

# Geometry through Linear Algebra



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Abstract—This book provides a vector approach to analytical geometry. The content and exercises are based on William Dresden's book on solid geometry.

### 1 Planes and Lines

## 1.1 Distance from a plane to a point

### 1.1.1. Solve the following

a) Find the foot of perpendicular from the point

$$\mathbf{A} = \begin{pmatrix} -3 \\ 2 \\ 1 \end{pmatrix} \text{ on the plane } \begin{pmatrix} 3 & 2 & -6 \end{pmatrix} \mathbf{x} = 2.$$

**Solution:** Consider orthogonal vectors  $m_1$  and  $m_2$  to the given normal vector n. Let,

$$\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}, \text{ then }$$

$$\mathbf{m}^{\mathbf{T}}\mathbf{n} = 0 \qquad (1.1.1.1)$$

$$\implies \begin{pmatrix} a & b & c \end{pmatrix} \begin{pmatrix} 3 \\ 2 \\ -6 \end{pmatrix} = 0 \qquad (1.1.1.2)$$

$$\implies 3a + 2b - 6c = 0$$
 (1.1.1.3)

Let a=1 and b=0 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ \frac{1}{2} \end{pmatrix} \tag{1.1.1.4}$$

Let a=0 and b=1 we get,

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ \frac{1}{2} \end{pmatrix} \tag{1.1.1.5}$$

Solving the equation,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.1.6}$$

Substituting (1.1.1.4) and (1.1.1.5) in (1.1.1.6),

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{1}{2} & \frac{1}{3} \end{pmatrix} \mathbf{x} = \begin{pmatrix} -3 \\ 2 \\ 1 \end{pmatrix} \tag{1.1.1.7}$$

Solving (1.1.1.7) using Singular Value De-

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composition on M as follows,

$$\mathbf{M} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T \tag{1.1.1.8}$$

Where the columns of V are the eigen vectors of  $M^TM$ , the columns of U are the eigen vectors of  $MM^T$  and S is diagonal matrix of singular value of eigenvalues of  $M^TM$ . We have,

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} \frac{5}{4} & \frac{1}{6} \\ \frac{1}{6} & \frac{10}{9} \end{pmatrix}$$
 (1.1.1.9)

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{1}{3} \\ \frac{1}{2} & \frac{1}{3} & \frac{13}{36} \end{pmatrix}$$
 (1.1.1.10)

Substituting (1.1.1.8) in (1.1.1.6),

$$\mathbf{U}\mathbf{\Sigma}\mathbf{V}^{T}\mathbf{x} = \mathbf{b} \tag{1.1.1.11}$$

$$\implies \mathbf{x} = \mathbf{V} \mathbf{\Sigma}^{-1} \mathbf{U}^{\mathrm{T}} \mathbf{b} \tag{1.1.1.12}$$

Where  $\Sigma^{-1}$  is Moore-Penrose Pseudo-Inverse of  $\Sigma$  and is obtained by inversing only non-zero elements in  $\Sigma$ 

Calculating eigen values of  $\mathbf{M}\mathbf{M}^T$ ,

$$\begin{vmatrix} \mathbf{M}\mathbf{M}^{T} - \lambda \mathbf{I} | = 0 \quad (1.1.1.13) \\ \Rightarrow \begin{vmatrix} 1 - \lambda & 0 & \frac{1}{2} \\ 0 & 1 - \lambda & \frac{1}{3} \\ \frac{1}{2} & \frac{1}{3} & \frac{13}{36} - \lambda \end{vmatrix} = 0 \quad (1.1.1.14) \\ \Rightarrow \lambda^{3} - \frac{85}{36}\lambda^{2} + \frac{49}{36}\lambda = 0 \quad (1.1.1.15)$$

From the characteristic equation (1.1.1.15), the eigen values of  $\mathbf{MM}^T$  are,

$$\lambda_1 = \frac{49}{36}$$
  $\lambda_2 = 1$   $\lambda_3 = 0$  (1.1.1.16)

The eigen vectors of  $\mathbf{M}\mathbf{M}^T$  are,

$$\mathbf{u_1} = \begin{pmatrix} \frac{18}{13} \\ \frac{12}{13} \\ 1 \end{pmatrix} \quad \mathbf{u_2} = \begin{pmatrix} \frac{-2}{3} \\ 1 \\ 0 \end{pmatrix} \quad \mathbf{u_3} = \begin{pmatrix} \frac{-1}{2} \\ \frac{-1}{3} \\ 1 \end{pmatrix}$$
(1.1.1.17)

Normalizing the eigen vectors in equation (1.1.1.17)

$$\mathbf{u_1} = \begin{pmatrix} \frac{18}{7\sqrt{13}} \\ \frac{12}{7\sqrt{13}} \\ \frac{\sqrt{13}}{7} \end{pmatrix} \quad \mathbf{u_2} = \begin{pmatrix} \frac{-2}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} \\ 0 \end{pmatrix} \quad \mathbf{u_3} = \begin{pmatrix} \frac{-7}{12} \\ \frac{-7}{18} \\ \frac{7}{6} \end{pmatrix}$$
(1.1.1.18)

Hence we obtain **U** as follows,

$$\mathbf{U} = \begin{pmatrix} \frac{18}{7\sqrt{13}} & \frac{-2}{\sqrt{13}} & \frac{-7}{12} \\ \frac{12}{7\sqrt{13}} & \frac{3}{\sqrt{13}} & \frac{-7}{18} \\ \frac{\sqrt{13}}{7} & 0 & \frac{7}{6} \end{pmatrix}$$
(1.1.1.19)

By computing the singular values from eigen values  $\lambda_1, \lambda_2, \lambda_3$  we get  $\Sigma$  as,

$$\Sigma = \begin{pmatrix} \frac{49}{36} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.1.1.20}$$

Calculating eigen values of  $\mathbf{M}^T\mathbf{M}$ ,

$$\left|\mathbf{M}^T \mathbf{M} - \lambda \mathbf{I}\right| = 0 \qquad (1.1.1.21)$$

$$\implies \begin{vmatrix} \frac{5}{4} - \lambda & \frac{1}{6} \\ \frac{1}{6} & \frac{10}{9} - \lambda \end{vmatrix} = 0 \qquad (1.1.1.22)$$

$$\implies \lambda^2 - \frac{85}{36}\lambda + \frac{49}{36} = 0 \qquad (1.1.1.23)$$

From the characteristic equation, the eigen values of  $\mathbf{M}^T \mathbf{M}$  are,

$$\lambda_1 = \frac{49}{36} \quad \lambda_2 = 1 \tag{1.1.1.24}$$

Hence the eigen vectors of  $\mathbf{M}^T \mathbf{M}$  are,

$$\mathbf{v}_1 = \begin{pmatrix} \frac{3}{2} \\ 1 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} \frac{-2}{3} \\ 1 \end{pmatrix} \tag{1.1.1.25}$$

Normalizing the eigen vectors,

$$\mathbf{v}_1 = \begin{pmatrix} \frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} \frac{-2}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} \end{pmatrix}$$
 (1.1.1.26)

Hence we obtain V as,

$$\mathbf{V} = \begin{pmatrix} \frac{3}{\sqrt{13}} & \frac{-2}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} & \frac{3}{\sqrt{13}} \end{pmatrix}$$
 (1.1.1.27)

From (1.1.1.6), the Singular Value Decomposition of **M** is as follows,

$$\mathbf{M} = \begin{pmatrix} \frac{18}{7\sqrt{13}} & \frac{-2}{\sqrt{13}} & \frac{-7}{12} \\ \frac{12}{7\sqrt{13}} & \frac{3}{\sqrt{13}} & \frac{-7}{18} \\ \frac{\sqrt{13}}{7} & 0 & \frac{7}{6} \end{pmatrix} \begin{pmatrix} \frac{49}{36} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{3}{\sqrt{13}} & \frac{-2}{\sqrt{13}} \\ \frac{\sqrt{13}}{\sqrt{13}} & \frac{3}{\sqrt{13}} \end{pmatrix}^{T}$$

$$(1.1.1.28)$$

And, the Moore-Penrose Pseudo inverse of  $\Sigma$  is given by,

$$\Sigma^{-1} = \begin{pmatrix} \frac{6}{7} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.1.29}$$

From (1.1.1.12) we get,

$$\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{-17}{7\sqrt{13}} \\ \frac{12}{\sqrt{13}} \\ \frac{77}{36} \end{pmatrix}$$
 (1.1.1.30)

$$\Sigma^{-1}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{-102}{49\sqrt{13}} \\ \frac{12}{\sqrt{13}} \end{pmatrix}$$
 (1.1.1.31)

$$\mathbf{x} = \mathbf{V} \mathbf{\Sigma}^{-1} \mathbf{U}^T \mathbf{b} = \begin{pmatrix} \frac{-114}{49} \\ \frac{120}{49} \end{pmatrix} \quad (1.1.1.32)$$

Now we verify the solution (1.1.1.32) using,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \implies \mathbf{M}^T \mathbf{M}\mathbf{x} = \mathbf{M}^T \mathbf{b}$$
 (1.1.1.33)

On evaluating the R.H.S in (1.1.1.33) we get,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \begin{pmatrix} \frac{-5}{2} \\ \frac{7}{3} \end{pmatrix} \tag{1.1.1.34}$$

$$\implies \begin{pmatrix} \frac{5}{4} & \frac{1}{6} \\ \frac{1}{6} & \frac{10}{9} \end{pmatrix} \mathbf{x} = \begin{pmatrix} \frac{-5}{2} \\ \frac{7}{3} \end{pmatrix} \tag{1.1.1.35}$$

On solving the augmented matrix of (1.1.1.35) we get,

$$\begin{pmatrix} \frac{5}{4} & \frac{1}{6} & \frac{-5}{2} \\ \frac{1}{6} & \frac{10}{9} & \frac{7}{3} \end{pmatrix} \stackrel{R_1 = \frac{4R_1}{5}}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{2}{15} & -2 \\ \frac{1}{6} & \frac{10}{9} & \frac{7}{3} \end{pmatrix} (1.1.1.36)$$

$$\stackrel{R_2 = R_2 - \frac{R_1}{6}}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{2}{15} & -2 \\ 0 & \frac{49}{45} & \frac{8}{3} \end{pmatrix} (1.1.1.37)$$

$$\stackrel{R_2 = \frac{45}{49}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{2}{15} & -2 \\ 0 & 1 & \frac{120}{49} \end{pmatrix} (1.1.1.38)$$

$$\stackrel{R_1 = R_1 - \frac{2R_2}{15}}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{-114}{49} \\ 0 & 1 & \frac{120}{49} \end{pmatrix}$$

$$\begin{array}{cccc}
 & & \downarrow \\
 &$$

From equation (1.1.1.39), solution is given by,

$$\mathbf{x} = \begin{pmatrix} \frac{-114}{49} \\ \frac{120}{49} \end{pmatrix} \tag{1.1.1.40}$$

From the equations (1.1.1.32) and (1.1.1.40), the solution  $\mathbf{x}$  is verified.

b) Find the foot of perpendicular from point  $B = \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix}$  to the plane  $\begin{pmatrix} 2 & 3 & -4 \end{pmatrix} \mathbf{x} = -5$ .

**Solution:** Let us consider orthogonal vectors  $\mathbf{m}_1$  and  $\mathbf{m}_2$  to the given normal vector  $\mathbf{n}$ . Let

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

Then.

$$\mathbf{m}^T \mathbf{n} = 0 \qquad (1.1.1.41)$$

$$\implies \left(a \quad b \quad c\right) \begin{pmatrix} 2\\3\\-4 \end{pmatrix} = 0 \quad (1.1.1.42)$$

$$\implies$$
 2a + 3b - 4c = 0 (1.1.1.43)

Let a = 1, b = 0, so that

$$\mathbf{m}_1 = \begin{pmatrix} 1\\0\\\frac{1}{2} \end{pmatrix} \tag{1.1.1.44}$$

and a = 0, b = 1, so that

$$\mathbf{m}_2 = \begin{pmatrix} 0\\1\\\frac{3}{4} \end{pmatrix} \tag{1.1.1.45}$$

We, now, solve the equation

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.1.46}$$

which, upon substitution, becomes

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{1}{2} & \frac{3}{4} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix} \tag{1.1.1.47}$$

Any  $m \times n$  matrix **M** can be factorized in SVD form as

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.1.1.48}$$

where  $\mathbf{U}$  and  $\mathbf{V}$  are matrices of eigen vectors which are orthogonal. Columns of  $\mathbf{V}$  are the eigen vectors of  $\mathbf{M}^T\mathbf{M}$ , columns of  $\mathbf{U}$  are the eigen vectors of  $\mathbf{M}\mathbf{M}^T$  and  $\mathbf{S}$  is the diagonal matrix of singular values of  $\mathbf{M}$  of the eigenvalues of  $\mathbf{M}^T\mathbf{M}$ .

$$\mathbf{M}^{T}\mathbf{M} = \begin{pmatrix} \frac{10}{8} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix}$$
 (1.1.1.49)

Putting (1.1.1.48) into (1.1.1.46), we get

$$\mathbf{USV}^T\mathbf{x} = \mathbf{b} \tag{1.1.1.50}$$

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} \qquad (1.1.1.51)$$

where  $S_+$  is the Moore-Penrose Pseudoinverse of S.

The eigenvalues of  $\mathbf{M}^T\mathbf{M}$ :

$$\left|\mathbf{M}^{T}\mathbf{M} - \lambda \mathbf{I}\right| = 0 \quad (1.1.1.52)$$

$$\implies \left| \frac{\frac{10}{8} - \lambda}{\frac{3}{8}} \right| \frac{\frac{3}{8}}{\frac{25}{16} - \lambda} = 0 \quad (1.1.1.53)$$

$$\implies \lambda^2 - \frac{45}{16}\lambda + \frac{116}{64} = 0 \quad (1.1.1.54)$$

So, the eigenvalues are

$$\lambda_1 = \frac{29}{16} \tag{1.1.1.55}$$

$$\lambda_2 = 1 \tag{1.1.1.56}$$

For  $\lambda_1 = \frac{29}{16}$ , the eigen vector  $\mathbf{v_1}$  can be calculated using row reduction as :

$$\begin{pmatrix} -\frac{9}{16} & \frac{3}{8} \\ \frac{3}{8} & -\frac{4}{16} \end{pmatrix} \stackrel{R_1 \leftarrow -\frac{16}{9}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & -\frac{2}{3} \\ \frac{3}{8} & -\frac{4}{16} \end{pmatrix} \quad (1.1.1.57)$$

$$\stackrel{R_2 \leftarrow R_2 - R_1}{\longleftrightarrow} \begin{pmatrix} 1 & -\frac{2}{3} \\ 0 & 0 \end{pmatrix} \quad (1.1.1.58)$$

Hence,

$$\mathbf{v_1} = \begin{pmatrix} \frac{2}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} \end{pmatrix} \tag{1.1.1.59}$$

Similarly, for  $\lambda_2 = 1$ ,

$$\mathbf{v_2} = \begin{pmatrix} -\frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{\sqrt{13}}} \end{pmatrix}$$
 (1.1.1.60)

Thus,

$$\mathbf{V} = \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}$$
 (1.1.1.61)

Now,

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} \end{pmatrix}$$
 (1.1.1.62)

Now, calculating eigenvalues of  $\mathbf{M}\mathbf{M}^T$ 

$$\begin{vmatrix} 1 - \lambda & 0 & \frac{1}{2} \\ 0 & 1 - \lambda & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} - \lambda \end{vmatrix} = 0 \quad (1.1.1.63)$$

So, the eigenvalues are

$$\lambda_1 = \frac{29}{16} \tag{1.1.1.64}$$

$$\lambda_2 = 1$$
 (1.1.1.65)

$$\lambda_3 = 0$$
 (1.1.1.66)

For  $\lambda_1 = \frac{29}{16}$ , the eigen vector can be computed as:

$$\begin{pmatrix}
1 - \frac{29}{16} & 0 & \frac{1}{2} \\
0 & 1 - \frac{29}{16} & \frac{3}{4} \\
\frac{1}{2} & \frac{3}{4} & \frac{13}{16} - \frac{29}{16}
\end{pmatrix}$$
(1.1.1.67)

$$\leftrightarrow \begin{pmatrix}
-\frac{13}{16} & 0 & \frac{1}{2} \\
0 & -\frac{13}{16} & \frac{3}{4} \\
\frac{1}{2} & \frac{3}{4} & -1
\end{pmatrix}$$
(1.1.1.68)

$$\stackrel{R_1 \leftarrow -\frac{16}{13}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -\frac{8}{3} \\ 0 & -\frac{13}{16} & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & -1 \end{pmatrix}$$
(1.1.1.69)

$$\stackrel{R_3 \leftarrow R_3 - \frac{1}{2}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -\frac{8}{3} \\ 0 & -\frac{13}{16} & \frac{3}{4} \\ 0 & \frac{3}{4} & -\frac{9}{13} \end{pmatrix} (1.1.1.70)$$

$$\stackrel{R_2 \leftarrow -\frac{16}{13}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -\frac{8}{3} \\ 0 & 1 & -\frac{12}{13} \\ 0 & \frac{3}{4} & -\frac{9}{13} \end{pmatrix}$$
(1.1.1.71)

$$\stackrel{R_2 \leftarrow R_3 - \frac{3}{4}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & -\frac{8}{3} \\ 0 & 1 & -\frac{12}{13} \\ 0 & 0 & 0 \end{pmatrix}$$
(1.1.1.72)

Hence, the eigen vector  $\mathbf{u}_1$ :

$$\mathbf{u_1} = \begin{pmatrix} \frac{8}{\sqrt{377}} \\ \frac{12}{\sqrt{377}} \\ \frac{13}{\sqrt{377}} \end{pmatrix} \tag{1.1.1.73}$$

For  $\lambda_2 = 1$ , the eigen vector is:

$$\begin{pmatrix}
1-1 & 0 & \frac{1}{2} \\
0 & 1-1 & \frac{3}{4} \\
\frac{1}{2} & \frac{3}{4} & \frac{13}{16} - 1
\end{pmatrix}$$
(1.1.1.74)

$$\longleftrightarrow \begin{pmatrix} 0 & 0 & \frac{1}{2} \\ 0 & 0 & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & -\frac{3}{16} \end{pmatrix}$$
 (1.1.1.75)

Hence, the eigen vector  $\mathbf{u}_2$ :

$$\mathbf{u_2} = \begin{pmatrix} \frac{3}{\sqrt{13}} \\ -\frac{2}{\sqrt{13}} \\ 0 \end{pmatrix} \tag{1.1.1.76}$$

Similarly, for  $\lambda_3 = 0$ , the eigen vector is:

$$\begin{pmatrix}
1 & 0 & \frac{1}{2} \\
0 & 1 & \frac{3}{4} \\
\frac{1}{2} & \frac{3}{4} & \frac{13}{16}
\end{pmatrix}$$
(1.1.1.77)

$$\xrightarrow{R_3 \leftarrow R_3 - \frac{1}{2}R_1 - \frac{3}{4}R_2} \begin{pmatrix} 0 & 0 & \frac{1}{2} \\ 0 & 0 & \frac{3}{4} \\ 0 & 0 & 0 \end{pmatrix}$$
 (1.1.1.78)

Hence, the eigen vector  $\mathbf{u}_3$ :

$$\mathbf{u_3} = \begin{pmatrix} \frac{2}{\sqrt{29}} \\ \frac{3}{\sqrt{29}} \\ -\frac{4}{\sqrt{\sqrt{90}}} \end{pmatrix}$$
 (1.1.1.79)

So, the orthonormal matrix U of eigen vectors is:

$$\mathbf{U} = \begin{pmatrix} \frac{8}{\sqrt{377}} & \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{29}} \\ \frac{12}{\sqrt{377}} & -\frac{2}{\sqrt{13}} & \frac{3}{\sqrt{29}} \\ \frac{13}{\sqrt{377}} & 0 & -\frac{4}{\sqrt{29}} \end{pmatrix}$$
(1.1.1.80)

The matrix of singular values of **M** is:

$$\mathbf{S} = \begin{pmatrix} \frac{\sqrt{29}}{4} & 0 & 0\\ 0 & 1 & 0\\ 0 & 0 & 0 \end{pmatrix} \tag{1.1.1.81}$$

The Moore-Penrose pseudoinverse of S is computed as

$$\mathbf{S}_{+} = (\mathbf{S}\mathbf{S}^{T})^{-1}\mathbf{S}^{T}$$
 (1.1.1.82)  
=  $\begin{pmatrix} \frac{4}{\sqrt{29}} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix}$  (1.1.1.83)

To solve for  $\mathbf{x}$  in (1.1.1.51), noting that  $\mathbf{b} =$  $\begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix}$ 

$$\mathbf{U}^T \mathbf{b} = \begin{pmatrix} 0 \\ \sqrt{13} \\ 0 \end{pmatrix} \tag{1.1.1.84}$$

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 0\\\sqrt{13} \end{pmatrix} \tag{1.1.1.85}$$

Thus, the foot of perpendicular is:

$$\mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix} \begin{pmatrix} 0 \\ \sqrt{13} \end{pmatrix}$$

$$(1.1.1.86)$$

$$\implies \quad \mathbf{x} = \begin{pmatrix} -3 \\ 2 \end{pmatrix} \qquad (1.1.1.87)$$

(1.1.1.87)

The solution can be verified using

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.1.88}$$

The LHS gives

$$\mathbf{M}^{T}\mathbf{M}\mathbf{x} = \begin{pmatrix} \frac{10}{8} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix} \begin{pmatrix} -3 \\ 2 \end{pmatrix} \quad (1.1.1.89)$$

$$\Longrightarrow \mathbf{M}^T \mathbf{M} \mathbf{x} = \begin{pmatrix} -3\\2 \end{pmatrix} \tag{1.1.1.90}$$

Now, finding x from

$$\begin{pmatrix} \frac{10}{8} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix} \mathbf{x} = \begin{pmatrix} -3 \\ 2 \end{pmatrix}$$
 (1.1.1.91)

Solving the augmented matrix, we get

$$\begin{pmatrix} \frac{10}{8} & \frac{3}{8} & -3\\ \frac{3}{8} & \frac{25}{16} & 2 \end{pmatrix} \xrightarrow{R_1 \leftarrow -\frac{3}{10}R_1} \begin{pmatrix} 1 & \frac{3}{10} & -\frac{24}{10}\\ \frac{3}{8} & \frac{25}{16} & 2 \end{pmatrix}$$
(1.1.1.92)

$$\xrightarrow{R_2 \leftarrow R_2 - \frac{3}{8}R_1} \begin{pmatrix} 1 & \frac{3}{10} & -\frac{24}{10} \\ 0 & \frac{29}{20} & \frac{58}{20} \end{pmatrix} \xrightarrow{R_2 \leftarrow \frac{20}{29}R_2} \begin{pmatrix} 1 & \frac{3}{10} & -\frac{24}{10} \\ 0 & 1 & 2 \end{pmatrix}$$

$$(1.1.1.93)$$

$$\xrightarrow{R_1 \leftarrow R_1 - \frac{3}{10}R_2} \begin{pmatrix} 1 & 0 & -3 \\ 0 & 1 & 2 \end{pmatrix}$$

$$(1.1.1.94)$$

Hence, the solution is given by

$$\mathbf{x} = \begin{pmatrix} -3\\2 \end{pmatrix} \tag{1.1.1.95}$$

Comparing the results in Eq.(1.1.1.87) and (1.1.1.95), it is concluded that the solution is verified.

### 1.1.2. Solve the following

a) Find the foot of the perpendicular from,

$$\mathbf{A} = \begin{pmatrix} 1\\4\\-3 \end{pmatrix} \tag{1.1.2.1}$$

to the plane,

$$(2 -3 1)\mathbf{x} = 0 (1.1.2.2)$$

**Solution:** The equation of plane is given as,

$$\mathbf{n}^T \mathbf{x} = c \tag{1.1.2.3}$$

Hence the normal vector **n** is,

$$\mathbf{n} = \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix} \tag{1.1.2.4}$$

Let, the normal vectors  $\mathbf{m_1}$  and  $\mathbf{m_2}$  to the normal vector  $\mathbf{n}$  be,

$$\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{1.1.2.5}$$

then, 
$$\mathbf{m}^T \mathbf{n} = 0$$
 (1.1.2.6)

$$\implies \begin{pmatrix} a & b & c \end{pmatrix} \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix} = 0 \qquad (1.1.2.7)$$

Let, a=0 and b=1 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ -2 \end{pmatrix} \tag{1.1.2.8}$$

Let, a=1 and b=0,

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix} \tag{1.1.2.9}$$

Now solving the equation,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.2.10}$$

Where,

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} \tag{1.1.2.11}$$

and, 
$$\mathbf{b} = \begin{pmatrix} 1\\4\\-3 \end{pmatrix}$$
 (1.1.2.12)

To solve (1.1.2.10) we perform singular value decomposition on M given by,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.1.2.13}$$

substituting the value of M from equation (1.1.2.13) to (1.1.2.10),

$$\implies$$
 **USV**<sup>T</sup>**x** = **b** (1.1.2.14)

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} \tag{1.1.2.15}$$

where,  $S_+$  is Moore-Pen-rose Pseudo-Inverse of S. Columns of U are eigenvectors of  $\mathbf{M}\mathbf{M}^T$ , columns of V are eigenvectors of  $\mathbf{M}^T\mathbf{M}$  and S is diagonal matrix of singular value of eigenvalues of  $\mathbf{M}^T\mathbf{M}$ . First calculating the eigenvectors corresponding to  $\mathbf{M}^T \mathbf{M}$ .

$$\mathbf{M}^{T}\mathbf{M} = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} = \begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix}$$
(1.1.2.16)

Eigenvalues corresponding to  $\mathbf{M}^T \mathbf{M}$  is,

$$\left|\mathbf{M}^{T}\mathbf{M} - \lambda \mathbf{I}\right| = 0 \qquad (1.1.2.17)$$

$$\implies \begin{pmatrix} 5 - \lambda & -6 \\ -6 & 10 - \lambda \end{pmatrix} \qquad (1.1.2.18)$$

$$\implies (\lambda - 14)(\lambda - 1) = 0$$
 (1.1.2.19)

$$\therefore \lambda_1 = 14 \qquad (1.1.2.20)$$

$$\lambda_2 = 1$$
 (1.1.2.21)

Hence the eigenvectors corresponding to  $\lambda_1$  and  $\lambda_2$  respectively is,

$$\mathbf{v_1} = \begin{pmatrix} \frac{-2}{3} \\ 1 \end{pmatrix} \tag{1.1.2.22}$$

$$\mathbf{v_2} = \begin{pmatrix} \frac{3}{2} \\ 1 \end{pmatrix} \tag{1.1.2.23}$$

Normalizing the eigenvectors we get,

$$\mathbf{v_1} = \frac{1}{\sqrt{13}} \begin{pmatrix} -2\\3 \end{pmatrix} \qquad (1.1.2.24)$$

$$\mathbf{v_2} = \frac{1}{\sqrt{13}} \begin{pmatrix} 3\\2 \end{pmatrix} \qquad (1.1.2.25)$$

$$\implies \mathbf{V} = \frac{1}{\sqrt{13}} \begin{pmatrix} -2 & 3\\ 3 & 2 \end{pmatrix} \qquad (1.1.2.26)$$

Now calculating the eigenvectors corresponding to  $\mathbf{MM}^T$ 

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \end{pmatrix} \quad (1.1.2.27)$$

$$\implies \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \\ -2 & 3 & 13 \end{pmatrix} \quad (1.1.2.28)$$

Eigenvalues corresponding to  $\mathbf{M}\mathbf{M}^T$  is,

$$\begin{aligned} |\mathbf{M}\mathbf{M}^{T} - \lambda \mathbf{I}| &= 0 \quad (1.1.2.29) \\ \Longrightarrow \begin{pmatrix} 1 - \lambda & 0 & -2 \\ 0 & 1 - \lambda & 3 \\ -2 & 3 & 13 - \lambda \end{pmatrix} \quad (1.1.2.30) \\ \Longrightarrow -\lambda^{3} + 15\lambda^{2} - 14\lambda &= 0 \quad (1.1.2.31) \\ \Longrightarrow -\lambda(\lambda - 1)(\lambda - 14) &= 0 \quad (1.1.2.32) \\ \therefore \lambda_{3} &= 14 \quad (1.1.2.33) \\ \lambda_{4} &= 1 \quad (1.1.2.34) \\ \lambda_{5} &= 0 \quad (1.1.2.35) \end{aligned}$$

Hence the eigenvectors corresponding to  $\lambda_3$ ,  $\lambda_4$  and  $\lambda_5$  respectively is,

$$\mathbf{v_3} = \begin{pmatrix} \frac{-2}{13} \\ \frac{3}{13} \\ 1 \end{pmatrix} \tag{1.1.2.36}$$

$$\mathbf{v_4} = \begin{pmatrix} \frac{3}{2} \\ 1 \\ 0 \end{pmatrix} \tag{1.1.2.37}$$

$$\mathbf{v_5} = \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix} \tag{1.1.2.38}$$

Normalizing the eigenvectors we get,

$$\mathbf{v_3} = \frac{1}{\sqrt{182}} \begin{pmatrix} -2\\3\\13 \end{pmatrix} = \begin{pmatrix} -\sqrt{\frac{2}{91}}\\\frac{3}{\sqrt{182}}\\\sqrt{\frac{13}{14}} \end{pmatrix} (1.1.2.39)$$

$$\mathbf{v_4} = \frac{1}{\sqrt{13}} \begin{pmatrix} 3\\2\\0 \end{pmatrix} = \begin{pmatrix} \frac{3}{\sqrt{13}}\\ \frac{2}{\sqrt{13}}\\0 \end{pmatrix} (1.1.2.40)$$

$$\mathbf{v_5} = \frac{1}{\sqrt{14}} \begin{pmatrix} 2\\ -3\\ 1 \end{pmatrix} = \begin{pmatrix} \sqrt{\frac{2}{7}}\\ -\frac{3}{\sqrt{14}}\\ \sqrt{\frac{1}{14}} \end{pmatrix} (1.1.2.41)$$

$$\implies \mathbf{U} = \begin{pmatrix} -\sqrt{\frac{2}{91}} & \frac{3}{\sqrt{13}} & \sqrt{\frac{2}{7}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{14}} \\ \sqrt{\frac{13}{14}} & 0 & \sqrt{\frac{1}{14}} \end{pmatrix} (1.1.2.42)$$

Now **S** corresponding to eigenvalues  $\lambda_3$ ,  $\lambda_4$ 

and  $\lambda_5$  is as follows,

$$\mathbf{S} = \begin{pmatrix} \sqrt{14} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \tag{1.1.2.43}$$

Now, Moore-Penrose Pseudo inverse of S is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{1}{\sqrt{14}} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.2.44}$$

Hence we get singular value decomposition of  $\mathbf{M}$  as,

$$\mathbf{M} = \frac{1}{\sqrt{13}} \begin{pmatrix} -\sqrt{\frac{2}{91}} & \frac{3}{\sqrt{13}} & \sqrt{\frac{2}{7}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{14}} \\ \sqrt{\frac{13}{14}} & 0 & \sqrt{\frac{1}{14}} \end{pmatrix} \begin{pmatrix} \sqrt{14} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} -2 & 3 \\ 3 & 2 \end{pmatrix}^{T}$$

$$(1.1.2.45)$$

Now substituting the values of (1.1.2.26), (1.1.2.44), (1.1.2.42) and (1.1.2.12) in (1.1.2.15),

$$\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} -\sqrt{\frac{2}{91}} & \frac{3}{\sqrt{13}} & \sqrt{\frac{2}{7}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{14}} \end{pmatrix}^{T} \begin{pmatrix} 1\\4\\-3 \end{pmatrix}$$

$$(1.1.2.46)$$

$$\implies \mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{-29}{\sqrt{182}} \\ \frac{11}{\sqrt{13}} \\ \frac{-13}{\sqrt{14}} \end{pmatrix}$$

$$(1.1.2.47)$$

$$\mathbf{VS}_{+} = \frac{1}{\sqrt{13}} \begin{pmatrix} -2 & 3 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{14}} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

$$(1.1.2.48)$$

$$\implies \mathbf{VS}_{+} = \frac{1}{\sqrt{13}\sqrt{14}} \begin{pmatrix} -2 & 3\sqrt{14} & 0 \\ 3 & 2\sqrt{14} & 0 \end{pmatrix}$$

$$(1.1.2.49)$$

 $\therefore$  from equation (1.1.2.15),

$$\mathbf{x} = \frac{1}{\sqrt{13}\sqrt{14}} \begin{pmatrix} -2 & 3\sqrt{14} & 0\\ 3 & 2\sqrt{14} & 0 \end{pmatrix} \begin{pmatrix} \frac{-29}{\sqrt{182}}\\ \frac{11}{\sqrt{13}}\\ \frac{-13}{\sqrt{14}} \end{pmatrix}$$
(1.1.2.50)

$$\implies \mathbf{x} = \begin{pmatrix} \frac{20}{7} \\ \frac{17}{14} \end{pmatrix} \tag{1.1.2.51}$$

Verifying the solution using,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.2.52}$$

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

$$(1.1.2.53)$$

$$\Rightarrow \begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 7 \\ -5 \end{pmatrix}$$

$$(1.1.2.54)$$

Solving the augmented matrix we get,

$$\begin{pmatrix}
5 & -6 & 7 \\
-6 & 10 & -5
\end{pmatrix}
\longleftrightarrow
\begin{pmatrix}
R_1 \leftarrow \frac{R_1}{5} \\
-6 & 10 & -5
\end{pmatrix}$$

$$(1.1.2.55)$$

$$\stackrel{R_2 \leftarrow R_2 + 6R_1}{\longleftrightarrow} \begin{pmatrix}
1 & -\frac{6}{5} & \frac{7}{5} \\
0 & \frac{14}{5} & \frac{17}{5}
\end{pmatrix}$$

$$(1.1.2.56)$$

$$\stackrel{R_2 \leftarrow \frac{5}{14}R_2}{\longleftrightarrow} \begin{pmatrix}
1 & -\frac{6}{5} & \frac{7}{5} \\
0 & 1 & \frac{17}{14}
\end{pmatrix}$$

$$(1.1.2.57)$$

$$\stackrel{R_1 \leftarrow R_1 + \frac{6}{5}R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & \frac{20}{7} \\
0 & 1 & \frac{17}{14}
\end{pmatrix}$$

$$(1.1.2.58)$$

$$\Longrightarrow \mathbf{x} = \begin{pmatrix}
\frac{20}{7} \\
\frac{17}{14}
\end{pmatrix}$$

$$(1.1.2.59)$$

Hence from equations (1.1.2.51) and (1.1.2.59) we conclude that the solution is verified.

b) Find the foot of the perpendicular from,

$$\mathbf{B} = \begin{pmatrix} 3 \\ -2 \\ 2 \end{pmatrix} \tag{1.1.2.60}$$

to the plane,

$$(2 -3 1)\mathbf{x} = 0 (1.1.2.62)$$

**Solution:** The equation of plane is give

$$\mathbf{n}^T \mathbf{x} = c \tag{1.1.2.63}$$

Hence the normal vector  $\mathbf{n}$  is,

$$\mathbf{n} = \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix} \tag{1.1.2.64}$$

Let, the normal vectors  $\mathbf{m_1}$  and  $\mathbf{m_2}$  to the normal vector  $\mathbf{n}$  be,

$$\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \qquad (1.1.2.65)$$

then, 
$$\mathbf{m}^T \mathbf{n} = 0$$
 (1.1.2.66)

$$\implies \begin{pmatrix} a & b & c \end{pmatrix} \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix} = 0 \qquad (1.1.2.67)$$

Let, a=0 and b=1 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1\\0\\-2 \end{pmatrix} \tag{1.1.2.68}$$

Let, a=1 and b=0,

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix} \tag{1.1.2.69}$$

Now solving the equation,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.2.70}$$

Where,

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 3 \\ -2 \\ 2 \end{pmatrix} \tag{1.1.2.71}$$

To solve (1.1.2.70) we perform singular value decomposition on M given by,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.1.2.72}$$

substituting the value of M from equation (1.1.2.72) to (1.1.2.70),

$$\implies \mathbf{U}\mathbf{S}\mathbf{V}^T\mathbf{x} = \mathbf{b} \tag{1.1.2.73}$$

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} \tag{1.1.2.74}$$

where,  $S_+$  is Moore-Pen-rose Pseudo-Inverse of S. Columns of U are eigenvectors of  $\mathbf{M}\mathbf{M}^T$ , columns of V are eigenvectors of  $\mathbf{M}^T\mathbf{M}$  and S is diagonal matrix of singular value of eigenvalues of  $\mathbf{M}^T\mathbf{M}$ . First calculating the eigenvectors corresponding to  $\mathbf{M}^T \mathbf{M}$ .

$$\mathbf{M}^{T}\mathbf{M} = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} = \begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix}$$
(1.1.2.75)

Eigenvalues corresponding to  $\mathbf{M}^T \mathbf{M}$  is,

$$\begin{vmatrix} \mathbf{M}^{T}\mathbf{M} - \lambda \mathbf{I} | = 0 & (1.1.2.76) \\ \Rightarrow \begin{pmatrix} 5 - \lambda & -6 \\ -6 & 10 - \lambda \end{pmatrix} & (1.1.2.77) \\ \Rightarrow (\lambda - 14)(\lambda - 1) = 0 & (1.1.2.78) \\ \therefore \lambda_{1} = 14, \lambda_{2} = 1, & (1.1.2.79) \end{vmatrix}$$

Hence the eigenvectors corresponding to  $\lambda_1$  and  $\lambda_2$  respectively is,

$$\mathbf{v_1} = \begin{pmatrix} \frac{-2}{3} \\ 1 \end{pmatrix}, \mathbf{v_2} = \begin{pmatrix} \frac{3}{2} \\ 1 \end{pmatrix}$$
 (1.1.2.80)

Normalizing the eigenvectors we get,

$$\mathbf{v_1} = \frac{1}{\sqrt{13}} \begin{pmatrix} -2\\3 \end{pmatrix} \qquad (1.1.2.81)$$

$$\mathbf{v_2} = \frac{1}{\sqrt{13}} \begin{pmatrix} 3\\2 \end{pmatrix} \qquad (1.1.2.82)$$

$$\implies \mathbf{V} = \frac{1}{\sqrt{13}} \begin{pmatrix} -2 & 3\\ 3 & 2 \end{pmatrix} \qquad (1.1.2.83)$$

Now calculating the eigenvectors corresponding to  $\mathbf{MM}^T$ 

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \end{pmatrix} \quad (1.1.2.84)$$

$$\implies \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \\ -2 & 3 & 13 \end{pmatrix} \quad (1.1.2.85)$$

Eigenvalues corresponding to  $\mathbf{M}\mathbf{M}^T$  is,

$$\begin{aligned} |\mathbf{M}\mathbf{M}^{T} - \lambda \mathbf{I}| &= 0 \quad (1.1.2.86) \\ \Longrightarrow \begin{pmatrix} 1 - \lambda & 0 & -2 \\ 0 & 1 - \lambda & 3 \\ -2 & 3 & 13 - \lambda \end{pmatrix} \quad (1.1.2.87) \\ \Longrightarrow -\lambda^{3} + 15\lambda^{2} - 14\lambda &= 0 \quad (1.1.2.88) \\ \Longrightarrow -\lambda(\lambda - 1)(\lambda - 14) &= 0 \quad (1.1.2.89) \\ \therefore \lambda_{3} &= 14, \lambda_{4} &= 1 \quad (1.1.2.90) \\ \lambda_{5} &= 0 \quad (1.1.2.91) \end{aligned}$$

Hence the eigenvectors corresponding to  $\lambda_3$ ,

 $\lambda_4$  and  $\lambda_5$  respectively is,

$$\mathbf{v_3} = \begin{pmatrix} \frac{-2}{13} \\ \frac{3}{13} \\ 1 \end{pmatrix}, \mathbf{v_4} = \begin{pmatrix} \frac{3}{2} \\ 1 \\ 0 \end{pmatrix}, \mathbf{v_5} = \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix}$$
 (1.1.2.92)

Normalizing the eigenvectors we get,

$$\mathbf{v_3} = \frac{1}{\sqrt{182}} \begin{pmatrix} -2\\3\\13 \end{pmatrix} = \begin{pmatrix} -\sqrt{\frac{2}{91}}\\\frac{3}{\sqrt{182}}\\\sqrt{\frac{13}{14}} \end{pmatrix} (1.1.2.93)$$

$$\mathbf{v_4} = \frac{1}{\sqrt{13}} \begin{pmatrix} 3\\2\\0 \end{pmatrix} = \begin{pmatrix} \frac{3}{\sqrt{13}}\\ \frac{2}{\sqrt{13}}\\0 \end{pmatrix} \quad (1.1.2.94)$$

$$\mathbf{v_5} = \frac{1}{\sqrt{14}} \begin{pmatrix} 2\\ -3\\ 1 \end{pmatrix} = \begin{pmatrix} \sqrt{\frac{2}{7}}\\ -\frac{3}{\sqrt{14}}\\ \sqrt{\frac{1}{14}} \end{pmatrix} (1.1.2.95)$$

$$\implies \mathbf{U} = \begin{pmatrix} -\sqrt{\frac{2}{91}} & \frac{3}{\sqrt{13}} & \sqrt{\frac{2}{7}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{14}} \\ \sqrt{\frac{13}{14}} & 0 & \sqrt{\frac{1}{14}} \end{pmatrix} (1.1.2.96)$$

Now **S** corresponding to eigenvalues  $\lambda_3$ ,  $\lambda_4$  and  $\lambda_5$  is as follows,

$$\mathbf{S} = \begin{pmatrix} \sqrt{14} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.1.2.97}$$

Now, Moore-Penrose Pseudo inverse of S is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{1}{\sqrt{14}} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.2.98}$$

Hence we get singular value decomposition of M as,

$$\mathbf{M} = \frac{1}{\sqrt{13}} \begin{pmatrix} -\sqrt{\frac{2}{91}} & \frac{3}{\sqrt{13}} & \sqrt{\frac{2}{7}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{14}} \\ \sqrt{\frac{13}{14}} & 0 & \sqrt{\frac{1}{14}} \end{pmatrix} \begin{pmatrix} \sqrt{14} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} -2 & 3 \\ 3 & 2 \end{pmatrix}^{T}$$

$$(1.1.2.99)$$

Now substituting the values of (1.1.2.83), (1.1.2.98), (1.1.2.96) and (1.1.2.71) in

(1.1.2.74),

$$\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} -\sqrt{\frac{2}{91}} & \frac{3}{\sqrt{13}} & \sqrt{\frac{2}{7}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{14}} \end{pmatrix} \begin{pmatrix} 3 \\ -2 \\ 2 \end{pmatrix}$$

$$(1.1.2.100)$$

$$\Rightarrow \mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{\sqrt{182}}{13} \\ \frac{5}{\sqrt{13}} \\ \sqrt{14} \end{pmatrix}$$

$$(1.1.2.101)$$

$$\mathbf{VS}_{+} = \frac{1}{\sqrt{13}} \begin{pmatrix} -2 & 3 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{14}} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

$$(1.1.2.102)$$

$$\Rightarrow \mathbf{VS}_{+} = \frac{1}{\sqrt{13}\sqrt{14}} \begin{pmatrix} -2 & 3\sqrt{14} & 0 \\ 3 & 2\sqrt{14} & 0 \end{pmatrix}$$

$$(1.1.2.103)$$

 $\therefore$  from equation (1.1.2.74),

$$\mathbf{x} = \frac{1}{\sqrt{13}\sqrt{14}} \begin{pmatrix} -2 & 3\sqrt{14} & 0\\ 3 & 2\sqrt{14} & 0 \end{pmatrix} \begin{pmatrix} \frac{\sqrt{182}}{13}\\ \frac{5}{\sqrt{13}}\\ \sqrt{14} \end{pmatrix}$$
(1.1.2.104)

$$\implies \mathbf{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \tag{1.1.2.105}$$

Verifying the solution using,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.2.106}$$

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

$$(1.1.2.107)$$

$$\Rightarrow \begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix} \mathbf{x} = \begin{pmatrix} -1 \\ 4 \end{pmatrix}$$

$$(1.1.2.108)$$

Solving the augmented matrix we get,

$$\begin{pmatrix}
5 & -6 & -1 \\
-6 & 10 & 4
\end{pmatrix}
\longleftrightarrow
\begin{pmatrix}
R_1 \leftarrow \frac{R_1}{5} \\
-6 & 10 & 4
\end{pmatrix}$$

$$(1.1.2.109)$$

$$\stackrel{R_2 \leftarrow R_2 + 6R_1}{\longleftrightarrow} \begin{pmatrix}
1 & -\frac{6}{5} & -\frac{1}{5} \\
0 & \frac{14}{5} & \frac{14}{5}
\end{pmatrix}$$

$$(1.1.2.110)$$

$$\stackrel{R_2 \leftarrow \frac{5}{14}R_2}{\longleftrightarrow} \begin{pmatrix}
1 & -\frac{6}{5} & -\frac{1}{5} \\
0 & 1 & 1
\end{pmatrix}$$

$$(1.1.2.111)$$

$$\stackrel{R_1 \leftarrow R_1 + \frac{6}{5}R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 1 \\
0 & 1 & 1
\end{pmatrix}$$

$$(1.1.2.112)$$

$$\Longrightarrow \mathbf{x} = \begin{pmatrix}
1 \\
1
\end{pmatrix}$$

$$(1.1.2.113)$$

Hence from equations (1.1.2.105) and (1.1.2.113) we conclude that the solution is verified.

c) Find the foot of the perpendicular from  $\begin{pmatrix} -3 \\ 1 \\ 3 \end{pmatrix}$  on the plane  $\begin{pmatrix} 2 & -3 & 1 \end{pmatrix} \mathbf{x} = 0$ Solution: Let orthogonal vectors be  $\mathbf{m_1}$  and  $\mathbf{m_2}$  to the given normal vector  $\mathbf{n}$ . Let,  $\mathbf{m} = \begin{pmatrix} a \\ b \end{pmatrix}$ , then

$$\mathbf{m}^{\mathbf{T}}\mathbf{n} = 0 \qquad (1.1.2.114)$$

$$(a \ b \ c)\begin{pmatrix} 2\\ -3\\ 1 \end{pmatrix} = 0$$
 (1.1.2.115)

$$\implies$$
  $-5a + b + 3c = 0$  (1.1.2.116)

Let a=1 and b=0 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ -2 \end{pmatrix} \tag{1.1.2.117}$$

Let a=0 and b=1 we get,

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix} \tag{1.1.2.118}$$

From (1.1.2.117) and (1.1.2.118),

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} \tag{1.1.2.119}$$

Now solving the equation

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.2.120}$$

Substituting the given point and (1.1.2.119) in (1.1.2.120)

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} \mathbf{x} = \begin{pmatrix} -5 \\ 1 \\ 3 \end{pmatrix} \tag{1.1.2.121}$$

Using the Singular value decomposition to solve (1.1.2.121) as follows,

$$\mathbf{M} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T \tag{1.1.2.122}$$

Where the columns of V are the eigen vectors of  $M^TM$ , the columns of U are the eigen vectors of  $MM^T$  and  $\Sigma$  is diagonal matrix of singular value of eigenvalues of  $M^TM$ .

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix} \tag{1.1.2.123}$$

$$\mathbf{M}\mathbf{M}^T = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \\ -2 & 3 & 13 \end{pmatrix}$$
 (1.1.2.124)

Substituting (1.1.2.122) in (1.1.2.120)

$$\mathbf{U}\mathbf{\Sigma}\mathbf{V}^T\mathbf{x} = \mathbf{b} \tag{1.1.2.125}$$

$$\mathbf{x} = \mathbf{V} \mathbf{\Sigma}^{-1} \mathbf{U}^{\mathbf{T}} \mathbf{b} \tag{1.1.2.126}$$

where  $\Sigma^{-1}$  is Moore-Penrose Pseudo-Inverse of  $\Sigma$ .

Now finding the eigen values of  $MM^T$ 

$$\left|\mathbf{M}\mathbf{M}^{T} - \lambda \mathbf{I}\right| = 0 \tag{1.1.2.127}$$

$$\begin{vmatrix} 1 - \lambda & 0 & -2 \\ 0 & 1 - \lambda & 3 \\ -2 & 3 & 13 - \lambda \end{vmatrix} = 0 \quad (1.1.2.128)$$

$$\implies \lambda^3 - 15\lambda^2 + 14\lambda = 0 \qquad (1.1.2.129)$$

Hence eigen values of  $\mathbf{M}\mathbf{M}^T$ ,

$$\lambda_1 = 1$$
  $\lambda_2 = 14$   $\lambda_3 = 0$  (1.1.2.130)

Therefore eigen vectors of  $\mathbf{M}\mathbf{M}^T$ ,

$$\mathbf{u_1} = \begin{pmatrix} \frac{3}{2} \\ 1 \\ 0 \end{pmatrix} \quad \mathbf{u_2} = \begin{pmatrix} \frac{-2}{13} \\ \frac{3}{13} \\ 1 \end{pmatrix} \quad \mathbf{u_3} = \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix}$$
(1.1.2.131)

Normalizing the eigen vectors,

$$\mathbf{u_1} = \begin{pmatrix} \frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \\ 0 \end{pmatrix} \quad \mathbf{u_2} = \begin{pmatrix} \frac{-2}{\sqrt{182}} \\ \frac{3}{\sqrt{182}} \\ \frac{13}{\sqrt{182}} \end{pmatrix} \quad \mathbf{u_3} = \begin{pmatrix} \frac{2}{\sqrt{14}} \\ \frac{-3}{\sqrt{14}} \\ \frac{1}{\sqrt{14}} \end{pmatrix}$$

$$(1.1.2.132)$$

Hence from the above we get,

$$\mathbf{U} = \begin{pmatrix} \frac{3}{\sqrt{13}} & \frac{-2}{\sqrt{182}} & \frac{2}{\sqrt{14}} \\ \frac{2}{\sqrt{13}} & \frac{3}{\sqrt{182}} & \frac{-3}{\sqrt{14}} \\ 0 & \frac{13}{\sqrt{182}} & \frac{1}{\sqrt{14}} \end{pmatrix}$$
(1.1.2.133)

By computing the singular values from eigen values  $\lambda_1, \lambda_2, \lambda_3$  we get  $\Sigma$  as,

$$\Sigma = \begin{pmatrix} 1 & 0 \\ 0 & 14 \\ 0 & 0 \end{pmatrix} \tag{1.1.2.134}$$

Now calculating eigen values of  $\mathbf{M}^T \mathbf{M}$ 

$$\left| \mathbf{M}^T \mathbf{M} - \lambda I \right| = 0 \qquad (1.1.2.135)$$

$$\begin{vmatrix} 5 - \lambda & -6 \\ -6 & 10 - \lambda \end{vmatrix} = 0 \qquad (1.1.2.136)$$

$$\implies \lambda^2 - 15\lambda + 14 = 0 \qquad (1.1.2.137)$$

hence the eigen values of  $\mathbf{M}^T \mathbf{M}$ 

$$\lambda_1 = 1 \quad \lambda_2 = 14 \quad (1.1.2.138)$$

Therefore eigen vectors  $\mathbf{M}^T \mathbf{M}$  are,

$$\mathbf{v_1} = \begin{pmatrix} \frac{3}{2} \\ 1 \end{pmatrix} \quad \mathbf{v_2} = \begin{pmatrix} \frac{-2}{3} \\ 1 \end{pmatrix}$$
 (1.1.2.139)

Normalizing the eigen vectors,

$$\mathbf{v_1} = \begin{pmatrix} \frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \end{pmatrix} \quad \mathbf{v_2} = \begin{pmatrix} \frac{-2}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} \end{pmatrix} \quad (1.1.2.140)$$

Hence V is given as,

$$\mathbf{V} = \begin{pmatrix} \frac{3}{\sqrt{13}} & \frac{-2}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} & \frac{3}{\sqrt{13}} \end{pmatrix}$$
 (1.1.2.141)

Moore Pseudo inverse of  $\Sigma$  is,

$$\Sigma^{-1} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{\sqrt{14}} & 0 \end{pmatrix}$$
 (1.1.2.142)

Substituting (1.1.2.133), (1.1.2.141) and (1.1.2.142) in (1.1.2.126),

$$\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} & 0\\ \frac{-2}{\sqrt{182}} & \frac{3}{\sqrt{182}} & \frac{13}{\sqrt{182}} \\ \frac{2}{\sqrt{14}} & \frac{-3}{\sqrt{14}} & \frac{1}{\sqrt{14}} \end{pmatrix} \begin{pmatrix} -5\\1\\3 \end{pmatrix} = \begin{pmatrix} \frac{-13}{\sqrt{13}}\\ \frac{52}{\sqrt{182}}\\ \frac{-10}{\sqrt{1}} \end{pmatrix}$$

$$(1.1.2.143)$$

$$\mathbf{\Sigma}^{-1}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{\sqrt{14}} & 0 \end{pmatrix} \begin{pmatrix} \frac{-13}{\sqrt{13}} \\ \frac{52}{\sqrt{182}} \\ \frac{-10}{\sqrt{14}} \end{pmatrix} = \begin{pmatrix} \frac{-13}{\sqrt{13}} \\ \frac{26}{7\sqrt{13}} \end{pmatrix}$$
(1.1.2.144)

$$\mathbf{V}\mathbf{\Sigma}^{-1}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{3}{\sqrt{13}} & \frac{-2}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} & \frac{3}{\sqrt{13}} \end{pmatrix} \begin{pmatrix} \frac{-13}{\sqrt{13}} \\ \frac{26}{7\sqrt{13}} \end{pmatrix} = \begin{pmatrix} \frac{-25}{7} \\ \frac{-8}{7} \end{pmatrix}$$
(1.1.2.145)

$$\implies \mathbf{x} = \begin{pmatrix} \frac{-25}{7} \\ \frac{-8}{7} \end{pmatrix}$$

$$(1.1.2.146)$$

Now verifying (1.1.2.146) using (1.1.2.120)

$$\mathbf{M}\mathbf{x} = \mathbf{b} \implies \mathbf{M}^T \mathbf{M}\mathbf{x} = \mathbf{M}^T \mathbf{b}$$
 (1.1.2.147)

Substituting (1.1.2.119), (1.1.2.123) and given point in (1.1.2.147)

$$\begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix} \mathbf{x} = \begin{pmatrix} -11 \\ 10 \end{pmatrix}$$
 (1.1.2.148) (1.1.2.149)

Solving the augmented matrix.

$$\begin{pmatrix}
5 & -6 & -11 \\
-6 & 10 & 10
\end{pmatrix}
\xrightarrow{R_1 = \frac{R_1}{5}}
\begin{pmatrix}
1 & \frac{-6}{5} & \frac{-11}{5} \\
-6 & 10 & 10
\end{pmatrix}$$

$$(1.1.2.150)$$

$$\stackrel{R_2 = R_2 + 6R_1}{\longleftrightarrow} \begin{pmatrix}
1 & \frac{-6}{5} & \frac{-11}{5} \\
0 & \frac{14}{5} & \frac{-16}{5} \\
(1.1.2.151)
\end{pmatrix}$$

$$\xrightarrow{R_2 = \frac{5R_2}{14}} \begin{pmatrix}
1 & \frac{-6}{5} & \frac{-11}{5} \\
0 & 1 & \frac{-8}{7}
\end{pmatrix}$$

$$(1.1.2.152)$$

$$\stackrel{R_1 = R_1 + \frac{6R_2}{5}}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & \frac{-25}{7} \\
0 & 1 & \frac{-8}{7}
\end{pmatrix}$$

$$(1.1.2.153)$$

From (1.1.2.153) we get,

$$\mathbf{x} = \begin{pmatrix} \frac{-25}{7} \\ \frac{-8}{7} \end{pmatrix} \tag{1.1.2.154}$$

Hence from (1.1.2.146) and (1.1.2.154) the  $\mathbf{x}$  is verified

d) Find the coordinates of foot of perpendicular

from 
$$\mathbf{D} = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}$$
 to the plane
$$2x - 3y + z = 0 \qquad (1.1.2.155)$$

**Solution:** First we find orthogonal vectors  $\mathbf{m_1}$  and  $\mathbf{m_2}$  to the given plane  $\mathbf{n}$ . Let,  $\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$ , then

$$\mathbf{m}^{\mathbf{T}}\mathbf{n} = 0$$

$$\implies (a \ b \ c) \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix} = 0$$

$$\implies 2a - 3b + c = 0 \qquad (1.1.2.156)$$

By substituting a = 1; b = 0 in (1.1.2.156),

$$\mathbf{m_1} = \begin{pmatrix} 1\\0\\-2 \end{pmatrix} \tag{1.1.2.157}$$

By substituting a = 0; b = 1 in (1.1.2.156),

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix} \tag{1.1.2.158}$$

Now M can be written as,

$$\mathbf{M} = \begin{pmatrix} \mathbf{m_1} & \mathbf{m_2} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} \quad (1.1.2.159)$$

such that solving Mx = b gives the required solution.

$$\implies \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -2 & 3 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} \qquad (1.1.2.160)$$

Applying Singular Value Decomposition on M,

$$\mathbf{M} = \mathbf{USV}^T \tag{1.1.2.161}$$

Where the columns of V are the eigenvectors of  $M^TM$ , the columns of U are the eigenvectors of  $MM^T$  and S is diagonal matrix of singular values of  $M^TM$ .

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix} \tag{1.1.2.162}$$

$$\mathbf{M}\mathbf{M}^T = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 3 \\ -2 & 3 & 13 \end{pmatrix}$$
 (1.1.2.163)

From (1.1.2.160) and (1.1.2.161),

$$\mathbf{USV}^{T}\mathbf{x} = \mathbf{b}$$

$$\implies \mathbf{x} = \mathbf{VS}_{+}\mathbf{U}^{T}\mathbf{b} \qquad (1.1.2.164)$$

Where  $S_+$  is Moore-Penrose Pseudo-Inverse of S. Calculating eigenvalues of  $MM^T$ ,

$$\begin{vmatrix} \mathbf{M}\mathbf{M}^T - \lambda \mathbf{I} | = 0 \\ \Rightarrow \begin{vmatrix} 1 - \lambda & 0 & -2 \\ 0 & 1 - \lambda & 3 \\ -2 & 3 & 13 - \lambda \end{vmatrix} = 0$$
$$\Rightarrow \lambda^3 + 15\lambda^2 - 14\lambda = 0$$

Hence eigenvalues of  $\mathbf{M}\mathbf{M}^T$  are,

$$\lambda_1 = 14; \quad \lambda_2 = 1; \quad \lambda_3 = 0 \quad (1.1.2.165)$$

And the corresponding eigenvectors are,

$$\mathbf{u_1} = \begin{pmatrix} \frac{-2}{13} \\ \frac{3}{13} \\ 1 \end{pmatrix}; \quad \mathbf{u_2} = \begin{pmatrix} \frac{3}{2} \\ 1 \\ 0 \end{pmatrix}; \quad \mathbf{u_3} = \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix}$$
(1.1.2.166)

Normalizing the above eigenvectors,

$$\mathbf{u_1} = \begin{pmatrix} \frac{-2}{\sqrt{182}} \\ \frac{3}{\sqrt{182}} \\ \frac{13}{\sqrt{182}} \end{pmatrix}; \quad \mathbf{u_2} = \begin{pmatrix} \frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \\ 0 \end{pmatrix}; \quad \mathbf{u_3} = \begin{pmatrix} \frac{2}{\sqrt{14}} \\ \frac{-3}{\sqrt{14}} \\ \frac{1}{\sqrt{14}} \end{pmatrix}$$
(1.1.2.167)

From (1.1.2.167) we obtain U as,

$$\mathbf{U} = \begin{pmatrix} \frac{-2}{\sqrt{182}} & \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{14}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & \frac{-3}{\sqrt{14}} \\ \frac{13}{\sqrt{182}} & 0 & \frac{1}{\sqrt{14}} \end{pmatrix}$$
 (1.1.2.168)

Using values from (1.1.2.165),

$$\mathbf{S} = \begin{pmatrix} \sqrt{14} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \tag{1.1.2.169}$$

Calculating the eigenvalues of  $\mathbf{M}^T\mathbf{M}$ ,

$$\begin{vmatrix} \mathbf{M}^T \mathbf{M} - \lambda \mathbf{I} | = 0 \\ \implies \begin{vmatrix} 5 - \lambda & -6 \\ -6 & 10 - \lambda \end{vmatrix} = 0 \\ \implies \lambda^2 - 15\lambda + 14 = 0$$

Hence, eigenvalues of  $\mathbf{M}^T \mathbf{M}$  are,

$$\lambda_4 = 14; \quad \lambda_5 = 1$$

And the corresponding eigenvectors are,

$$\mathbf{v}_1 = \begin{pmatrix} \frac{-2}{3} \\ 1 \end{pmatrix}; \quad \mathbf{v}_2 = \begin{pmatrix} \frac{3}{2} \\ 1 \end{pmatrix}$$

Normalizing the above eigenvectors,

$$\mathbf{v}_1 = \begin{pmatrix} \frac{-2}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} \end{pmatrix}; \quad \mathbf{v}_2 = \begin{pmatrix} \frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \end{pmatrix} \quad (1.1.2.170)$$

From(1.1.2.170) we obtain  $\mathbf{V}$  as,

$$\mathbf{V} = \begin{pmatrix} \frac{-2}{\sqrt{13}} & \frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}$$
 (1.1.2.171)

From (1.1.2.161) we get the Singular Value Decomposition of **M**,

$$\mathbf{M} = \begin{pmatrix} \frac{-2}{\sqrt{182}} & \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \\ \frac{3}{\sqrt{182}} & \frac{2}{\sqrt{13}} & \frac{-3}{\sqrt{14}} \\ \frac{13}{\sqrt{182}} & 0 & \frac{1}{\sqrt{14}} \end{pmatrix} \begin{pmatrix} \sqrt{14} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{-2}{\sqrt{13}} & \frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}^{T}$$

$$(1.1.2.172)$$

Moore-Penrose Pseudo inverse of S is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{1}{\sqrt{14}} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.2.173}$$

From (1.1.2.164),

$$\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{12\sqrt{2}}{\sqrt{91}} \\ \frac{3}{\sqrt{13}} \\ \frac{2\sqrt{2}}{7} \end{pmatrix}$$

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{12}{7\sqrt{13}} \\ \frac{3}{\sqrt{13}} \end{pmatrix}$$

$$\mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{3}{7} \\ \frac{6}{7} \end{pmatrix}$$
 (1.1.2.174)

To verify the solution obtained from (1.1.2.174),

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.2.175}$$

Substituting the values from (1.1.2.162) in (1.1.2.175),

$$\begin{pmatrix} 5 & -6 \\ -6 & 10 \end{pmatrix} \mathbf{x} = \begin{pmatrix} -3 \\ 6 \end{pmatrix}$$

Converting the above equation into augmented form and solving for  $\mathbf{x}$ ,

$$\begin{pmatrix} 5 & -6 & -3 \\ -6 & 10 & 6 \end{pmatrix} \xrightarrow{R_2 \leftarrow \frac{5R_2 + 6R_1}{14}} \begin{pmatrix} 5 & -6 & -3 \\ 0 & 1 & \frac{6}{7} \end{pmatrix}$$

$$\xrightarrow{R_1 \leftarrow \frac{R_1 + 6R_2}{5}} \begin{pmatrix} 1 & 0 & \frac{3}{7} \\ 0 & 1 & \frac{6}{7} \end{pmatrix}$$

$$(1.1.2.176)$$

From (1.1.2.176) it can be observed that,

$$\mathbf{x} = \begin{pmatrix} \frac{3}{7} \\ \frac{6}{7} \end{pmatrix} \tag{1.1.2.177}$$

1.1.3. a) Find the foot of the perpendicular to the plane

$$2x + 3y - 2z + 4 = 0 (1.1.3.1)$$

from the point  $\begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix}$  using SVD. **Solution:** 

The given plane equation is

$$(2 \ 3 \ -2)\mathbf{x} = 0 \tag{1.1.3.2}$$

(1.1.3.3)

The equation of plane is

$$\mathbf{n}^T \mathbf{x} = c \tag{1.1.3.4}$$

Hence the normal vector  $\mathbf{n}$  is,

$$\mathbf{n} = \begin{pmatrix} 2\\3\\-2 \end{pmatrix} \tag{1.1.3.5}$$

Let, the normal vectors  $\mathbf{m_1}$  and  $\mathbf{m_2}$  to the normal vector  $\mathbf{n}$  be,

$$\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \tag{1.1.3.6}$$

then, 
$$\mathbf{m}^T \mathbf{n} = 0$$
 (1.1.3.7)

$$\implies \begin{pmatrix} a & b & c \end{pmatrix} \begin{pmatrix} 2 \\ 3 \\ -2 \end{pmatrix} = 0 \qquad (1.1.3.8)$$

Let, a=1 and b=0 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \tag{1.1.3.9}$$

Let, a=0 and b=1,

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ \frac{3}{2} \end{pmatrix} \tag{1.1.3.10}$$

Now solving the equation,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.3.11}$$

Where,

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & \frac{3}{2} \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix}$$
 (1.1.3.12)

To solve (1.1.3.11) we perform singular value decomposition on M given by,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.1.3.13}$$

substituting the value of M from equation (1.1.3.13) to (1.1.3.11),

$$\implies$$
 **USV**<sup>T</sup>**x** = **b** (1.1.3.14)

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} \tag{1.1.3.15}$$

where,  $S_+$  is Moore-Pen-rose Pseudo-Inverse of S.

Columns of  $\mathbf{U}$  are eigenvectors of  $\mathbf{M}\mathbf{M}^T$ , columns of  $\mathbf{V}$  are eigenvectors of  $\mathbf{M}^T\mathbf{M}$  and  $\mathbf{S}$  is diagonal matrix of singular value of eigenvalues of  $\mathbf{M}^T\mathbf{M}$ .

First calculating the eigenvectors corresponding to  $\mathbf{M}^T \mathbf{M}$ .

$$\mathbf{M}^{T}\mathbf{M} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & \frac{3}{2} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & \frac{3}{2} \end{pmatrix} = \begin{pmatrix} 2 & \frac{3}{2} \\ \frac{3}{2} & \frac{13}{4} \end{pmatrix}$$
(1.1.3.16)

Eigenvalues corresponding to  $\mathbf{M}^T \mathbf{M}$  is,

$$\left|\mathbf{M}^T \mathbf{M} - \lambda \mathbf{I}\right| = 0 \qquad (1.1.3.17)$$

$$\Longrightarrow \begin{pmatrix} 2 - \lambda & \frac{3}{2} \\ \frac{3}{2} & \frac{13}{4} - \lambda \end{pmatrix} \qquad (1.1.3.18)$$

$$\implies (\lambda - \frac{17}{4})(\lambda - 1) = 0$$
 (1.1.3.19)

$$\therefore \lambda_1 = \frac{17}{4}, \lambda_2 = 1, \qquad (1.1.3.20)$$

Hence the eigenvectors corresponding to  $\lambda_1$  and  $\lambda_2$  respectively is,

$$\mathbf{v_1} = \begin{pmatrix} \frac{2}{3} \\ 1 \end{pmatrix}, \mathbf{v_2} = \begin{pmatrix} \frac{-3}{2} \\ 1 \end{pmatrix}$$
 (1.1.3.21)

Normalizing the eigenvectors we get,

$$\mathbf{v_1} = \frac{1}{\sqrt{13}} \begin{pmatrix} 2\\3 \end{pmatrix} \qquad (1.1.3.22)$$

$$\mathbf{v_2} = \frac{1}{\sqrt{13}} \begin{pmatrix} -3\\2 \end{pmatrix} \qquad (1.1.3.23)$$

$$\implies \mathbf{V} = \frac{1}{\sqrt{13}} \begin{pmatrix} 2 & -3 \\ 3 & 2 \end{pmatrix} \qquad (1.1.3.24)$$

Now calculating the eigenvectors corresponding to  $\mathbf{M}\mathbf{M}^T$ 

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & \frac{3}{2} \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & \frac{3}{2} \end{pmatrix}$$
 (1.1.3.25)

$$\implies \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & \frac{3}{2} \\ 1 & \frac{3}{2} & \frac{13}{4} \end{pmatrix} \qquad (1.1.3.26)$$

Eigenvalues corresponding to  $MM^T$  is,

$$\left|\mathbf{M}\mathbf{M}^{T} - \lambda \mathbf{I}\right| = 0 \quad (1.1.3.27)$$

$$\implies \begin{pmatrix} 1 - \lambda & 0 & 1\\ 0 & 1 - \lambda & \frac{3}{2}\\ 1 & \frac{3}{2} & \frac{13}{4} - \lambda \end{pmatrix} \quad (1.1.3.28)$$

$$\implies \lambda(\lambda - 1)(\lambda - \frac{17}{4}) = 0 \quad (1.1.3.29)$$

$$\therefore \lambda_3 = \frac{17}{4}, \lambda_4 = 1, \lambda_5 = 0 \quad (1.1.3.30)$$

Hence the eigenvectors corresponding to  $\lambda_3$ ,  $\lambda_4$  and  $\lambda_5$  respectively is,

$$\mathbf{v_3} = \begin{pmatrix} \frac{4}{13} \\ \frac{6}{13} \\ 1 \end{pmatrix}, \mathbf{v_4} = \begin{pmatrix} -\frac{3}{2} \\ 1 \\ 0 \end{pmatrix}, \mathbf{v_5} = \begin{pmatrix} -1 \\ \frac{-3}{2} \\ 1 \end{pmatrix}$$
 (1.1.3.31)

Normalizing the eigenvectors we get,

$$\mathbf{v_3} = \begin{pmatrix} \frac{4}{\sqrt{221}} \\ \frac{6}{\sqrt{221}} \\ \frac{13}{\sqrt{221}} \end{pmatrix} \quad (1.1.3.32)$$

$$\mathbf{v_4} = \begin{pmatrix} \frac{-3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \\ 0 \end{pmatrix} \quad (1.1.3.33)$$

$$\mathbf{v_5} = \begin{pmatrix} \frac{-2}{\sqrt{17}} \\ \frac{-3}{\sqrt{17}} \\ \frac{2}{\sqrt{17}} \end{pmatrix} \quad (1.1.3.34)$$

$$\implies \mathbf{U} = \begin{pmatrix} \frac{4}{\sqrt{221}} & \frac{-3}{\sqrt{13}} & \frac{-2}{\sqrt{17}} \\ \frac{6}{\sqrt{221}} & \frac{2}{\sqrt{13}} & \frac{-3}{\sqrt{17}} \\ \frac{13}{\sqrt{221}} & 0 & \frac{2}{\sqrt{17}} \end{pmatrix}$$
(1.1.3.35)

Now **S** corresponding to eigenvalues  $\lambda_3$ ,  $\lambda_4$  and  $\lambda_5$  is as follows,

$$\mathbf{S} = \begin{pmatrix} \sqrt{\frac{17}{4}} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.1.3.36}$$

Now, Moore-Penrose Pseudo inverse of S is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{2}{\sqrt{17}} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.3.37}$$

Hence we get singular value decomposition of  $\mathbf{M}$  as,

$$\mathbf{M} = \frac{1}{\sqrt{13}} \begin{pmatrix} \frac{4}{\sqrt{221}} & \frac{-3}{\sqrt{13}} & \frac{-2}{\sqrt{17}} \\ \frac{6}{\sqrt{221}} & \frac{2}{\sqrt{13}} & \frac{-3}{\sqrt{17}} \\ \frac{13}{\sqrt{221}} & 0 & \frac{2}{\sqrt{17}} \end{pmatrix} \begin{pmatrix} \sqrt{\frac{17}{4}} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 2 & -3 \\ 3 & 2 \end{pmatrix}^{T}$$

$$(1.1.3.38)$$

Now substituting the values of (1.1.3.24), (1.1.3.37), (1.1.3.35) and (1.1.3.12) in (1.1.3.15),

$$\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{4}{\sqrt{221}} & \frac{-3}{\sqrt{13}} & \frac{-2}{\sqrt{17}} \\ \frac{6}{\sqrt{221}} & \frac{2}{\sqrt{13}} & \frac{-3}{\sqrt{17}} \\ \frac{13}{\sqrt{221}} & 0 & \frac{2}{\sqrt{17}} \end{pmatrix}^{T} \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix}$$
 (1.1.3.39)

$$\implies \mathbf{U}^T \mathbf{b} = \begin{pmatrix} 0 \\ -\sqrt{13} \\ 0 \end{pmatrix} \quad (1.1.3.40)$$

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{2}{\sqrt{17}} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ -\sqrt{13} \\ 0 \end{pmatrix} \quad (1.1.3.41)$$
$$\implies \mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 0 \\ -\sqrt{13} \end{pmatrix} \quad (1.1.3.42)$$

$$\mathbf{VS}_{+}\mathbf{U}^{T}\mathbf{b} = \frac{1}{\sqrt{13}} \begin{pmatrix} 2 & -3 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} 0 \\ -\sqrt{13} \end{pmatrix}$$

$$(1.1.3.43)$$

$$\implies \mathbf{VS}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 3 \\ -2 \end{pmatrix}$$

$$(1.1.3.44)$$

 $\therefore$  from equation (1.1.3.15),

$$\mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix} \tag{1.1.3.45}$$

Verifying the solution using,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.3.46}$$

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

$$(1.1.3.47)$$

$$\implies \begin{pmatrix} 2 & \frac{3}{2} \\ \frac{3}{2} & \frac{13}{4} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix}$$

$$(1.1.3.48)$$

Solving the augmented matrix we get,

$$\begin{pmatrix} 2 & \frac{3}{2} & 3 \\ \frac{3}{2} & \frac{13}{4} & -2 \end{pmatrix} \stackrel{R_1 \leftarrow \frac{R_1}{2}}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{3}{4} & \frac{3}{2} \\ \frac{3}{2} & \frac{13}{4} & -2 \end{pmatrix}$$

$$(1.1.3.49)$$

$$\stackrel{R_2 \leftarrow R_2 - \frac{3}{2}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{3}{4} & \frac{3}{2} \\ 0 & \frac{17}{8} & -\frac{17}{4} \end{pmatrix}$$

$$(1.1.3.50)$$

$$\stackrel{R_2 \leftarrow \frac{8}{17}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{3}{4} & \frac{3}{2} \\ 0 & 1 & -2 \end{pmatrix}$$

$$(1.1.3.51)$$

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

$$(1.1.3.52)$$

$$\Longrightarrow \mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix}$$

$$(1.1.3.53)$$

from equations (1.1.3.45) (1.1.3.53) we conclude that the solution is verified.

a) Determine the distance from the Y-axis to the plane 5x - 2z - 3 = 0

Solution: Equation of plane can be expressed as

$$\mathbf{n}^T \mathbf{x} = c \tag{1.1.3.54}$$

Rewriting given equation of plane in (1.1.3.54) form

$$(5 \ 0 \ -2)\begin{pmatrix} x \\ y \\ z \end{pmatrix} = 3$$
 (1.1.3.55)

where : 
$$\mathbf{n} = \begin{pmatrix} 5 \\ 0 \\ -2 \end{pmatrix}$$
,  $\mathbf{x} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$  and  $c = 3$ 

We need to represent equation of plane in parametric form,

$$\mathbf{x} = \mathbf{p} + \lambda_1 \mathbf{q} + \lambda_2 \mathbf{r} \tag{1.1.3.56}$$

Here p is any point on plane and  $\mathbf{q}$ ,  $\mathbf{r}$  are two vectors parallel to plane and hence  $\perp$  to **n**. Find two vectors that are  $\perp$  to **n** 

$$(5 \quad 0 \quad -2) \begin{pmatrix} a \\ b \\ c \end{pmatrix} = 0$$
 (1.1.3.57)

Put a = 0 and b = 1 in (1.1.3.56),  $\implies c = 0$ Put a = 1 and b = 0 in (1.1.3.56),  $\implies c = \frac{5}{2}$ 

Hence 
$$\mathbf{q} = \begin{pmatrix} 1 \\ 0 \\ \frac{5}{2} \end{pmatrix}$$
,  $\mathbf{r} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$   
Let us find point  $\mathbf{p}$  on the plane. Put  $x = \mathbf{p}$ 

1, 
$$y = 0$$
 in (1.1.3.55), we get  $\mathbf{p} = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ 

Since given plane is parallel to y-axis, we can use any point P on y-axis to compute shortest distance.

$$\mathbf{P} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \tag{1.1.3.58}$$

Let **Q** be the point on plane with shortest distance to **P**. **Q** can be expressed in (1.1.3.57) form as

$$\mathbf{Q} = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + \lambda_1 \begin{pmatrix} 1 \\ 0 \\ \frac{5}{2} \end{pmatrix} + \lambda_2 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$
 (1.1.3.59)

Equation **P** and **Q**, and computing pseudo inverse using SVD should give the value of  $\lambda_1$  and  $\lambda_2$  (since plane and y-axis never intersect pseudo inverse should give the points which are closest)

$$\begin{pmatrix}
1 \\ 0 \\ 1
\end{pmatrix} + \lambda_1 \begin{pmatrix}
1 \\ 0 \\ \frac{5}{2}
\end{pmatrix} + \lambda_2 \begin{pmatrix}
0 \\ 1 \\ 0
\end{pmatrix} = \begin{pmatrix}
0 \\ 0 \\ 0
\end{pmatrix}$$

$$\lambda_1 \begin{pmatrix}
1 \\ 0 \\ \frac{5}{2}
\end{pmatrix} + \lambda_2 \begin{pmatrix}
0 \\ 1 \\ 0
\end{pmatrix} = \begin{pmatrix}
-1 \\ 0 \\ -1
\end{pmatrix}$$

$$\begin{pmatrix}
1 & 0 \\ 0 & 1 \\ \frac{5}{2} & 0
\end{pmatrix} \begin{pmatrix}
\lambda_1 \\ \lambda_2
\end{pmatrix} = \begin{pmatrix}
-1 \\ 0 \\ -1
\end{pmatrix}$$

$$(1.1.3.62)$$

$$\mathbf{M}\mathbf{x} = \mathbf{b}$$

$$\mathbf{1}.1.3.63)$$

$$\mathbf{x} = \mathbf{M}^+ \mathbf{b}$$

$$(1.1.3.64)$$

where 
$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{5}{2} & 0 \end{pmatrix}$$
,  $\mathbf{x} = \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix}$  and  $\mathbf{b} = \begin{pmatrix} -1 \\ 0 \\ -1 \end{pmatrix}$ 

Diagonalize  $\mathbf{M}\mathbf{M}^T$ 

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{5}{2} & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & \frac{5}{2} \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 & \frac{3}{2} \\ 0 & 1 & 0 \\ \frac{5}{2} & 0 & \frac{25}{4} \end{pmatrix}$$

$$(1.1.3.65)$$

$$= \begin{pmatrix} 0 & \frac{2}{5} & -\frac{5}{2} \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} \frac{29}{4} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ \frac{2}{5} & 0 & 1 \\ -\frac{5}{2} & 0 & 1 \end{pmatrix}$$

$$(1.1.3.66)$$

$$= \mathbf{U}\Sigma^{T}\Sigma\mathbf{U}^{T} \qquad (1.1.3.67)$$

Verify (1.1.3.66) from,

codes/diagonalize1.py

Diagonalize  $\mathbf{M}^T \mathbf{M}$ 

$$\mathbf{M}^{T}\mathbf{M} = \begin{pmatrix} 1 & 0 & \frac{5}{2} \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{5}{2} & 0 \end{pmatrix} = \begin{pmatrix} \frac{29}{4} & 0 \\ 0 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & \frac{29}{4} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad (1.1.3.69)$$

$$= \mathbf{V}\Sigma^{T}\Sigma\mathbf{V}^{T} \qquad (1.1.3.70)$$

Verify (1.1.3.69) from,

codes/diagonalize2.py

Compute SVD of  $\mathbf{M}$  from (1.1.3.66) and (1.1.3.71),

$$\mathbf{M} = \mathbf{U}\Sigma\mathbf{V}^{T} \qquad (1.1.3.71)$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{5}{2} & 0 \end{pmatrix} = \begin{pmatrix} 0 & \frac{2}{5} & -\frac{5}{2} \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} \frac{\sqrt{29}}{2} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$(1.1.3.72)$$

$$\mathbf{M}^{+} = \mathbf{V}\Sigma^{T}\mathbf{U}^{T} \qquad (1.1.3.73)$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \frac{\sqrt{29}}{2} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ \frac{2}{5} & 0 & 1 \\ -\frac{5}{2} & 0 & 1 \end{pmatrix}$$

$$(1.1.3.74)$$

$$= \begin{pmatrix} \frac{4}{29} & 0 & \frac{10}{29} \\ 0 & 1 & 0 \end{pmatrix} \qquad (1.1.3.75)$$

Verify (1.1.3.75) from,

codes/pseudo inverse.py

Substitute (1.1.3.75) in (1.1.3.64),

$$\mathbf{x} = \begin{pmatrix} \frac{4}{29} & 0 & \frac{10}{29} \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} -1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} -\frac{14}{29} \\ 0 \end{pmatrix} = \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix}$$
(1.1.3.76)

Substituting  $\lambda_1$ ,  $\lambda_2$  in (1.1.3.59)

$$\mathbf{Q} = \begin{pmatrix} \frac{15}{29} \\ 0 \\ -\frac{6}{29} \end{pmatrix} \tag{1.1.3.77}$$

Distance between point P and Q is

$$\|\mathbf{P} - \mathbf{Q}\| = \sqrt{\left(\frac{15}{29}\right)^2 + 0 + \left(-\frac{6}{29}\right)^2} = \frac{3}{\sqrt{29}}$$
(1.1.3.78)

Hence, distance from y-axis to 5x-2z-3=0 is  $\frac{3}{\sqrt{29}}$ .

Verifying solution to (1.1.3.63) by least squares method

$$\mathbf{M}^{T}(\mathbf{b} - \mathbf{M}\mathbf{x}) = 0 \tag{1.1.3.79}$$

$$\implies \mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.3.80}$$

Substituting  $\mathbf{M}, \mathbf{b}$  from (1.1.3.62) in

(1.1.3.80)

$$\begin{pmatrix} 1 & 0 & \frac{5}{2} \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{5}{2} & 0 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 1 & 0 & \frac{5}{2} \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} -1 \\ 0 \\ -1 \end{pmatrix}$$

$$(1.1.3.81)$$

$$\begin{pmatrix} \frac{29}{4} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} = \begin{pmatrix} -\frac{7}{2} \\ 0 \end{pmatrix} \qquad (1.1.3.82)$$

$$\implies \frac{29}{4} \lambda_1 = -\frac{7}{2} \qquad (1.1.3.83)$$

$$\lambda_1 = -\frac{7}{2} \times \frac{4}{29} = -\frac{14}{29}$$

$$(1.1.3.84)$$
and  $\lambda_2 = 0 \qquad (1.1.3.85)$ 

$$\mathbf{x} = \begin{pmatrix} -\frac{14}{29} \\ 0 \end{pmatrix} \qquad (1.1.3.86)$$

Comparing (1.1.3.76) and (1.1.3.86) solution is verified.

b) Determine the distance from the Z-axis to the plane 5x - 12y - 8 = 0

**Solution:** Equation of plane can be expressed as

$$\mathbf{n}^T \mathbf{x} = c \tag{1.1.3.87}$$

Rewriting given equation of plane in (1.1.3.87) form

$$(5 -12 \ 0)\begin{pmatrix} x \\ y \\ z \end{pmatrix} = 8$$
 (1.1.3.88)

where the value of

$$\mathbf{n} = \begin{pmatrix} 5 \\ -12 \\ 0 \end{pmatrix} \tag{1.1.3.89}$$

$$\mathbf{x} = \begin{pmatrix} x \\ y \\ z \end{pmatrix} \tag{1.1.3.90}$$

$$c = 8$$
 (1.1.3.91)

We need to represent the equation of plane in parametric form,

$$\mathbf{Q} = \mathbf{p} + \lambda_1 \mathbf{q} + \lambda_2 \mathbf{r} \tag{1.1.3.92}$$

Here p is any point on plane and  $\mathbf{q}$ ,  $\mathbf{r}$  are two vectors parallel to plane and hence  $\perp$  to  $\mathbf{n}$ . Now, we need to find these two vectors  $\mathbf{q}$ 

and  $\mathbf{r}$  which are  $\perp$  to  $\mathbf{n}$ 

$$(5 -12 \ 0) \begin{pmatrix} a \\ b \\ c \end{pmatrix} = 0 \implies 5a - 12b = 0$$
(1.1.3.93)

Put a = 0 and c = 1 in (1.1.3.93),  $\implies b = 0$ Put a = 1 and c = 0 in (1.1.3.93),  $\implies b = \frac{5}{12}$ 

Hence 
$$\mathbf{q} = \begin{pmatrix} 1 \\ \frac{5}{12} \\ 0 \end{pmatrix}, \mathbf{r} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

Let us find point **p** on the plane. Put x =

1, 
$$z = 0$$
 in (1.1.3.88), we get  $\mathbf{p} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ 

Since given plane is parallel to Z-axis, we can use any point *P* on Z-axis to compute shortest distance.

$$\mathbf{P} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \tag{1.1.3.94}$$

Let  $\mathbf{Q}$  be the point on plane with shortest distance to  $\mathbf{P}$ .  $\mathbf{Q}$  can be expressed in (1.1.3.93) form as

$$\mathbf{Q} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + \lambda_1 \begin{pmatrix} \frac{1}{5} \\ \frac{12}{12} \\ 0 \end{pmatrix} + \lambda_2 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$
 (1.1.3.95)

Computation of Pseudo Inverse using SVD in order to determine the value of  $\lambda_1$  and  $\lambda_2$ :

$$\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + \lambda_1 \begin{pmatrix} 1 \\ \frac{5}{12} \\ 0 \end{pmatrix} + \lambda_2 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$
 (1.1.3.96)

$$\lambda_1 \begin{pmatrix} 1 \\ \frac{5}{12} \\ 0 \end{pmatrix} + \lambda_2 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ 0 \end{pmatrix}$$
 (1.1.3.97)

$$\begin{pmatrix} 1 & 0 \\ \frac{5}{12} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ 0 \end{pmatrix} \quad (1.1.3.98)$$

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.3.99}$$

$$\implies$$
  $\mathbf{x} = \mathbf{M}^+ \mathbf{b}$  (1.1.3.100)

where,

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ \frac{5}{12} & 0 \\ 0 & 1 \end{pmatrix} \tag{1.1.3.101}$$

$$\mathbf{x} = \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} \tag{1.1.3.102}$$

$$\mathbf{b} = \begin{pmatrix} -1 \\ -1 \\ 0 \end{pmatrix} \tag{1.1.3.103}$$

Applying Singular Value Decomposition on **M**,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.1.3.104}$$

Where the columns of V are the eigenvectors of  $M^TM$ , the columns of U are the eigenvectors of  $MM^T$  and S is diagonal matrix of Singular values of  $M^TM$ .

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} \frac{169}{144} & 0\\ 0 & 1 \end{pmatrix}$$
 (1.1.3.105)

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & \frac{5}{12} & 0\\ \frac{5}{12} & \frac{25}{144} & 0\\ 0 & 0 & 1 \end{pmatrix}$$
 (1.1.3.106)

As we know that,

$$\mathbf{USV}^{T}\mathbf{x} = \mathbf{b}$$

$$\implies \mathbf{x} = \mathbf{VS}_{+}\mathbf{U}^{T}\mathbf{b} \qquad (1.1.3.107)$$

Where  $S_+$  is Moore-Penrose Pseudo-Inverse of S. Calculating eigenvalues of  $\mathbf{M}\mathbf{M}^T$ ,

$$\begin{vmatrix} \mathbf{M}\mathbf{M}^T - \lambda \mathbf{I} | = 0 \\ \Rightarrow \begin{vmatrix} 1 - \lambda & \frac{5}{12} & 0 \\ \frac{5}{12} & \frac{25}{144} - \lambda & 0 \\ 0 & 0 & 1 - \lambda \end{vmatrix} = 0$$
$$\Rightarrow \lambda^3 - \frac{313}{144}\lambda^2 + \frac{169}{144}\lambda = 0$$

Hence eigenvalues of  $\mathbf{M}\mathbf{M}^T$  are,

$$\lambda_1 = \frac{169}{144}; \quad \lambda_2 = 1; \quad \lambda_3 = 0 \quad (1.1.3.108)$$

And the corresponding eigenvectors are,

$$\mathbf{u_1} = \begin{pmatrix} 1 \\ \frac{5}{12} \\ 0 \end{pmatrix}; \quad \mathbf{u_2} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}; \quad \mathbf{u_3} = \begin{pmatrix} -\frac{5}{12} \\ 1 \\ 0 \end{pmatrix}$$
(1.1.3.109)

$$\mathbf{U} = \begin{pmatrix} 1 & 0 & -\frac{5}{12} \\ \frac{5}{12} & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.3.110}$$

Using values from (1.1.3.108),

$$\mathbf{S} = \begin{pmatrix} \frac{13}{12} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.1.3.111}$$

Calculating the eigenvalues of  $\mathbf{M}^T\mathbf{M}$ ,

$$\begin{vmatrix} \mathbf{M}^T \mathbf{M} - \lambda \mathbf{I} | = 0 \\ \Rightarrow \begin{vmatrix} \frac{169}{144} - \lambda & 0 \\ 0 & 1 - \lambda \end{vmatrix} = 0 \\ \Rightarrow \lambda^2 - \frac{313}{144}\lambda + \frac{169}{144} = 0$$

Hence, eigenvalues of  $\mathbf{M}^T\mathbf{M}$  are,

$$\lambda_4 = \frac{169}{144}; \quad \lambda_5 = 1$$

And the corresponding eigenvectors are,

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}; \quad \mathbf{v}_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \tag{1.1.3.112}$$

From (1.1.3.112) we obtain  $\mathbf{V}$  as,

$$\mathbf{V} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{1.1.3.113}$$

Now, we can compute SVD of M:

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^{T}$$
 (1.1.3.114)  

$$= \begin{pmatrix} 1 & 0 & -\frac{5}{12} \\ \frac{5}{12} & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} \frac{13}{12} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$
 (1.1.3.115)  

$$\mathbf{M}^{+} = \mathbf{V}\mathbf{S}^{T}\mathbf{U}^{T}$$
 (1.1.3.116)  

$$= \begin{pmatrix} \frac{144}{169} & \frac{60}{169} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
 (1.1.3.117)

Substitute (1.1.3.117) in (1.1.3.100),

$$\mathbf{x} = \begin{pmatrix} \frac{144}{169} & \frac{60}{169} & 0\\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} -1\\ -1\\ 0 \end{pmatrix}$$

$$(1.1.3.118)$$

$$\mathbf{x} = \begin{pmatrix} -\frac{204}{169}\\ 0 \end{pmatrix}$$

$$(1.1.3.120)$$

$$\Rightarrow \begin{pmatrix} \lambda_1\\ \lambda_2 \end{pmatrix} = \begin{pmatrix} -\frac{204}{169}\\ 0 \end{pmatrix}$$

$$(1.1.3.120)$$

Substituting  $\lambda_1$ ,  $\lambda_2$  in (1.1.3.95)

$$\mathbf{Q} = \begin{pmatrix} -\frac{204}{169} \\ -\frac{85}{169} \\ 0 \end{pmatrix} \tag{1.1.3.121}$$

Distance between point P and Q is

$$\|\mathbf{P} - \mathbf{Q}\| = \sqrt{\left(-\frac{204}{169}\right)^2 + \left(-\frac{85}{169}\right)^2 + 0}$$
(1.1.3.122)

$$\|\mathbf{P} - \mathbf{Q}\| = \frac{17}{13} \tag{1.1.3.123}$$

Hence, the distance from the Z-axis to the plane 5x - 12y - 8 = 0 is  $\frac{17}{13}$ . Now, we can verify the solution using Least Squares Method,

$$\mathbf{M}^{T}(\mathbf{b} - \mathbf{M}\mathbf{x}) = 0 \tag{1.1.3.124}$$

$$\implies \mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.3.125}$$

Substituting **M**, **b** from (1.1.3.98) in (1.1.3.125)

$$\begin{pmatrix} 1 & 0 \\ \frac{5}{12} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{5}{12} & 0 \\ 0 & 0 & 1 \end{pmatrix} \mathbf{x} = \begin{pmatrix} 1 & \frac{5}{12} & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} -1 \\ -1 \\ 0 \end{pmatrix}$$
(1.1.3.126)

$$\begin{pmatrix} \frac{169}{144} & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} = \begin{pmatrix} -\frac{17}{12} \\ 0 \end{pmatrix} \quad (1.1.3.127)$$

$$\implies \frac{169}{144} \lambda_1 = -\frac{17}{12} \quad (1.1.3.128)$$

$$\lambda_1 = -\frac{17}{12} \times \frac{144}{169} = -\frac{204}{169}$$

$$(1.1.3.129)$$

and 
$$\lambda_2 = 0$$
 (1.1.3.130)

$$\implies \mathbf{x} = \begin{pmatrix} -\frac{204}{169} \\ 0 \end{pmatrix} \quad (1.1.3.131)$$

Comparing (1.1.3.118) and (1.1.3.131) solution is verified.

1.1.4. Find the foot of the perpendicular using svd

drawn from  $\begin{pmatrix} -3\\1\\2 \end{pmatrix}$  to the plane

$$(2 -1 -2)\mathbf{x} + 4 = 0$$
 (1.1.4.1)

Solution: Let us consider orthogonal vectors  $m_1$  and  $m_2$  to the given normal vector n. Let,

$$\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}, \text{ then }$$

$$\mathbf{m}^{\mathbf{T}}\mathbf{n} = 0 \tag{1.1.4.2}$$

$$\implies \begin{pmatrix} a & b & c \end{pmatrix} \begin{pmatrix} 2 \\ -1 \\ -2 \end{pmatrix} = 0 \tag{1.1.4.3}$$

$$\implies 2a - b - 2c = 0 \tag{1.1.4.4}$$

Let a=1 and b=0 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \tag{1.1.4.5}$$

Let a=0 and b=1 we get,

$$\mathbf{m_2} = \begin{pmatrix} 0\\1\\-\frac{1}{2} \end{pmatrix} \tag{1.1.4.6}$$

Let us solve the equation,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.4.7}$$

Substituting (1.1.4.5) and (1.1.4.6) in (1.1.4.7),

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & -\frac{1}{2} \end{pmatrix} \mathbf{x} = \begin{pmatrix} -3 \\ 1 \\ 2 \end{pmatrix}$$
 (1.1.4.8)

To solve (1.1.4.8), we will perform Singular Value Decomposition on  $\mathbf{M}$  as follows,

$$\mathbf{M} = \mathbf{USV}^T \tag{1.1.4.9}$$

Where the columns of V are the eigen vectors of  $M^TM$ , the columns of U are the eigen vectors of  $MM^T$  and S is diagonal matrix of singular value of eigenvalues of  $M^TM$ .

$$\mathbf{M}^{T}\mathbf{M} = \begin{pmatrix} 2 & \frac{-1}{2} \\ \frac{-1}{2} & \frac{5}{4} \end{pmatrix}$$
 (1.1.4.10)

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 & 1\\ 0 & 1 & -\frac{1}{2}\\ 1 & -\frac{1}{2} & \frac{5}{4} \end{pmatrix}$$
 (1.1.4.11)

Substituting (1.1.4.9) in (1.1.4.7),

$$\mathbf{USV}^T \mathbf{x} = \mathbf{b} \tag{1.1.4.12}$$

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{\mathsf{T}}\mathbf{b} \tag{1.1.4.13}$$

Where  $S_+$  is Moore-Penrose Pseudo-Inverse of S.

Let us calculate eigen values of  $\mathbf{M}\mathbf{M}^T$ ,

$$\begin{vmatrix} \mathbf{M}\mathbf{M}^{T} - \lambda \mathbf{I} | = 0 & (1.1.4.14) \\ \Rightarrow \begin{pmatrix} 1 - \lambda & 0 & 1 \\ 0 & 1 - \lambda & -\frac{1}{2} \\ 1 & -\frac{1}{2} & \frac{5}{4} - \lambda \end{pmatrix} = 0 & (1.1.4.15) \\ \Rightarrow \lambda^{3} - \frac{13}{4}\lambda^{2} + \frac{9}{4}\lambda = 0 & (1.1.4.16)$$

From equation (1.1.4.16) eigen values of  $\mathbf{M}\mathbf{M}^T$  are,

$$\lambda_1 = \frac{9}{4}$$
  $\lambda_2 = 1$   $\lambda_3 = 0$  (1.1.4.17)

The eigen vectors of  $\mathbf{M}\mathbf{M}^T$  are,

$$\mathbf{u}_{1} = \begin{pmatrix} -\frac{4}{5} \\ \frac{2}{5} \\ -1 \end{pmatrix} \quad \mathbf{u}_{2} = \begin{pmatrix} -\frac{1}{2} \\ -1 \\ 0 \end{pmatrix} \quad \mathbf{u}_{3} = \begin{pmatrix} -1 \\ \frac{1}{2} \\ 1 \end{pmatrix}$$
(1.1.4.18)

Normalizing the eigen vectors in equation (1.1.4.18)

$$\mathbf{u}_{1} = \begin{pmatrix} -\frac{4}{3\sqrt{5}} \\ \frac{2}{3\sqrt{5}} \\ -\frac{\sqrt{5}}{3} \end{pmatrix} \quad \mathbf{u}_{2} = \begin{pmatrix} -\frac{1}{\sqrt{5}} \\ -\frac{2}{\sqrt{5}} \\ 0 \end{pmatrix} \quad \mathbf{u}_{3} = \begin{pmatrix} -\frac{2}{3} \\ \frac{1}{3} \\ \frac{2}{3} \end{pmatrix}$$

$$(1.1.4.19)$$

Hence we obtain **U** as follows,

$$\mathbf{U} = \begin{pmatrix} -\frac{4}{3\sqrt{5}} & -\frac{1}{\sqrt{5}} & -\frac{2}{3} \\ \frac{2}{3\sqrt{5}} & -\frac{2}{\sqrt{5}} & \frac{1}{3} \\ -\frac{\sqrt{5}}{3} & 0 & \frac{2}{3} \end{pmatrix}$$
(1.1.4.20)

After computing the singular values from eigen values  $\lambda_1, \lambda_2, \lambda_3$  we get **S** as follows,

$$\mathbf{S} = \begin{pmatrix} \frac{3}{2} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.1.4.21}$$

Now, lets calculate eigen values of  $\mathbf{M}^T \mathbf{M}$ ,

$$\left|\mathbf{M}^{T}\mathbf{M} - \lambda \mathbf{I}\right| = 0 \tag{1.1.4.22}$$

$$\implies \begin{pmatrix} 2 - \lambda & -\frac{1}{2} \\ -\frac{1}{2} & \frac{5}{4} - \lambda \end{pmatrix} = 0 \tag{1.1.4.23}$$

$$\implies \lambda^2 - \frac{13}{4}\lambda + \frac{9}{4} = 0 \qquad (1.1.4.24)$$

Hence eigen values of  $\mathbf{M}^T\mathbf{M}$  are,

$$\lambda_1 = \frac{9}{4} \quad \lambda_2 = 1 \tag{1.1.4.25}$$

Hence the eigen vectors of  $\mathbf{M}^T \mathbf{M}$  are,

$$\mathbf{v}_1 = \begin{pmatrix} -2\\1 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} -\frac{1}{2}\\-1 \end{pmatrix} \tag{1.1.4.26}$$

Normalizing the eigen vectors,

$$\mathbf{v}_1 = \begin{pmatrix} -\frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} -\frac{1}{\sqrt{5}} \\ -\frac{2}{\sqrt{5}} \end{pmatrix} \tag{1.1.4.27}$$

Hence we obtain V as,

$$\mathbf{V} = \begin{pmatrix} -\frac{2}{\sqrt{5}} & -\frac{1}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \end{pmatrix}$$
 (1.1.4.28)

From (1.1.4.7), the Singular Value Decomposition of  $\mathbf{M}$  is as follows,

$$\mathbf{M} = \begin{pmatrix} -\frac{4}{3\sqrt{5}} & -\frac{1}{\sqrt{5}} & -\frac{2}{3} \\ \frac{2}{3\sqrt{5}} & -\frac{2}{\sqrt{5}} & \frac{1}{3} \\ -\frac{\sqrt{5}}{3} & 0 & \frac{2}{3} \end{pmatrix} \begin{pmatrix} \frac{3}{2} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} -\frac{2}{\sqrt{5}} & -\frac{1}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} & -\frac{2}{\sqrt{5}} \end{pmatrix}^{T}$$

$$(1.1.4.29)$$

Now, Moore-Penrose Pseudo inverse of S is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{2}{3} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.4.30}$$

From (1.1.4.13) we get,

$$\mathbf{U}^T \mathbf{b} = \begin{pmatrix} \frac{4}{3\sqrt{5}} \\ \frac{1}{\sqrt{5}} \\ \frac{11}{2} \end{pmatrix}$$
 (1.1.4.31)

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{8}{9\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{pmatrix}$$
 (1.1.4.32)

$$\mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} -\frac{5}{9} \\ -\frac{2}{9} \end{pmatrix}$$
(1.1.4.33)

Verifying the solution of (1.1.4.33) using,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.4.34}$$

Evaluating the R.H.S in (1.1.4.34) we get,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \tag{1.1.4.35}$$

$$\implies \begin{pmatrix} 2 & -\frac{1}{2} \\ -\frac{1}{2} & \frac{5}{4} \end{pmatrix} \mathbf{x} = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \tag{1.1.4.36}$$

Solving the augmented matrix of (1.1.4.36) we

get,

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

$$(1.1.4.37)$$

$$\xrightarrow{R_2 = R_2 + \frac{R_1}{2}} \begin{pmatrix}
1 & -\frac{1}{4} & -\frac{1}{2} \\
0 & \frac{9}{8} & -\frac{1}{4}
\end{pmatrix}$$

$$(1.1.4.38)$$

$$\xrightarrow{R_2 = \frac{8}{9}R_2} \begin{pmatrix}
1 & -\frac{1}{4} & -\frac{1}{2} \\
0 & 1 & -\frac{2}{9}
\end{pmatrix}$$

$$(1.1.4.39)$$

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

$$(1.1.4.40)$$

From equation (1.1.4.40), solution is given by,

$$\mathbf{x} = \begin{pmatrix} -\frac{5}{9} \\ -\frac{2}{9} \end{pmatrix} \tag{1.1.4.41}$$

Comparing results of  $\mathbf{x}$  from (1.1.4.33) and (1.1.4.41), we can say that the solution is verified.

1.1.5. Find the foot of the perpendicular to the given plane

$$2x + 3y - 4z + 5 = 0 (1.1.5.1)$$

from

a)

$$\mathbf{B} = \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix} \tag{1.1.5.2}$$

**Solution:** The given equation of plane can be represented as

$$(2 \quad 3 \quad -4)\mathbf{x} = -5 \tag{1.1.5.3}$$

$$\mathbf{n} = \begin{pmatrix} 2\\3\\-4 \end{pmatrix} \tag{1.1.5.4}$$

We need to find two vectors  $\mathbf{m_1}$  and  $\mathbf{m_2}$  that are  $\perp$  to  $\mathbf{n}$ 

$$\implies \left(2 \quad 3 \quad -4\right) \begin{pmatrix} a \\ b \\ c \end{pmatrix} = 0 \qquad (1.1.5.5)$$

Put a = 1 and b = 0 in (1.1.5.5), we get,

$$\mathbf{m_1} = \begin{pmatrix} 1\\0\\\frac{1}{2} \end{pmatrix} \tag{1.1.5.6}$$

Put a = 0 and b = 1 in (1.1.5.5),we get,

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ \frac{3}{4} \end{pmatrix} \tag{1.1.5.7}$$

Now, solving the equation

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.1.5.8}$$

where,

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{1}{2} & \frac{3}{4} \end{pmatrix} \tag{1.1.5.9}$$

$$\mathbf{b} = \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix} \tag{1.1.5.10}$$

Now, to solve equation (1.1.5.8), we perform Singular Value Decomposition on  $\mathbf{M}$  as follows,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.1.5.11}$$

Substituting the value of M from equation (1.1.5.11) to(1.1.5.8),

$$\mathbf{USV}^T \mathbf{x} = \mathbf{b} \tag{1.1.5.12}$$

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} \tag{1.1.5.13}$$

Where,  $S_+$  is the Moore-Pen-rose Pseudo-Inverse of S. Columns of V are the eigen vectors of  $M^TM$ , columns of U are the eigen vectors of  $MM^T$  and S is diagonal matrix of singular value of eigenvalues of  $M^TM$ .

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} \frac{5}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix}$$
 (1.1.5.14)

Eigen values corresponding to  $\mathbf{M}^T \mathbf{M}$  are given by,

$$\left|\mathbf{M}^T \mathbf{M} - \lambda \mathbf{I}\right| = 0 \quad (1.1.5.15)$$

$$\implies \left| \begin{pmatrix} \frac{5}{4} - \lambda & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} - \lambda \end{pmatrix} \right| = 0 \quad (1.1.5.16)$$

$$\implies \lambda^2 - \frac{45}{16}\lambda + \frac{29}{16} = 0 \qquad (1.1.5.17)$$

Hence eigen values of  $\mathbf{M}^T\mathbf{M}$  are,

$$\lambda_1 = \frac{29}{16} \tag{1.1.5.18}$$

$$\lambda_2 = 1 \tag{1.1.5.19}$$

Hence the eigen vectors of  $\mathbf{M}^T \mathbf{M}$  are,

$$\mathbf{v}_1 = \begin{pmatrix} \frac{2}{3} \\ 1 \end{pