.4.4.15)$$

$$\xrightarrow{R_2 \leftarrow \frac{R_2}{i-1}}
\begin{pmatrix}
1 & 1 \\
0 & 1 \\
0 & 1
\end{pmatrix}
(2.4.4.16)$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_2}
\begin{pmatrix}
1 & 0 \\
0 & 1 \\
0 & 0
\end{pmatrix}
(2.4.4.17)$$

Since **B** is a full rank matrix the column vectors are linearly independent.

Let  $\alpha$  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  $\mathbf{x_2}$  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  $\mathbf{W}$ 

c) Since  $\alpha_1, \alpha_2 \in \mathbf{W}$  and  $\{\beta_1, \beta_2\}$  are ordered basis for  $\mathbf{W}$  there must exist unique value of  $\mathbf{x}$  such that

$$(\beta_1 \quad \beta_2) \mathbf{x} = (\alpha_1 \quad \alpha_2) \tag{2.4.4.22}$$

Using row reduction on (2.4.4.22) we get,

$$\begin{pmatrix} 1 & 1 & | & 1 & 1+i \\ 1 & i & | & 0 & 1 \\ 0 & 1+i & | & i & -1 \end{pmatrix}$$

$$(2.4.4.23)$$

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

$$(2.4.4.24)$$

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

$$(2.4.4.25)$$

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

$$(2.4.4.26)$$

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

$$(2.4.4.27)$$

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

$$(2.4.4.28)$$

Thus the column vectors of (2.4.4.28) are corresponding coordinates of  $\alpha_1$  and  $\alpha_2$  in ordered basis  $\{\beta_1, \beta_2\}$ .

expressed as span $\{\beta_1, \beta_2\}$ . Therefore  $\{\beta_1, \beta_2\}$  is basis for **W**. 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  $\mathbb{R}^2$  Since  $\alpha_1, \alpha_2 \in \mathbf{W}$  and  $\{\beta_1, \beta_2\}$  are ordered 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

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  $\alpha$  and  $\beta$  say, geometrically, that  $\alpha$  and  $\beta$  are perpendicular and each has length 1).

**Solution:** we need to show that  $\alpha$  and  $\beta$  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.  $\mathbf{A}^T$  is the inverse of **A**. Thus, the columns of **A** are linearly independent i.e,  $\alpha$  and  $\beta$  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 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  $\beta$ 

**Solution:** We start by taking,

$$\mathbf{f} = \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} \tag{2.4.6.1}$$

Let's start by proving that g is linearly inde-

pendent.

$$\mathbf{g} = \mathbf{Bf} \tag{2.4.6.2}$$

where,

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 \\ t & 1 & 0 \\ t^2 & 2t & 1 \end{pmatrix} \tag{2.4.6.3}$$

Now,

$$\mathbf{v}^T \mathbf{g} = 0 \tag{2.4.6.4}$$

$$\implies \mathbf{v}^T \mathbf{B} \mathbf{f} = 0 \tag{2.4.6.5}$$

Since  $\mathbf{f}$  is linearly independent,

$$\mathbf{v}^T \mathbf{B} = 0 \tag{2.4.6.6}$$

$$\mathbf{B}^T \mathbf{v} = 0 \tag{2.4.6.7}$$

Since  $\mathbf{B}^T$  is an upper triangular matrix with non zero values in principal diagonal, it is invertible matrix and hence  $\mathbf{v}$  will be zero vector. Now, Finding the inverse of  $\mathbf{B}^T$ 

$$\begin{pmatrix}
1 & t & t^2 & 1 & 0 & 0 \\
0 & 1 & 2t & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 0 & 1
\end{pmatrix} (2.4.6.8)$$

$$\stackrel{R_1=R_1-tR_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & -t^2 & 1 & -t & 0 \\
0 & 1 & 2t & 0 & 1 & 0 \\
0 & 0 & 1 & 0 & 0 & 1
\end{pmatrix} (2.4.6.9)$$

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

$$(2.4.6.10)$$

$$\stackrel{R_2=R_2-2tR_3}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 0 & 1 & -t & t^2 \\ 0 & 1 & 0 & 0 & 1 & -2t \\ 0 & 0 & 1 & 0 & 0 & 1 \end{pmatrix}$$

$$(2.4.6.11)$$

So.

$$(\mathbf{B}^T)^{-1} = \begin{pmatrix} 1 & -t & t^2 \\ 0 & 1 & -2t \\ 0 & 0 & 1 \end{pmatrix}$$
 (2.4.6.12)

Now, to find the coordinates,

$$f(x) = \mathbf{w}^T \mathbf{g} \tag{2.4.6.13}$$

So,

$$\mathbf{c}^T \mathbf{f} = \mathbf{w}^T \mathbf{g} \tag{2.4.6.14}$$

$$\mathbf{c}^T \mathbf{f} = \mathbf{w}^T \mathbf{B} \mathbf{f} \tag{2.4.6.15}$$

$$(\mathbf{c}^T - \mathbf{w}^T \mathbf{B})\mathbf{f} = 0 (2.4.6.16)$$

Since, **f** is linearly independent,

$$\mathbf{c}^T - \mathbf{w}^T \mathbf{B} = 0 \tag{2.4.6.17}$$

$$\mathbf{c}^T = \mathbf{w}^T \mathbf{B} \tag{2.4.6.18}$$

$$\mathbf{c}^T \mathbf{B}^{-1} = \mathbf{w}^T \tag{2.4.6.19}$$

$$(\mathbf{B}^{-1})^T \mathbf{c} = \mathbf{w} \tag{2.4.6.20}$$

Using (2.4.6.12) in (2.4.6.20)

$$\mathbf{w} = \begin{pmatrix} 1 & -t & t^2 \\ 0 & 1 & -2t \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} c_0 \\ c_1 \\ c_2 \end{pmatrix}$$
 (2.4.6.21)

# 2.5 Summary of Row Equivalence

2.5.1. Consider the vectors in  $\mathbb{R}^4$  defined by:

$$\alpha_1 = \begin{pmatrix} -1\\0\\1\\2 \end{pmatrix}, \ \alpha_2 = \begin{pmatrix} 3\\4\\-2\\5 \end{pmatrix} \text{ and } \alpha_3 = \begin{pmatrix} 1\\4\\0\\9 \end{pmatrix}.$$

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  $\mathbf{AX} = 0$  and  $\mathbf{RX} = 0$  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)

 $\rho_1$ ,  $\rho_2$  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:

$$\begin{pmatrix} 1 & 0 & x_1 \\ 0 & 4 & x_2 \\ -1 & 1 & x_3 \\ -2 & 11 & x_4 \end{pmatrix} \xrightarrow{R_3 = 4R_3 + 4R_1 - R_2} (2.5.1.8)$$

$$\begin{pmatrix} 1 & 0 & x_1 \\ 0 & 4 & x_2 \\ 0 & 0 & 4x_1 - x_2 + 4x_3 \\ -2 & 11 & x_4 \end{pmatrix} \xrightarrow{R_4 = 4R_4 + 8R_1 - 11R_2} (2.5.1.9)$$

$$\begin{pmatrix} 1 & 0 & x_1 \\ 0 & 4 & x_2 \\ 0 & 0 & 4x_1 - x_2 + 4x_3 \\ 0 & 0 & 8x_1 - 11x_2 + 4x_4 \end{pmatrix}$$

Using 2.5.1.10, The required homogeneous equation is given as:

$$\begin{pmatrix} 4 & -1 & 4 & 0 \\ 8 & -11 & 0 & 4 \end{pmatrix} \mathbf{X} = 0$$
 (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  $\alpha$ ,  $\beta$ ,  $\gamma$  are in the subspace of  $\mathbb{R}^4$  spanned by  $\alpha_i$ ?
- b) Which of the vectors  $\alpha$ ,  $\beta$ ,  $\gamma$  are in the subspace of  $\mathbb{C}^4$  spanned by  $\alpha_i$ ?
- c) Does this suggest a theorem?

#### **Solution:**

a) The linear combination of  $\alpha_i$  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  $\alpha_i$  are independent as the rank of matrix is 3. Hence  $\alpha_i$  forms basis for subspace S.

i) Checking for  $\alpha$ : 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),

$$(\mathbf{A} \quad \alpha) = \begin{pmatrix} 1 & 0 & 0 & -3 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$
 (2.5.3.10)

As Rank( $(\mathbf{A} \ \alpha)$ )=Rank( $(\mathbf{A})$ )=3, the vector  $\alpha$  can be represented as linear combination of  $\alpha_i$ . 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  $\alpha$  is in the subspace S.

ii) Checking for  $\beta$ : To check if a solution by  $\alpha_i$ , i=1,2,3...,n. exists for  $AX = \beta$ . The corresponding 2.5.4. In  $C^3$ , let  $\alpha_1 = (1,0,-i)$ ,  $\alpha_2 = (1+i,1-i,1)$  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( $(A \ \beta)$ )=4 and Rank(A)=3, Solution doesn't exist for  $AX = \beta$  and hence  $\beta$  is not in the subspace S.

iii) Checking for  $\gamma$ : To check if a solution exists for  $AX = \gamma$ . 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  $\gamma$  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  $\alpha$  is in the subspace and  $\beta$  and  $\gamma$  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  $\alpha_i$ , i=1,2,3...,n forms basis for subspace of  $\mathbb{F}_2^n$  and a vector  $\alpha \in \mathbb{F}_2^n$ . Then  $\alpha$  is in the subspace of  $\mathbb{F}_2^n$  spanned by  $\alpha_i$ , i=1,2,3...,n if only if  $\alpha$  is in the subspace of  $\mathbb{F}_1^n$  spanned by  $\alpha_i$ , i=1,2,3...,n.

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)

$$\implies C_1 \begin{pmatrix} 1 \\ 0 \\ -i \end{pmatrix} + C_2 \begin{pmatrix} 1+i \\ 1-i \\ 1 \end{pmatrix} + C_3 \begin{pmatrix} i \\ i \\ i \end{pmatrix} = \mathbf{0}$$
(2.5.4.2)

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)

Considering the co-efficient matrix *A*:

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

$$\xrightarrow{R_3 \leftarrow R_3/i}
\begin{pmatrix}
1 & 1+i & i \\
0 & 1-i & i \\
0 & 1 & 1+i
\end{pmatrix}
\xrightarrow{R_3 \leftarrow (1-i)R_3}$$

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

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

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

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  $\alpha_1,\alpha_2$  and  $\alpha_3$  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  $A^{-1}$ 

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. Let  $\mathbb{V}$  be a vector space which is spanned by the rows of matrix

$$\mathbf{A} = \begin{pmatrix} 3 & 21 & 0 & 9 & 0 \\ 1 & 7 & -1 & -2 & -1 \\ 2 & 14 & 0 & 6 & 1 \\ 6 & 42 & -1 & 13 & 0 \end{pmatrix}$$
 (2.5.5.1)

- 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 (2.5.5.1)

$$\begin{pmatrix}
3 & 21 & 0 & 9 & 0 \\
1 & 7 & -1 & -2 & -1 \\
2 & 14 & 0 & 6 & 1 \\
6 & 42 & -1 & 13 & 0
\end{pmatrix}$$

$$\xrightarrow{R_1 \leftarrow \frac{R_1}{3}}
\begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
1 & 7 & -1 & -2 & -1 \\
2 & 14 & 0 & 6 & 1 \\
6 & 42 & -1 & 13 & 0
\end{pmatrix}$$

$$\frac{R_{3} \leftarrow R_{3} - 2R_{1}}{R_{2} \leftarrow R_{2} - R_{1}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & -1 & -5 & -1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & -1 & -5 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{2}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & -1 & -5 & -1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_{2} \leftarrow -R_{2}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 1
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 1 & 5 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

$$\xrightarrow{R_{4} \leftarrow R_{4} - R_{3}} \begin{pmatrix}
1 & 7 & 0 & 3 & 0 \\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

a. For the basis of  $\mathbb{V}$ , we can take the non zero rows of (2.5.5.2)

$$\rho_1 = \begin{pmatrix} 1 \\ 7 \\ 0 \\ 3 \\ 0 \end{pmatrix} \rho_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \\ 5 \\ 0 \end{pmatrix} \rho_3 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$
 (2.5.5.3)

b. Vectors which are elements of  $\mathbb{V}$  are of the form:

$$c_{1}\rho_{1} + c_{2}\rho_{2} + c_{3}\rho_{3}$$

$$= \begin{pmatrix} c_{1} \\ 7c_{1} \\ c_{2} \\ 3c_{1} + 5c_{2} \\ c_{3} \end{pmatrix}$$

$$(2.5.5.4)$$

where  $c_1, c_2, c_3$  are scalars.

c. Expressing (2.5.5.4) in matrix form, if 
$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 is 
$$\begin{pmatrix} x_1 \\ x_3 \\ x_5 \end{pmatrix}$$
 (2.5.5.8)

in V,it must be of the form

$$\begin{pmatrix} 1 & 0 & 0 \\ 7 & 0 & 0 \\ 0 & 1 & 0 \\ 3 & 5 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix}$$
 (2.5.5.5)

The augmented matrix form

$$\begin{pmatrix}
1 & 0 & 0 & x_1 \\
7 & 0 & 0 & x_2 \\
0 & 1 & 0 & x_3 \\
3 & 5 & 0 & x_4 \\
0 & 0 & 1 & x_5
\end{pmatrix}$$
(2.5.5.6)

Converting the above matrix into row reduced echelon form

$$\begin{pmatrix} 1 & 0 & 0 & x_1 \\ 7 & 0 & 0 & x_2 \\ 0 & 1 & 0 & x_3 \\ 3 & 5 & 0 & x_4 \\ 0 & 0 & 1 & x_5 \end{pmatrix} \xrightarrow{R_4 \leftarrow R_4 - 3R_1} \begin{pmatrix} 1 & 0 & 0 & x_1 \\ 0 & 0 & 0 & x_2 - 7x_1 \\ 0 & 1 & 0 & x_3 \\ 0 & 5 & 0 & x_4 - 3x_1 \\ 0 & 0 & 1 & x_5 \end{pmatrix}$$

$$\stackrel{R_4 \leftarrow R_4 - 5R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & x_1 \\
0 & 1 & 0 & x_3 \\
0 & 0 & 0 & x_2 - 7x_1 \\
0 & 0 & 0 & x_4 - 3x_1 - 5x_3 \\
0 & 0 & 1 & x_5
\end{pmatrix} \qquad b)$$

$$\stackrel{R_5 \leftarrow R_3}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 0 & x_1 \\
0 & 1 & 0 & x_3 \\
0 & 0 & 1 & x_5 \\
0 & 0 & 0 & x_4 - 3x_1 - 5x_3 \\
0 & 0 & 0 & x_2 - 7x_1 \\
0 & 0 & 0 & x_2 - 7x_1
\end{pmatrix} \qquad (2.5.5.7)$$

From (2.5.5.7), the coordinates of  $\begin{bmatrix} x_2 \\ x_3 \\ x_4 \end{bmatrix}$  in the

3 Linear Transformations

# 3.1 Linear Transformations

3.1.1. Find weather given functions **T** from  $\mathbb{R}^2$  into  $\mathbb{R}^2$  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:-

$$x_1 = x_2 = 0 (3.1.1.2)$$

$$T\begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{3.1.1.3}$$

(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,

$$\mathbf{x}, \mathbf{y} \in \mathbb{R}^2 \tag{3.1.1.5}$$

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)  
$$= \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} c\mathbf{x} + \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \mathbf{y}$$
 (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,

$$\mathbf{T}(c\mathbf{x} + \mathbf{y}) = c\mathbf{T}(\mathbf{x}) + \mathbf{T}(\mathbf{y}) \qquad (3.1.1.12)$$

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

$$\mathbf{T} \begin{pmatrix} -1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \qquad (3.1.1.17)$$
$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} \neq \begin{pmatrix} -1 \\ 0 \end{pmatrix}. \qquad (3.1.1.18)^{3}$$

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.

(3.1.1.18) 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  $\mathbf{F}$  be a field and let  $\mathbf{V}$  be the space of polynomial functions f from  $\mathbf{F}$  into  $\mathbf{F}$ , given by

$$f(x) = c_0 + c_1 x + \dots + c_n x^n$$

Let **D** be a linear differentiation transforma-

Properties	Zero Transformation	<b>Identity Transformation</b>
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  $\alpha$  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  $D:V\to V$ . 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  $\alpha$  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

$$\mathbf{T} \begin{pmatrix}
1 & 0 & 0 & \cdots & 0 & 0 \\
0 & 1 & 0 & \cdots & 0 & 0 \\
0 & 0 & 1 & \cdots & 0 & 0 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
0 & 0 & 0 & \cdots & 1 & 0 \\
0 & 0 & 0 & \cdots & 0 & 1
\end{pmatrix}$$

$$= \begin{pmatrix}
0 & 0 & 0 & \cdots & 0 & 0 \\
1 & 0 & 0 & \cdots & 0 & 0 \\
0 & \frac{1}{2} & 0 & \cdots & 0 & 0 \\
0 & \frac{1}{2} & 0 & \cdots & 0 & 0 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
0 & 0 & 0 & \cdots & 0 & 0 \\
0 & 0 & 0 & \cdots & \frac{1}{n-1} & 0 \\
0 & 0 & 0 & \cdots & \frac{1}{n-1} & 1
\end{pmatrix}$$

$$(3.1.3.22)$$

Thus the **T** transformation coefficient matrix

is

$$T = \begin{pmatrix} 0 & 0 & 0 & \cdots & 0 & 0 \\ 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & \frac{1}{2} & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 0 & \cdots & \frac{1}{n-1} & 0 \\ 0 & 0 & 0 & \cdots & 0 & \frac{1}{n} \end{pmatrix}$$
(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  $T: V \rightarrow V$  is

$$\mathbf{N} = \left\{ \begin{pmatrix} 0\\0\\\vdots\\0 \end{pmatrix} : k \in \mathbf{R} \right\} \tag{3.1.3.30}$$

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,

$$T: \mathbb{F}^2 \to \mathbb{F}^2 \tag{3.1.4.1}$$

The transformation for  $\in_1$  and  $\in_2$  can be written as,

$$T(\epsilon_1) = \begin{pmatrix} a \\ b \end{pmatrix}$$
 (3.1.4.2)  
$$T(\epsilon_2) = \begin{pmatrix} c \\ d \end{pmatrix}$$
 (3.1.4.3)

$$T(\epsilon_2) = \begin{pmatrix} c \\ d \end{pmatrix} \tag{3.1.4.3}$$

Now,let's assume  $\in_1$  and  $\in_2$  as linearly independent. So the linear transformation T for any vector v in two dimensional space will be,

$$T(\mathbf{v}) = \begin{pmatrix} T(\epsilon_1) & T(\epsilon_2) \end{pmatrix} \mathbf{v}$$
 (3.1.4.4)  
=  $\begin{pmatrix} a & c \\ b & d \end{pmatrix} \mathbf{v}$  (3.1.4.5)

$$= \begin{pmatrix} a & c \\ b & d \end{pmatrix} \mathbf{v} \tag{3.1.4.5}$$

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 (3.1.4.5) we can say that,

$$Range(T) = \text{columnspace of} \begin{pmatrix} a & c \\ b & d \end{pmatrix}$$
 (3.1.4.6)

Now, considering the case when linear transformation will be linearly dependent,

$$Range(T) = \text{columnspace of} \begin{pmatrix} a \\ b \end{pmatrix}$$
 (3.1.4.7)

Now, considering that vectors  $\in_1$  and  $\in_2$  itself are linearly dependent. Let  $\mathbf{v} = \epsilon_1 + \epsilon_2$ 

$$T(\mathbf{v}) = T(\epsilon_1) + T(\epsilon_2)$$
 (3.1.4.8)

$$= T(\in_1) + T(k \in_1)$$
 (3.1.4.9)

$$= (k+1)T(\epsilon_1) \tag{3.1.4.10}$$

$$= (k+1) \binom{a}{b}$$
 (3.1.4.11)

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.

3.1.5. 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

<b>Vectors Independent</b>	<b>Vectors Dependent</b>
$T(\mathbf{v}) = \begin{pmatrix} a & c \\ b & d \end{pmatrix} \mathbf{v}$	$T(\mathbf{v}) = (\mathbf{k} + 1) \begin{pmatrix} a \\ b \end{pmatrix}$
Output:	Output:
On the plane	On the line

**TABLE 3.1.4** 

by

$$\mathbb{T}(x_1, x_2, x_3) =$$

$$(3.1.5.1)$$

$$(x_1 - x_2 + 2x_3, 2x_1 + x_2, -x_1 - 2x_2 + 2x_3)$$

$$(3.1.5.2)$$

- 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

$$\mathbb{T}(x_1, x_2, x_3) = \mathbf{Tx} \tag{3.1.5.3}$$

$$\mathbf{T} = \begin{pmatrix} 1 & -1 & 2 \\ 2 & 1 & 0 \\ -1 & -2 & 2 \end{pmatrix} \tag{3.1.5.4}$$

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \tag{3.1.5.5}$$

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

$$\mathbf{T}(c\mathbf{x} + \mathbf{y}) = \mathbf{T}(c\mathbf{x}) + \mathbf{T}\mathbf{y} \tag{3.1.5.6}$$

$$= c\mathbf{T}\mathbf{x} + \mathbf{T}\mathbf{y} \tag{3.1.5.7}$$

So, T is a linear transformation. Part (b)

$$range(\mathbf{T}) = \{ \mathbf{y} : \mathbf{T}\mathbf{x} = \mathbf{y} \text{ where } \mathbf{x}, \mathbf{y} \in \mathbb{F}^3 \}$$
(3.1.5.8)

$$\mathbf{y} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \quad (3.1.5.9)$$

$$Tx = y (3.1.5.10)$$

$$\implies$$
 **BTx** = **By** (3.1.5.11)

$$\implies \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0\\ \frac{-2}{3} & \frac{1}{3} & 0\\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 & 2\\ 2 & 1 & 0\\ -1 & -2 & 2 \end{pmatrix} \mathbf{x} = (3.1.5.12)$$

$$\begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0\\ \frac{-2}{3} & \frac{1}{3} & 0\\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} a\\b\\c \end{pmatrix} (3.1.5.13)$$

$$\begin{pmatrix} 1 & 0 & \frac{2}{3} \\ 0 & 1 & \frac{-4}{3} \\ 0 & 0 & 0 \end{pmatrix} \mathbf{x} = \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0 \\ \frac{-2}{3} & \frac{1}{3} & 0 \\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} (3.1.5.14)$$

So, rank(T)=2 and comparing the third row element in LHS and RHS of (3.1.5.14)

$$-a + b + c = 0 (3.1.5.15)$$

All vectors  $\begin{pmatrix} a \\ b \\ c \end{pmatrix} \in \mathbb{F}^3$  that satisfy (3.1.5.15) lie in the range of **T** Part (c)

$$nullspace(\mathbf{T}) = \left\{ \mathbf{x} : \mathbf{T}\mathbf{x} = \mathbf{0} \text{ where } \mathbf{x} \in \mathbb{F}^3 \right\}$$
(3.1.5.16)

$$\mathbf{x} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{3.1.5.17}$$

$$\mathbf{T}\mathbf{x} = \mathbf{0}$$
 (3.1.5.18)

$$BTx = 0$$
 (3.1.5.19)

where BT is in reduced row echelon form

$$\begin{pmatrix} \frac{1}{3} & \frac{1}{3} & 0\\ \frac{-2}{3} & \frac{1}{3} & 0\\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 & 2\\ 2 & 1 & 0\\ -1 & -2 & 2 \end{pmatrix} \mathbf{x} = \mathbf{0}$$
 (3.1.5.20)

$$\implies \begin{pmatrix} 1 & 0 & \frac{2}{3} \\ 0 & 1 & \frac{-4}{3} \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad (3.1.5.21)$$

$$\implies a + \frac{2}{3}c = 0$$
 (3.1.5.22)

$$b - \frac{4}{3}c = 0 \qquad (3.1.5.23)$$

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

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix} \in \mathbb{F}^3$$
 that satisfy (3.1.5.22) and (3.1.5.23)

#### Note

 $rank(\mathbf{T})+nullity(\mathbf{T})=2+1=dim(\mathbb{F}^3)$ 

3.1.6. 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,

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.1.6.1}$$

$$T(\mathbf{x}) = \begin{pmatrix} 1 & 1 \\ 0 & 2 \\ -1 & 2 \end{pmatrix} \mathbf{x}$$
 (3.1.6.2)

3.1.7. 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,

$$\mathbf{A_1} \in \mathbf{V} \tag{3.1.7.1}$$

$$\mathbf{A_2} \in \mathbf{V} \tag{3.1.7.2}$$

If c be any scalar of the field  $\mathbb{F}$  we get,

$$c\mathbf{A_1} + \mathbf{A_2} \in \mathbf{V} \tag{3.1.7.3}$$

Applying transformation T on  $(c\mathbf{A_1} + \mathbf{A_2})$  we get,

$$T(c\mathbf{A}_{1} + \mathbf{A}_{2}) = (c\mathbf{A}_{1} + \mathbf{A}_{2})\mathbf{B} - \mathbf{B}(c\mathbf{A}_{1} + \mathbf{A}_{2})$$

$$(3.1.7.4)$$

$$= c\mathbf{A}_{1}\mathbf{B} + \mathbf{A}_{2}\mathbf{B} - c\mathbf{B}\mathbf{A}_{1} - \mathbf{B}\mathbf{A}_{2}$$

$$(3.1.7.5)$$

$$= c(\mathbf{A}_{1}\mathbf{B} - \mathbf{B}\mathbf{A}_{1}) + (\mathbf{A}_{2}\mathbf{B} - \mathbf{B}\mathbf{A}_{2})$$

$$(3.1.7.6)$$

$$= cT(\mathbf{A}_{1}) + T(\mathbf{A}_{2})$$

$$(3.1.7.7)$$

From (3.1.7.7) we conclude that T is a linear transformation from vector space V to V.

3.1.8. 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

$$T: V \to V \tag{3.1.8.1}$$

be a function such that,

$$T(x + iy) = Re(x + iy) = x$$
 (3.1.8.2)

$$\implies T: x + iy \rightarrow x$$
 (3.1.8.3)

where  $x, y \in \mathbb{R}$ .

Let,  $\alpha = a + ib$ ,  $\beta = c + id$ .

$$T (kα + β) = T (ka + ikb + c + id)$$

$$(3.1.8.4)$$

$$= T (ka + c + i(kb + d))$$

$$(3.1.8.5)$$

$$= ka + c$$

$$= kT (α) + T (β)$$

$$(3.1.8.7)$$

Now, let  $z \in V$  such that,

$$z = i$$
 (3.1.8.8)

$$\implies T(z) = T(i) = 0$$
 (3.1.8.9)

We can also write,

$$T(i) = T(i(1)) = iT(1) = i \neq 0$$
 (3.1.8.10)

Thus from (3.1.8.7), T is real linear transformation and from (3.1.8.10), T is not complex linear.

3.1.9. 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

$$\mathbf{AX} = \mathbf{0}_{m \times 1} \tag{3.1.9.1}$$

Let,

$$A = (A_1 \dots A_j \dots A_n)_{1 \times n}$$
 and

(3.1.9.2)

$$\mathbf{X_{j}} = \begin{pmatrix} x_{1} \\ \vdots \\ x_{j} \\ \vdots \\ x_{n} \end{pmatrix}, \text{ where } x_{i} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

$$(3.1.9.3)$$

If  $A_{m \times n}$  is zero transformation, then for any vector  $X_{n \times 1}$ , AX = 0. Consider,

$$\mathbf{AX_i} = \mathbf{0}_{m \times 1} \qquad (3.1.9.4)$$

$$\left(\mathbf{A_1} \dots \mathbf{A_j} \dots \mathbf{A_n}\right) \begin{pmatrix} x_1 \\ \vdots \\ x_j \\ \vdots \\ x_n \end{pmatrix} = \mathbf{0}_{m \times 1} \qquad (3.1.9.5)$$

From (3.1.9.3) and (3.1.9.5)

$$\mathbf{A_i} = \mathbf{0}_{m \times 1} \text{ for } j = 1, 2, ...n$$
 (3.1.9.6)

Substitute (3.1.9.6) in (3.1.9.2)

$$\mathbf{A} = \begin{pmatrix} \mathbf{0}_{m \times 1} & \mathbf{0}_{m \times 1} & \dots & \mathbf{0}_{m \times 1} \end{pmatrix}_{1 \times n} \quad (3.1.9.7)$$

$$\therefore \mathbf{A} = \mathbf{0}_{m \times n} \tag{3.1.9.8}$$

Hence **A** is zero matrix.

Let us assume  $A_{m \times n}$  is a zero matrix

$$\mathbf{A} = \mathbf{0}_{m \times n} \tag{3.1.9.9}$$

Then,

$$T(\mathbf{X}) = \mathbf{AX} \tag{3.1.9.10}$$

$$= 0.X (3.1.9.11)$$

$$= \mathbf{0}_{m \times 1} , \forall \mathbf{X} \in F$$
 (3.1.9.12)

Hence  $T(\mathbf{X}) = \mathbf{A}\mathbf{X}$  is the zero transformation.

From (3.1.9.8) and (3.1.9.12) it is proved that T is the zero transformation if and only if **A** is the zero matrix.

3.1.10. 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  $\mathbf{V}$  and  $\mathbf{W}$  be vector spaces over the field  $\mathbf{F}$  and let  $\mathbf{T}$  be a linear transformation from  $\mathbf{V}$  into  $\mathbf{W}$ . Then,

$$rank(\mathbf{T}) + nullity(\mathbf{T}) = \dim \mathbf{V}$$
 (3.1.10.1)

It is given that range and null space of T are same, let us assume it to be m. Substituting in equation (3.1.10.1)

$$m + m = n \tag{3.1.10.2}$$

$$\implies n = 2m \tag{3.1.10.3}$$

From equation (3.1.10.3), 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  $\mathbf{M}$ 

$$\mathbf{T} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ 0 \end{pmatrix} \tag{3.1.10.4}$$

$$\mathbf{M} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \tag{3.1.10.5}$$

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.

$$\mathbf{T} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{3.1.10.6}$$

$$\mathbf{T} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{3.1.10.7}$$

From (3.1.10.5),

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,

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{3.1.10.8}$$

$$\implies x_1 = t \quad x_2 = 0 \tag{3.1.10.9}$$

$$\implies \mathbf{X} = \begin{pmatrix} t \\ 0 \end{pmatrix} = t \begin{pmatrix} 1 \\ 0 \end{pmatrix} \tag{3.1.10.10}$$

The basis for null space is  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .

$$rank(T) = 1$$
  $nullity(T) = 1$  (3.1.10.11)  
 $dim(V) = 2$  (3.1.10.12)

Thus the range and null space are equal, and n is even.

- 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  $\mathbf{U}(x_1, x_2) = (x_1, 0)$ .
  - a) Let T and U be the linear operators on  $\mathbb{R}^2$  defined by

$$\mathbf{T}(x_1, x_2) = (x_2, x_1) \tag{3.2.1.1}$$

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 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.

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  $x_2$ 

$$\mathbf{x_1} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \mathbf{x_2} = \begin{pmatrix} 4 \\ 3 \end{pmatrix} \tag{3.2.1.8}$$

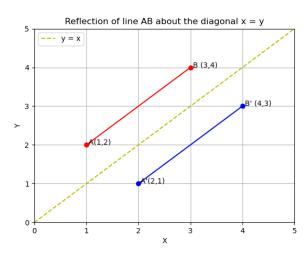


Fig. 3.2.1.1: Reflection of line AB about the x = y

# 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

$$\mathbf{A} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 3 \\ 0 \end{pmatrix} \quad (3.2.1.13)$$

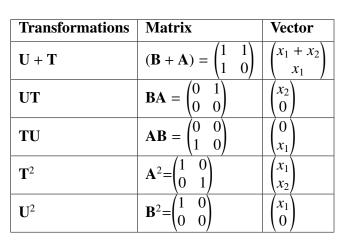


TABLE 3.2.1.1: Summary

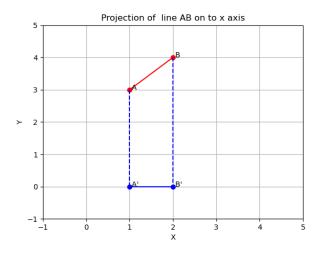


Fig. 3.2.1.2: Projection of AB onto x-axis

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  $\epsilon_i$  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  $rank(\mathbf{T}) = 3$ , **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  $2 \times 2$ 

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

$$\mathbf{T}(\mathbf{A}_{11}) = \mathbf{A}_{11} - 4\mathbf{A}_{21} \tag{3.2.5.7}$$

$$\mathbf{T}(\mathbf{A}_{12}) = \mathbf{B}\mathbf{A}_{12} \tag{3.2.5.8}$$

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

$$\mathbf{T}(\mathbf{A}_{21}) = -\mathbf{A}_{11} + 4\mathbf{A}_{21} \tag{3.2.5.15}$$

$$T(A_{22}) = BA_{22} (3.2.5.16)$$

$$= \begin{pmatrix} 1 & -1 \\ -4 & 4 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.2.5.17}$$

$$= \begin{pmatrix} 0 & -1 \\ 0 & 4 \end{pmatrix} \tag{3.2.5.18}$$

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(T) = 2. Now, we compute

$$T^{2}(A) = T(T(A))$$
 (3.2.5.23)

$$= T(BA)$$
 (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)

$$\begin{pmatrix} 1 & 0 & 1 \\ 1 & 0 & -1 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} = rref(\mathbf{A})$$

$$(3.2.6.9)$$

$$\implies Rank(\mathbf{A}) = 2$$

$$(3.2.6.10)$$

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  $Rank(\mathbf{C}) < dim(\mathbf{C})$ , 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)

... 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)

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 3.2.9.1

Now,

Range(T) = ColumnSpace 
$$\left\{ \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \right\}$$

$$(3.2.8.19)$$

$$= k \begin{pmatrix} 0 \\ 1 \end{pmatrix}; k \in \mathbf{R}$$

$$(3.2.8.20)$$

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 **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, 
$$T: \mathbb{R}^3 \to \mathbb{R}^2$$
 (3.2.9.1)

$$T(\mathbf{X}) = \mathbf{AX} = \mathbf{b} \tag{3.2.9.2}$$

Let, 
$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix}$$
 (3.2.9.3)

Consider, 
$$\mathbf{X} = \begin{pmatrix} 2\\4\\1 \end{pmatrix}$$
 (3.2.9.4)

$$\implies \mathbf{AX} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 4 \\ 1 \end{pmatrix} \quad (3.2.9.5)$$

$$= \begin{pmatrix} 6 \\ 5 \end{pmatrix} \tag{3.2.9.6}$$

Hence T is onto.

Consider, 
$$\mathbf{X} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$
 (3.2.9.7)

$$\Rightarrow \mathbf{AX} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \quad (3.2.9.8)$$
$$= \mathbf{0} \qquad (3.2.9.9)$$

Since  $\exists X \neq 0$  such that AX = 0, T is singular.

.. T is both onto and singular.

b) m > n

Let, 
$$T: \mathbb{R}^3 \to \mathbb{R}^2$$
 (3.2.9.10)

$$T(\mathbf{X}) = \mathbf{AX} = \mathbf{b} \tag{3.2.9.11}$$

Let, 
$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}$$
 (3.2.9.12)

Consider, 
$$\mathbf{X} = \begin{pmatrix} -1\\2 \end{pmatrix}$$
 (3.2.9.13)

$$\implies \mathbf{AX} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} -1 \\ 2 \end{pmatrix} \quad (3.2.9.14)$$

$$= \begin{pmatrix} -1\\2\\-1 \end{pmatrix}$$
 (3.2.9.15)

(3.2.9.16)

.. T is not onto, and is also non-singular.

3.2.10. 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:** 

$$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A} \tag{3.2.10.1}$$

So, **B** is the transformation matrix.

**B** is invertible if

a)  $\mathbb{T}$  is one to one mapping, that is

$$\mathbf{BA} = \mathbf{BA'} \tag{3.2.10.2}$$

$$\implies \mathbf{A} = \mathbf{A}' \tag{3.2.10.3}$$

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 W to V that satisfies

$$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A} \in \mathbf{W} \qquad (3.2.10.4)$$

$$\implies \mathbb{T}_1(\mathbf{B}\mathbf{A}) = \mathbf{A} \in \mathbf{V}$$
 (3.2.10.5)

$$\dim(\mathbf{V}) = mn, \dim(\mathbf{W}) = pn \qquad (3.2.10.6)$$

Since  $\mathbb{T}$  is one-one mapping, the zero vector in  $\mathbf{V}$ ,  $\mathbf{0}_{m \times n}$  is uniquely mapped to

$$\mathbb{T}(\mathbf{0}_{m \times n}) = \mathbf{B}\mathbf{0}_{m \times n} = \mathbf{0}_{p \times n} \tag{3.2.10.7}$$

So, 
$$BA = 0 \iff A = 0$$
 (3.2.10.8)

Let  $\{V_1, V_2, \dots, V_{mn}\}$  be the basis for V

$$c_1 \mathbf{V}_1 + c_2 \mathbf{V}_2 + \ldots + c_{mn} \mathbf{V}_{mn} = \mathbf{0}$$
 (3.2.10.9)

$$\iff c_1, c_2, \dots, c_{mn} \in \mathbb{F} = 0 \quad (3.2.10.10)$$

Any matrix  $A \in V$  can be written as

$$\mathbf{A} = \sum_{i=1}^{mn} \alpha_i \mathbf{V}_i \tag{3.2.10.11}$$

Since  $\mathbb{T}$  is onto, any matrix  $\mathbf{C} \in \mathbf{W}$  can be expressed as

$$\mathbf{C} = \mathbf{B} \left( \sum_{i=1}^{mn} \alpha_i \mathbf{V}_i \right) \tag{3.2.10.12}$$

$$=\sum_{i=1}^{mn}\alpha_i(\mathbf{B}\mathbf{V}_i) \tag{3.2.10.13}$$

So, the set  $S = \{BV_1, BV_2, ..., BV_{mn}\}$  forms basis of W if all matrices in it are linearly

Let <b>A</b> 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	Since $m < n$ , by definition $T$ can be onto	Since $m > n$ , by definition $T$ is not onto.

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	$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A}$
$\mathbb{T}: \mathbf{V} \to \mathbf{W}$	

TABLE 3.2.10.1: Input Parameters

independent.

$$c_{1}(\mathbf{B}\mathbf{V}_{1}) + c_{2}(\mathbf{B}\mathbf{V}_{2}) + \dots + c_{mn}(\mathbf{B}\mathbf{V}_{mn}) = \mathbf{0}$$

$$(3.2.10.14)$$

$$\mathbf{B}(c_{1}\mathbf{V}_{1} + c_{2}\mathbf{V}_{2} + \dots + c_{mn}\mathbf{V}_{mn}) = \mathbf{0}$$

$$(3.2.10.15)$$

$$(3.2.10.8) \implies c_{1}\mathbf{V}_{1} + \dots + c_{mn}\mathbf{V}_{mn} = 0$$

$$(3.2.10.16)$$

$$\iff c_{1}, c_{2}, \dots, c_{mn} = 0 \text{ (from (3.2.10.10))}$$

So, the set S with cardinality mn is basis for W

$$(3.2.10.6) \implies pn = mn \qquad (3.2.10.18)$$

$$p = m (3.2.10.19)$$

(3.2.10.17)

(3.2.10.8),(3.2.10.19) 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  $\{A_1, A_2, ..., A_{mn}\}$  be the basis for **V** 

Any matrix  $A \in V$  can be written as

$$\mathbf{A} = \sum_{i=1}^{mn} \alpha_i \mathbf{A}_i \tag{3.2.10.20}$$

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  $\mathbf{W}$ 

$$c_1(\mathbf{B}\mathbf{A}_1) + c_2(\mathbf{B}\mathbf{A}_2) + \dots + c_{mn}(\mathbf{B}\mathbf{A}_{mn}) = \mathbf{0}$$
(3.2.10.21)

$$\implies \mathbf{B}(c_1\mathbf{A}_1 + c_2\mathbf{A}_2 + \ldots + c_{mn}\mathbf{A}_{mn}) = \mathbf{0}$$
(3.2.10.22)

Since **B** is non-singular,

$$(c_1\mathbf{A}_1 + c_2\mathbf{A}_2 + \dots + c_{mn}\mathbf{A}_{mn}) = \mathbf{0} \quad (3.2.10.23)$$
  
 $\iff c_1, c_2, \dots, c_{mn} = 0 \quad (3.2.10.24)$ 

because  $\{\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_{mn}\}$  are linearly independent

So,M forms basis for W

Any vector  $C \in W$  can be written as

$$\mathbf{C} = \sum_{i=1}^{mn} \beta_i \mathbf{B} \mathbf{A}_i \text{ where } \beta_i \in \mathbb{F} \qquad (3.2.10.25)$$

$$=\mathbf{B}(\sum_{i=1}^{mn}\beta_i\mathbf{A}_i) \qquad (3.2.10.26)$$

$$=$$
 **BA** (from (3.2.10.20)) (3.2.10.27)

So,range(B)=W

Consider the matrix  $A, A' \in V$  such that

$$BA = BA'$$
 (3.2.10.28)

$$\mathbf{B}^{-1}(\mathbf{B}\mathbf{A}) = \mathbf{B}^{-1}(\mathbf{B}\mathbf{A}') \tag{3.2.10.29}$$

$$(\mathbf{B}^{-1}\mathbf{B})\mathbf{A} = (\mathbf{B}^{-1}\mathbf{B})\mathbf{A}'$$
 (3.2.10.30)

$$\implies \mathbf{A} = \mathbf{A}' \tag{3.2.10.31}$$

So,  $\mathbb{T}$  is invertible. Conclusion: From case 1,case 2  $\mathbb{T}$  is invertible if and only if p=m and **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  $\mathbf{A} \in \mathbf{V}$ 

The elementary matrix that performs this is

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \tag{3.2.10.32}$$

Let 
$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 4 & 9 & 2 & 6 \end{pmatrix}$$
 (3.2.10.33)

$$\mathbb{T}(\mathbf{A}) = \mathbf{B}\mathbf{A} \qquad (3.2.10.34)$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 4 & 9 & 2 & 6 \end{pmatrix}$$
 (3.2.10.35)

$$= \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 5 & 12 & 8 & 13 \end{pmatrix}$$
 (3.2.10.36)

$$= \mathbf{C} \in \mathbf{W}$$
 (3.2.10.37)

Let transformation  $\mathbb{T}_1: \mathbf{W} \to \mathbf{V}$  subtracts row2 from row 3 for a matrix  $\mathbf{C} \in \mathbf{W}$  and is

performed by elementary matrix

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

$$(3.2.10.38)$$

$$Let \mathbf{C} = \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 5 & 12 & 8 & 13 \end{pmatrix}$$

$$(3.2.10.39)$$

$$\mathbb{T}_{1}(\mathbf{C}) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 5 & 12 & 8 & 13 \end{pmatrix}$$

$$(3.2.10.40)$$

$$= \begin{pmatrix} 1 & 2 & 2 & 5 \\ 1 & 3 & 6 & 7 \\ 4 & 9 & 2 & 6 \end{pmatrix}$$

$$(3.2.10.41)$$

$$= \mathbf{A}$$

$$(3.2.10.42)$$

$$\implies \mathbb{T}_{1}(\mathbf{C}) = \mathbf{A}$$

$$(3.2.10.43)$$

$$\mathbb{T}_{1}(\mathbb{T}(\mathbf{A})) = \mathbf{A}$$

$$(3.2.10.44)$$
and 
$$\mathbb{T}(\mathbf{A}) = \mathbf{C}$$

$$(3.2.10.45)$$

$$\implies \mathbb{T}(\mathbb{T}_{1}(\mathbf{C})) = \mathbf{C}$$

$$(3.2.10.46)$$

So,  $\mathbb{T}_1$  is the inverse transformation of  $\mathbb{T}$  and

$$\mathbf{UB} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix}$$
 (3.2.10.48)  

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
 (3.2.10.49)  

$$\mathbf{BU} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{pmatrix}$$
 (3.2.10.50)  

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
 (3.2.10.51)  

$$\Rightarrow \mathbf{B}^{-1} = \mathbf{U}$$
 (3.2.10.52)

So,  $\mathbb{T}$  is invertible and  $\mathbf{B}$  is an invertible  $3 \times 3$ matrix.

## 3.3 Isomorphism

3.3.1. 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  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$ ,  $\alpha_4$  be vectors in V such that

$$T(\alpha_1) = \begin{pmatrix} 1\\0\\i \end{pmatrix}, T(\alpha_2) = \begin{pmatrix} -2\\1+i\\0 \end{pmatrix},$$
$$T(\alpha_3) = \begin{pmatrix} -1\\1\\1 \end{pmatrix}, T(\alpha_4) = \begin{pmatrix} \sqrt{2}\\i\\3 \end{pmatrix}$$
(3.3.1.1)

- a) Is  $\alpha_1$  in the subspace spanned by  $\alpha_2$  and  $\alpha_3$ ? **Solution:**  $T: V \rightarrow W$  is an isomorphism if (1) T is one one.
  - (2) T is onto.

$$\begin{pmatrix} 1 & -2 & -1 & \sqrt{2} \\ 0 & 1+i & 1 & i \\ i & 0 & 1 & 3 \end{pmatrix} \xrightarrow{ref} \begin{pmatrix} 1 & 0 & -i & -3i \\ 0 & 2 & 1-i & i+1 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

(3.3.1.2)

(3.3.1.7)

T is one one over  $C^3$  if

$$T(\alpha) = 0 \implies \alpha = 0$$
 (3.3.1.3)

now.

(3.2.10.47)

$$\begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ i & 0 & 3 \end{pmatrix} \alpha = 0$$
 (3.3.1.4)

consider the row reduced matrix

$$\begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ i & 0 & 3 \end{pmatrix} \xrightarrow{R_3 \to R_3 - iR_1} \begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 1+i & i \\ 0 & -2 & \sqrt{2} + 3i \end{pmatrix}$$

$$(3.3.1.5)$$

$$\xrightarrow{R_2 \leftarrow (1-i)R_2} \begin{pmatrix} 1 & -2 & \sqrt{2} \\ 0 & 2 & i+1 \\ 0 & 0 & \sqrt{2} + 4i + 1 \end{pmatrix}$$

$$(3.3.1.6)$$

$$\alpha = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \qquad (3.3.1.7)$$

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  $C^3$ .

$$T(\alpha_1) = c_1 T(\alpha_2) + c_2 T(\alpha_3)$$
 (3.3.1.8)

 $c_1$  and  $c_2$  are scalar.

$$\begin{pmatrix} 1 \\ 0 \\ i \end{pmatrix} = c_1 \begin{pmatrix} -2 \\ 1+i \\ 0 \end{pmatrix} + c_2 \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$
 (3.3.1.9)

$$\begin{pmatrix} 1 \\ 0 \\ i \end{pmatrix} = \begin{pmatrix} -2 & -1 \\ 1+i & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}$$
 (3.3.1.10)

Now we find  $c_i$  by row reducing augmented matrix.

$$\begin{pmatrix}
-2 & -1 & 1 \\
1+i & 1 & 0 \\
0 & 1 & i
\end{pmatrix}
\xrightarrow{R_1 \to -R_1/2}
\xrightarrow{R_2 \to R_3}
\begin{pmatrix}
1 & \frac{1}{2} & -\frac{1}{2} \\
0 & 1 & i \\
1+i & 1 & 0
\end{pmatrix}$$

$$(3.3.1.11)$$

$$\xrightarrow{R_1 \leftarrow R_1 - R_2/2}
\xrightarrow{R_3 \leftarrow R_3 - (1+i)R_1}
\begin{pmatrix}
1 & 0 & \frac{-1-i}{2} \\
0 & 1 & i \\
0 & \frac{1-i}{2} & \frac{1+i}{2}
\end{pmatrix}$$

$$(3.3.1.12)$$

$$\xrightarrow{R_3 \leftarrow R_3 - (1-i)/2R_2}
\begin{pmatrix}
1 & 0 & \frac{-1-i}{2} \\
0 & 1 & i \\
0 & 0 & 0
\end{pmatrix}$$

$$(3.3.1.13)$$

Therefore the coordinate matrix of the vector is

$$\begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = \begin{pmatrix} \frac{-1-i}{2} \\ i \end{pmatrix}$$
 (3.3.1.14)

substituting the  $c_i$  in (3.3.1.8)

$$T(\alpha_1) = -\frac{1+i}{2}T(\alpha_2) + iT(\alpha_3)$$
 (3.3.1.15)

Hence  $\alpha_1$  belongs to the subspace spanned by  $\alpha_2$  and  $\alpha_3$ .

b) Find a basis for the subspace of V spanned by the 4 vectors  $\alpha_i$ .

**Solution:** V is a vector space and V is 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

$$dim(V) = dim(C^3) = 3$$
 (3.3.1.16)

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  $\alpha_1$ ,  $\alpha_2$  and  $\alpha_4$  are the basis of V.

3.3.2. Let  $\mathbb{W}$  be the set of all  $2 \times 2$  complex Hermitian matrices, that is the sset of  $2 \times 2$  complex matrices  $\mathbf{A}$  ssuch that  $\mathbf{A}_{ij} = \overline{\mathbf{A}_{ji}}$  (the bar denoting complex conjugation).  $\mathbb{W}$  is a vector space over the field of real numbers, under the usual operations. Verify that

$$\begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} \rightarrow \begin{pmatrix} t + x & y + iz \\ y - iz & t - x \end{pmatrix}$$
 (3.3.2.1)

is an isomorphism of  $\mathbb{R}^4$  onto  $\mathbb{W}$ .

## **Solution:**

a) Check for linearity: The transformation T

is given by

$$T: \mathbb{R}^4 \to \mathbb{W} \tag{3.3.2.2}$$

$$T \begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix} = \begin{pmatrix} t + x & y + iz \\ y - iz & t - x \end{pmatrix}$$
 (3.3.2.3)

Let  $\mathbf{x} = \begin{bmatrix} x \\ y \\ t \end{bmatrix}$ . Expressing R.H.S of equation

(3.3.2.3) using Kronecker Product,

$$T(\mathbf{x}) = \begin{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 1 \end{pmatrix} \mathbf{x} & \begin{pmatrix} 0 & 1 & i & 0 \end{pmatrix} \mathbf{x} \\ \begin{pmatrix} 0 & 1 & -i & 0 \end{pmatrix} \mathbf{x} & \begin{pmatrix} -1 & 0 & 0 & 1 \end{pmatrix} \mathbf{x} \end{pmatrix}$$

$$= \left(\begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & -i & 0 \end{pmatrix}\right) \mathbf{x} \quad \begin{pmatrix} 0 & 1 & i & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix} \mathbf{x}$$

$$= \left(\begin{pmatrix} 1 & 0 & 0 & 1 & 0 & 1 & i & 0 \\ 0 & 1 & -i & 0 & -1 & 0 & 0 & 1 \end{pmatrix}\right) \begin{pmatrix} x & 0 \\ y & 0 \\ z & 0 \\ t & 0 \\ 0 & x \\ 0 & y \\ 0 & z \\ 0 & t \end{pmatrix}$$

$$= \left(3.3.2.4\right)$$

$$= \left(\begin{pmatrix} 1 & 0 & 0 & 1 & 0 & 1 & i & 0 \\ 0 & 1 & -i & 0 & -1 & 0 & 0 & 1 \end{pmatrix}\right) \begin{pmatrix} x & 0 \\ y & 0 \\ z & 0 \\ t & 0 \\ 0 & x \\ 0 & y \\ 0 & z \\ 0 & t \end{pmatrix}$$

$$= \left(3.3.2.6\right)$$

$$\implies T(\mathbf{x}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{x} & \mathbf{0}_{4\times 1} \\ \mathbf{0}_{4\times 1} & \mathbf{x} \end{pmatrix} (3.3.2.7)$$

Where **A** and **B** are block matrices.

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

$$\mathbf{B} = \begin{pmatrix} 0 & 1 & i & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix}$$
(3.3.2.8) one-one and that implies  $T : \mathbb{R}^4$  isomorphism.

3.3.3. Show that  $\mathbf{F}^{\mathbf{m} \times \mathbf{n}}$  is isomorphic to  $\mathbf{F}^{\mathbf{m} \cdot \mathbf{n}}$ .

Solution: See Tables 3.3.3.1, 3.3.1

$$\mathbf{B} = \begin{pmatrix} 0 & 1 & i & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix} \tag{3.3.2.9}$$

The Kronecker Product of  $I_2$  and x gives the block matrix in equation (3.3.2.7).

$$\mathbf{I}_{2\times2}\otimes\mathbf{x}_{4\times1} = \begin{pmatrix} \mathbf{x} & \mathbf{0} \\ \mathbf{0} & \mathbf{x} \end{pmatrix}_{8\times2} \tag{3.3.2.10}$$

Hence we can write equation (3.3.2.7) as,

$$T(\mathbf{x}) = (\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{x}) \tag{3.3.2.11}$$

Let  $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{R}^4$  and  $\alpha, \beta \in \mathbb{R}$ .

$$T (\alpha \mathbf{x}_1 + \beta \mathbf{x}_2) = (\mathbf{A} \quad \mathbf{B}) (\mathbf{I} \otimes (\alpha \mathbf{x}_1 + \beta \mathbf{x}_2))$$

$$= \alpha (\mathbf{A} \quad \mathbf{B}) (\mathbf{I} \otimes \mathbf{x}_1) + \beta (\mathbf{A} \quad \mathbf{B}) (\mathbf{I} \otimes \mathbf{x}_2)$$

$$= \alpha T \mathbf{x}_1 + \beta T \mathbf{x}_2$$

$$(3.3.2.14)$$

Therefore from equation (3.3.2.14), 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 (3.3.2.11),

$$T\left(\mathbf{x}\right) = \mathbf{0} \tag{3.3.2.15}$$

$$(\mathbf{A} \quad \mathbf{B})(\mathbf{I} \otimes \mathbf{x}) = \mathbf{0} \tag{3.3.2.16}$$

$$\implies \begin{pmatrix} 1 & 0 & 0 & 1 & 0 & 1 & i & 0 \\ 0 & 1 & -i & 0 & -1 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x & 0 \\ y & 0 \\ z & 0 \\ t & 0 \\ 0 & x \\ 0 & y \\ 0 & z \\ 0 & t \end{pmatrix} = \mathbf{0}_{2 \times 2}$$

$$(3.3.2.17)$$

From equation (3.3.2.3),

$$\begin{pmatrix} t+x & y+iz \\ y-iz & t-x \end{pmatrix} = \mathbf{0}_{2\times 2} \quad (3.3.2.18)$$

$$\implies x = 0, y = 0, z = 0, t = 0 \quad (3.3.2.19)$$

$$\implies \mathbf{x} = \mathbf{0} \quad (3.3.2.20)$$

Hence from (3.3.2.15) and (3.3.2.20), T is one-one and that implies  $T: \mathbb{R}^4 \to \mathbb{W}$  is

Solution: See Tables 3.3.3.1, 3.3.3.2 and

 $\mathbb{R}^{2\times 2}$  is isomorphic to  $\mathbb{R}^4$  ie,  $\mathbb{R}^{2\times 2} \cong \mathbb{R}^4$ .

3.3.4. 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  $2 \times 2$  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 <b>V</b> and <b>W</b> 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 3.3.3.1: Definition

Result 1	The space of all $m \times n$ matrices over the field <b>F</b> 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 3.3.3.2: 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  $2 \times 2$  real matrices.

**Solution:** The kronecker product also called as matrix direct product is defined as

$$\mathbf{A} \otimes \mathbf{B} = \begin{pmatrix} a_{11}\mathbf{B} & \cdots & a_{1n}\mathbf{B} \\ \vdots & \ddots & \vdots \\ a_{m1}\mathbf{B} & \cdots & a_{mn}\mathbf{B} \end{pmatrix}$$
(3.3.4.1)

Also,

$$\mathbf{A} \otimes (\mathbf{B} + \mathbf{C}) = \mathbf{A} \otimes \mathbf{B} + \mathbf{A} \otimes \mathbf{C}$$
 (3.3.4.2)

$$\mathbf{A} \otimes (k\mathbf{B}) = k(\mathbf{A} \otimes \mathbf{B}) \tag{3.3.4.3}$$

Given,

$$\mathbf{T}: \mathbf{C} \to \mathbf{R}^{2\times 2}$$

$$\mathbf{T}(x+iy) = \begin{pmatrix} x+7y & 5y \\ -10y & x-7y \end{pmatrix}$$
 (3.3.4.4)

Let,

$$z = x + iy;$$
  $w = a + ib;$   $z, w \in \mathbb{C}$ 

Also the RHS of (3.3.4.4) can be expressed as,

$$\mathbf{T}(\mathbf{z}) = \begin{pmatrix} 1 & 7 \\ 0 & -10 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} & \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \end{pmatrix}$$
$$= \begin{pmatrix} 1 & 7 & 0 & 5 \\ 0 & -10 & 1 & -7 \end{pmatrix} \begin{pmatrix} x & 0 \\ y & 0 \\ 0 & x \\ 0 & y \end{pmatrix}$$
$$= \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \begin{pmatrix} \mathbf{x} & 0 \\ 0 & \mathbf{x} \end{pmatrix} \tag{3.3.4.5}$$

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  ${\bf x}$ 

$$\mathbf{I} \otimes \mathbf{x} = \begin{pmatrix} \mathbf{x} & 0 \\ 0 & \mathbf{x} \end{pmatrix} \tag{3.3.4.6}$$

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 $(a,b) \to (a-1)n + b$
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 3.3.3.3: Proof

Where I is an identity matrix. (3.3.4.5) can be rewritten as,

$$\mathbf{T}(\mathbf{z}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes \mathbf{x}) \tag{3.3.4.7}$$

Consider,

$$\mathbf{T}(\alpha \mathbf{z} + \mathbf{w}) = (\mathbf{A} \ \mathbf{B})(\mathbf{I} \otimes (\alpha \mathbf{z} + \mathbf{w}))$$

Using properties (3.3.4.2), (3.3.4.3), the above equation can be expressed as,

$$\mathbf{T}(\alpha \mathbf{z} + \mathbf{w}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes (\alpha \mathbf{z})) + \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes \mathbf{x})$$
$$= \alpha \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes \mathbf{z}) + \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} (\mathbf{I} \otimes \mathbf{w})$$
$$= \alpha \mathbf{T}(\mathbf{z}) + \mathbf{T}(\mathbf{w}) \qquad (3.3.4.8)$$

From (3.3.4.8), it can be proved that **T** is a linear operator.

b) How would you describe the range of  $\mathbb{T}$ ? Solution:

$$\mathbb{T}: \mathbf{V} \to \mathbb{R}^{2 \times 2} \tag{3.3.4.9}$$

where  $\mathbb{R}^{2\times 2}$ , is the space of all  $2\times 2$  real matrices

$$\mathbb{T}(z) = \begin{pmatrix} x + 7y & 5y \\ -10y & x - 7y \end{pmatrix} \quad (3.3.4.10)$$

$$\mathbf{x} = \begin{pmatrix} x \\ y \end{pmatrix} \quad (3.3.4.11)$$

$$\mathbb{T}(\mathbf{x}) = \begin{pmatrix} (1 & 7)\mathbf{x} & (0 & 5)\mathbf{x} \\ (0 & -10)\mathbf{x} & (1 & -7)\mathbf{x} \end{pmatrix} \quad (3.3.4.12)$$

$$= \begin{pmatrix} (1 & 7 \\ 0 & -10 \end{pmatrix} \mathbf{x} \quad \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \mathbf{x} \quad (3.3.4.13)$$

$$\text{Let } \mathbf{A} = \begin{pmatrix} 1 & 7 \\ 0 & -10 \end{pmatrix} \quad (3.3.4.14)$$

$$\mathbf{B} = \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix} \quad (3.3.4.15) \quad 3.3.4.16$$

$$\implies \mathbb{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A} \mathbf{B} \mathbf{B} \begin{pmatrix} \mathbf{x} & \mathbf{0}_{2\times 1} \\ \mathbf{0}_{2\times 1} & \mathbf{x} \end{pmatrix} \quad (3.3.4.17)$$

The kronecker product of  $\mathbf{I}$ ,  $\mathbf{x}$  gives the block matrix

$$\mathbf{I}_{2\times2} \otimes \mathbf{x}_{2\times1} = \begin{pmatrix} \mathbf{x} & \mathbf{0} \\ \mathbf{0} & \mathbf{x} \end{pmatrix}_{4\times2}$$

$$(3.3.4.18)$$

$$(3.3.4.17) \implies \mathbb{T}(\mathbf{x}) = \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \mathbf{x}$$

$$(3.3.4.19)$$

$$= \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{bmatrix} \begin{pmatrix} x \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ y \end{pmatrix} \end{bmatrix}$$

$$(3.3.4.20)$$

$$= x \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} + y \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$(3.3.4.21)$$

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

$$(3.3.4.22)$$

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

Kronecker product in the first term of (3.3.4.21) picks out 1st columns of **A**, **B** and in the second term picks out 2nd columns of

(3.3.4.23)

A, B so basis for range( $\mathbb{T}$ ) is

$$\left\{ \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} \mathbf{A} & \mathbf{B} \end{pmatrix} \mathbf{I} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\} = (3.3.4.24)$$

$$\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 7 & 5 \\ -10 & -7 \end{pmatrix} \right\}$$

$$(3.3.4.25)$$

$$\operatorname{range}(\mathbb{T}) = (3.3.4.26)$$

$$\operatorname{span of} \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 7 & 5 \\ -10 & -7 \end{pmatrix} \right\}$$

$$(3.3.4.27)$$

 $\mathbf{B} = \begin{pmatrix} 0 & 5 \\ 1 & -7 \end{pmatrix}$  (3.3.4.15) 3.3.5. 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

$$U = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \tag{3.3.5.1}$$

here U is an isomorphism from  $\mathbb{R}^{2\times 2}$  to  $\mathbb{R}^{2\times 2}$  since inverse of U exists and

$$U^{-1} = \begin{pmatrix} -2 & -\frac{3}{2} \\ -1 & -\frac{1}{2} \end{pmatrix}$$
 (3.3.5.2)

Consider

$$T = \begin{pmatrix} -1 & 2\\ 3 & 1 \end{pmatrix} \in \mathbb{R}^{2 \times 2} \tag{3.3.5.3}$$

Now

$$UTU^{-1} = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} -1 & 2 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} -2 & -1 \\ -\frac{3}{2} & -\frac{1}{2} \end{pmatrix}$$
 (3.3.5.4)  
= 
$$\begin{pmatrix} -16 & -7 \\ -33 & -14 \end{pmatrix} \in \mathbb{R}^{2 \times 2}$$
 (3.3.5.5)

Also inverse exists for T

$$S = T^{-1} = \begin{pmatrix} -\frac{1}{2} & \frac{2}{7} \\ \frac{3}{7} & \frac{1}{7} \end{pmatrix}$$
 (3.3.5.6)

Since T inverse exists  $\mathcal{T}(T) = UTU^{-1}$  is an isomorphism from  $\mathbb{R}^{2\times 2}$  onto  $\mathbb{R}^{2\times 2}$ .

3.3.6. 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.

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 $\mathcal{T}(U^{-1}SU)$ ,
	$\int \mathcal{T}(U^{-1}SU) = U(U^{-1}SU)U^{-1} = S$
	$\implies TS = I \text{ and } ST = I$
	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 3.3.5.1: Proof

**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.

$$TU = I$$
 (3.3.6.1)

Where, *I* is an identity transformation. Now we know that linear transformations are functions. Hence,

$$TU = I$$
 is a function (3.3.6.2)

$$\Longrightarrow I: \mathbb{V} \to \mathbb{V} \tag{3.3.6.3}$$

Such that T(V) = V. Defining  $TU : \mathbb{V} \to \mathbb{V}$  to be a linear operator, we have,

$$T[U(V_i)] = V_i \qquad [V_i \in \mathbb{V}] \qquad (3.3.6.4)$$

Now we show in the below Table that T is one-one and onto as follows,

Hence we get from Table 3.3.6.1 that, T is invertible. Hence we get the following,

$$TT^{-1} = I$$
 (3.3.6.5)

Where  $T^{-1}$  is an inverse function of linear operator T. Hence,

$$TT^{-1} = I = TU$$
 (3.3.6.6)

$$\implies T^{-1}(TT^{-1}) = T^{-1}(TU)$$
 (3.3.6.7)

$$\implies T^{-1}(I) = IU \tag{3.3.6.8}$$

$$\implies T^{-1} = U \tag{3.3.6.9}$$

Hence from (3.3.6.9) 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 functions in one variable x over  $\mathbb{R}$  as follows,

$$D(c_0 + c_1 x + \dots + c_n x^n) = c_1 + c_2' x + \dots + c_n' x^{n-1}$$
(3.3.6.10)

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,

$$span(\mathbb{V}) = \{p_1, p_2, \dots, p_k\}$$
 (3.3.6.11)

Also let m be the maximum of the degree of these k polynomials in (3.3.6.11). Now let an

linear transformation	Let $V$ and $W$ be vector spaces over field $F$ . A <b>linear transformation</b> $V$ into $W$ is a function $T$ from $V$ into $W$ such that $T(c\alpha + \beta) = c(T\alpha) + T\beta$ for all $\alpha$ and $\beta$ 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 <b>isormorphism of</b> $V$ <b>onto</b> $W$
one-one	A linear transformation $T: \mathbb{R}^n \to \mathbb{R}^m$ is said to be <b>one-one</b> if for every $\mathbf{b} \in \mathbb{R}^m$ , $\mathbf{A}\mathbf{X} = \mathbf{b}$ has atmost 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 <b>one-one</b>
invertible	A linear transformation $T: V \to W$ is <b>invertible</b> 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 <b>invertible</b> if and only if $T$ is $one - one$ and $onto$

TABLE 3.3.5.2: 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 3.3.6.1: Proof of Invertibility of transformation

element of the vector space V be,

$$cx^{m+1} \in \mathbb{V} \tag{3.3.6.12}$$

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,

$$cx^{m+1} \neq \sum_{i=1}^{k} \alpha_i p_i \quad [\alpha_i \in \mathbb{F} \ \forall i] \qquad (3.3.6.13)$$

Hence,  $cx^{m+1}$  is not in the span of

 $p_1, p_2, \dots, p_k$ . 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,

$$c_1 + x^m \in \mathbb{V} \tag{3.3.6.14}$$

$$c_2 + x^m \in \mathbb{V} \tag{3.3.6.15}$$

From definition (3.3.6.10) of operator D we have,

$$D(c_1 + x^m) = mx^{m-1} (3.3.6.16)$$

$$D(c_2 + x^m) = mx^{m-1} (3.3.6.17)$$

From (3.3.6.16) and (3.3.6.17),

$$c_1 + x^m \neq c_2 + x^m \tag{3.3.6.18}$$

$$D(c_1 + x^m) = D(c_2 + x^m)$$
 (3.3.6.19)

Hence from (3.3.6.19) we see that D is not One-One operator.

And,  $U: \mathbb{V} \to \mathbb{V}$  is another linear operator

such that,

$$U(c_0 + c_1 x + \dots + c_n x^n) = c_0 x + c_1 \frac{x^2}{2} + \dots + c_n \frac{x^{n+1}}{n+1}$$
(3.3.6.20)

Now,  $DU : \mathbb{V} \to \mathbb{V}$  is a linear operator such that,

$$DU(c_0 + c_1 x + \dots + c_n x^n)$$
 (3.3.6.21)  
=  $D[U(c_0 x + c_1 \frac{x^2}{2} + \dots + c_n \frac{x^{n+1}}{n+1})]$  (3.3.6.22)  
 $x^2$   $x^{n+1}$ 

$$= D[c_0x + c_1\frac{x^2}{2} + \dots + c_n\frac{x^{n+1}}{n+1}] \quad (3.3.6.23)$$

$$= c_0 + c_1 \frac{2x}{2} + \dots + c_n \frac{(n+1)x^n}{n+1}$$
 (3.3.6.24)

$$= c_0 + c_1 x + \dots + c_n x^n \tag{3.3.6.25}$$

Hence, from (3.3.6.25),

$$DU = I$$
 (3.3.6.26)

Again  $UD : \mathbb{V} \to \mathbb{V}$  is a linear operator such that,

$$UD(c_0 + c_1 x + \dots + c_n x^n)$$
 (3.3.6.27)  

$$= U[D(c_0 x + c_1 \frac{x^2}{2} + \dots + c_n \frac{x^{n+1}}{n+1})]$$
 (3.3.6.28)  

$$= U[c_1 + c_2' x + \dots + c_n' x^{n-1}]$$
 (3.3.6.29)  

$$= c_1 x + c_2 \frac{x^2}{2} + \dots + c_n \frac{x^n}{n}$$
 (3.3.6.30)

Hence, from (3.3.6.30),

$$UD \neq I$$
 (3.3.6.31)

Hence, from (3.3.6.26) and (3.3.6.31), *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  $\beta$  be the standard ordered basis for  $\mathbb{C}^2$  and  $\beta' = \{\alpha_1, \alpha_2\}$  be the ordered basis defined by  $\alpha_1 = (1, i), \alpha_2 = (-i, 2)$ .
  - a) 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

$$\mathbf{e_1} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \mathbf{e_2} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \tag{3.4.1.1}$$

$$\beta = \begin{pmatrix} \mathbf{e_1} & \mathbf{e_2} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.4.1.2}$$

$$\beta' = \begin{pmatrix} \alpha_2 & \alpha_1 \end{pmatrix} = \begin{pmatrix} -i & 1\\ 2 & i \end{pmatrix} \tag{3.4.1.3}$$

T in the ordered basis  $\beta$  is:

$$[\mathbf{T}]_{\beta} = \begin{pmatrix} -i & 1\\ 2 & i \end{pmatrix} \tag{3.4.1.4}$$

T is defined by

$$T(\mathbf{x}) = \mathbf{A}\mathbf{x} \tag{3.4.1.5}$$

$$T(\mathbf{x}) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \mathbf{x} \tag{3.4.1.6}$$

$$T(\alpha_2) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} -i \\ 2 \end{pmatrix} = \begin{pmatrix} -i \\ 0 \end{pmatrix}$$
 (3.4.1.7)

$$T(\alpha_1) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ i \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$
 (3.4.1.8)

$$[\alpha]_{\beta} = \begin{pmatrix} -i & 1\\ 0 & 0 \end{pmatrix} \tag{3.4.1.9}$$

The matrix of T in the ordered basis  $\{\alpha_2, \alpha_1\}$  is given as:

$$[\mathbf{T}_{\alpha}]_{\beta} = [\mathbf{T}]_{\beta}[\alpha]_{\beta} = \begin{pmatrix} -i & 1 \\ 2 & i \end{pmatrix} \begin{pmatrix} -i & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} -1 & -i \\ -2i & 2 \end{pmatrix}$$
(3.4.1.10)

3.4.2. 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

$$[T]_B = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \tag{3.4.2.1}$$

Prove that

$$T^{2} - (a+d)T + (ad - bc)I = 0 (3.4.2.2)$$

**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:

$$\begin{vmatrix} A - \lambda I \end{vmatrix} = 0 \quad (3.4.2.3)$$

$$\implies \begin{vmatrix} a - \lambda & b \\ c & d - \lambda \end{vmatrix} = 0 \quad (3.4.2.4)$$

$$\implies \lambda^2 - (a+d)\lambda + (ad-bc) = 0 \quad (3.4.2.5)$$

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.

$$\lambda^{2} - (a+d)\lambda + (ad-bc) = 0$$

$$(3.4.2.6)$$

$$\implies A^{2} - (a+d)A + (ad-bc)I = 0$$

$$(3.4.2.7)$$

We can also write the equation 3.4.2.7 as:

$$[T]_B^2 - (a+d)[T]_B + (ad-bc)I = 0$$
 (3.4.2.8)  
or,  $T^2 - (a+d)T + (ad-bc)I = 0$  (3.4.2.9)

3.4.3. Let T be a linear operator on  $\mathbb{R}^3$ , the matrix of which in the standard ordered basis is,

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ -1 & 3 & 4 \end{pmatrix} \tag{3.4.3.1}$$

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 columnspace 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,

$$\begin{pmatrix}
1 & 2 & 1 \\
0 & 1 & 1 \\
-1 & 3 & 4
\end{pmatrix}
\xrightarrow{R_3 = R_3 + R_1}
\begin{pmatrix}
1 & 0 & -1 \\
0 & 1 & 1 \\
0 & 5 & 5
\end{pmatrix}$$
(3.4.3.2)
$$\mathbf{T} = \begin{pmatrix}
3 & 0 & 1 \\
-2 & 1 & 0 \\
-1 & 2 & 4
\end{pmatrix}$$
(3.4.5. The linear operator  $\mathbf{T}$  on  $\mathbf{R}^2$  defined by
$$\mathbf{T} \begin{pmatrix}
x_1 \\
x_2 \\
x_3 \\
x_4 \\
x_5 \\
x_6 \\
x_6$$

From (3.4.3.3) the basis of the range of linear operator T are as follows,

$$\mathbf{a_1} = \begin{pmatrix} 1 & 0 & 0 \end{pmatrix} \tag{3.4.3.4}$$

$$\mathbf{a_2} = \begin{pmatrix} 0 & 1 & 0 \end{pmatrix} \tag{3.4.3.5}$$

Again, the basis for null-space of linear oper-

ator T or  $N(\mathbf{A})$  is a solution of the equation Ax = 0. From (3.4.3.3) we have,

$$\mathbf{A}\mathbf{x} = 0$$
 (3.4.3.6)

$$\implies \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = 0 \tag{3.4.3.7}$$

Setting the value of the free variable  $x_3 = 1$  we get the solution,

$$\mathbf{x} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \tag{3.4.3.8}$$

Hence, the basis of the null-space of the linear operator T is given by,

$$\mathbf{b} = \begin{pmatrix} 1 & -1 & 1 \end{pmatrix} \tag{3.4.3.9}$$

3.4.4. Let  $\mathbb{T}$  be the linear operator on  $\mathbb{R}^3$  defined by

$$\mathbb{T}(x_1, x_2, x_3) =$$
(3.4.4.1)

$$(3x_1 + x_3, -2x_1 + x_2, -x_1 + 2x_2 + 4x_3)$$
 (3.4.4.2)

What is the matrix of  $\mathbb{T}$  in the standard ordered basis of  $\mathbb{R}^3$ ?

**Solution:** 

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \tag{3.4.4.3}$$

$$\mathbb{T}(\mathbf{x}) = \mathbf{T}\mathbf{x} \tag{3.4.4.4}$$

The matrix of  $\mathbb{T}$  in the standard ordered basis from (3.4.4.2) is

$$\mathbf{T} = \begin{pmatrix} 3 & 0 & 1 \\ -2 & 1 & 0 \\ -1 & 2 & 4 \end{pmatrix} \tag{3.4.4.5}$$

$$\mathbf{T} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ 0 \end{pmatrix} \tag{3.4.5.1}$$

is represented by the matrix

$$\mathbf{A} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.4.5.2}$$

(3.4.3.5) 3.4.6. Prove that if **S** is a linear operator on  $\mathbb{R}^2$  such that  $S^2 = S$ , then S = 0, or S = I, or there is an ordered basis **B** for  $\mathbb{R}^2$  such that  $[S]_B = A$ . 3.4.7. Let V be an n-dimensional vector space over **Solution:** If a linear operator **S** is defined on  $\mathbf{R}^2$  such that  $\mathbf{S}^2 = \mathbf{S}$ , then

$$S^2 - S = 0 (3.4.6.1)$$

$$S(S - I) = 0 (3.4.6.2)$$

$$\implies$$
 **S** = **0**, **S** = **I** (3.4.6.3)

The transformation of a vector  $\mathbf{x} \in \mathbf{R}^2$  can be represented as

$$Sx = v$$
 (3.4.6.4)

$$\implies$$
 S(Sx) = Sy (3.4.6.5)

$$\implies \mathbf{S}^2 \mathbf{x} = \mathbf{S} \mathbf{y} \tag{3.4.6.6}$$

$$\implies$$
 **Sx** = **Sy** (3.4.6.7)

$$\implies$$
  $\mathbf{x} = \mathbf{y}$  (3.4.6.8)

Therefore the transformation of a vector  $\mathbf{x} \in \mathbf{R}^2$ can be given as

$$\mathbf{S}\mathbf{x} = \mathbf{x} \ \forall \ \mathbf{x} \in \mathbf{R}^2 \tag{3.4.6.9}$$

Consider the ordered basis set

$$B = \{\epsilon_1, \epsilon_2\} \in \mathbf{R}^2 \tag{3.4.6.10}$$

and if

$$[S]_B = A$$
 (3.4.6.11)

$$\implies [\mathbf{S}]_{\mathbf{B}} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \tag{3.4.6.12}$$

Thus we can re-write the column vectors of  $[S]_B$  using (3.4.6.9) as

$$\mathbf{S}\begin{pmatrix}1\\0\end{pmatrix} = \begin{pmatrix}1\\0\end{pmatrix} = 1\begin{pmatrix}1\\0\end{pmatrix} + 0\begin{pmatrix}0\\1\end{pmatrix} \tag{3.4.6.13}$$

$$\mathbf{S}\begin{pmatrix}0\\0\end{pmatrix} = \begin{pmatrix}0\\0\end{pmatrix} = 0\begin{pmatrix}1\\0\end{pmatrix} + 0\begin{pmatrix}0\\1\end{pmatrix} \tag{3.4.6.14}$$

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 [S]<sub>B</sub>. 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  $\binom{1}{0}$ . One of 3.4.8. Let V and W be finite-dimensional vector the ordered basis set can be

$$B = \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\} \tag{3.4.6.15}$$

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

$$T\alpha_j = \alpha_{j+1}, j = 1, \dots, n-1$$
 (3.4.7.1)

$$T\alpha_n = 0. ag{3.4.7.2}$$

a) What is the matrix A of T in the ordered basis  $\mathcal{B}$ ?

**Solution:** Given that,

$$T: V \to V \tag{3.4.7.3}$$

$$T: V \to V$$
 (3.4.7.3)  
 $[T(\alpha)]_{\mathcal{B}} = A[\alpha]_{\mathcal{B}}$  (3.4.7.4)

$$T\alpha_i = \alpha_{i+1} \tag{3.4.7.5}$$

$$T\alpha_n = 0 \tag{3.4.7.6}$$

where j = 1, ..., n - 1. The matrix A of T in the ordered basis  $\mathcal{B}$  is given by,

$$\implies A = ([T\alpha_1]_{\mathcal{B}} \cdots [T\alpha_n]_{\mathcal{B}}) \quad (3.4.7.7)$$

For  $j = 1, \dots, n-1$  we have,

$$T\alpha_i = \alpha_{i+1} \tag{3.4.7.8}$$

we can write,

$$T\alpha_j = 0\alpha_1 + \ldots + 0\alpha_j + 1\alpha_{j+1} + \ldots + 0\alpha_n$$
(3.4.7.9)

$$\implies [T\alpha_j]_{\mathcal{B}} = (0, \dots, 0, 1, 0, \dots, 0)^T$$
(3.4.7.10)

where 1 is in (j + 1)th position. Now,

$$T\alpha_n = 0 \tag{3.4.7.11}$$

$$\implies [T\alpha_n]_{\mathcal{B}} = 0 \tag{3.4.7.12}$$

Thus from (3.4.7.7), (3.4.7.10) (3.4.7.12) we get matrix A of T in the ordered basis  $\mathcal{B}$  as,

$$A = \begin{pmatrix} 0 & 0 & 0 & 0 & \cdots & 0 & 0 \\ 1 & 0 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & 0 & 1 & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & 1 & 0 \end{pmatrix}$$
 (3.4.7.13)

spaces over the field F and let T be a linear

transformation from V into W. If

$$\mathcal{B} = \{\alpha_1, \dots, \alpha_n\} \text{ and } \mathcal{B}' = \{\beta_1, \dots, \beta_m\}$$
(3.4.8.1)

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

$$T = \sum_{p=1}^{m} \sum_{q=1}^{n} A_{pq} E^{p,q}$$
 (3.4.8.2)

for certain scalars  $A_{pq}$  (the coordinates of T in this basis for L(V, 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,

$$T = \sum_{p=1}^{m} \sum_{q=1}^{n} A_{pq} E^{p,q}$$
 (3.4.8.3) 3.5

where

$$E^{p,q}(\alpha_i) = \begin{cases} \beta_p & p = i \\ 0 & \text{otherwise} \end{cases}$$

$$= \delta_{i\alpha}\beta_p$$
(3.4.8.4)

where  $\mathcal{B} = \{\alpha_1, \dots, \alpha_n\}$  is basis of V and  $\mathcal{B}' = \{\beta_1, \dots, \beta_n\}$  is basis of W.

Consider a vector  $\mathbf{x} = \{x_1, x_2, \dots, x_n\} \in V$ ,

$$\mathbf{x} = \sum_{q=1}^{n} x_q \alpha_q \tag{3.4.8.6}$$

$$\therefore E^{p,q}(\mathbf{x}) = \sum_{q=1}^{n} x_q E^{p,q}(\alpha_q)$$
 (3.4.8.7)

$$= x_q \delta_{iq} \beta_p \tag{3.4.8.8}$$

Consider  $T(\mathbf{x})$ , from (3.4.8.3)

$$T(\mathbf{x}) = \sum_{p=1}^{m} \sum_{q=1}^{n} A_{pq} E^{p,q}(\mathbf{x})$$
 (3.4.8.9)

Substitute (3.4.8.8) in (3.4.8.9)

$$T(\mathbf{x}) = \sum_{p=1}^{m} \sum_{q=1}^{n} A_{pq} x_p \delta_{iq} \beta_q$$
 (3.4.8.10)

From (3.4.8.5),  $\delta_{iq}\beta_q$  is the transformation of basis of V to W. Hence  $T:V\to W$  is

$$T = \begin{pmatrix} \sum_{p=1}^{n} A_{p1} x_{p} \\ \sum_{p=1}^{n} A_{p2} x_{p} \\ \vdots \\ \sum_{p=1}^{n} A_{pm} x_{p} \end{pmatrix}$$
(3.4.8.11)

$$T = \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & A_{22} & \dots & A_{2n} \\ \vdots & \vdots & \dots & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$
(3.4.8.12)

 $\therefore$  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

(3.4.8.3) 3.5.1. Let  $\mathbb{V}$  be the vector space of all  $2 \times 2$  matrices over the field of real numbers, and let

$$\mathbf{B} = \begin{pmatrix} 2 & -2 \\ -1 & 1 \end{pmatrix} \tag{3.5.1.1}$$

Let  $\mathbb{W}$  be the subspace of  $\mathbb{V}$  consisting of all  $\mathbf{A}$  such that  $\mathbf{A}\mathbf{B} = 0$ . 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  $\mathbf{I}$  is the  $2 \times 2$  identity matrix and

$$\mathbf{C} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \tag{3.5.1.2}$$

Find  $f(\mathbf{B})$  Solution: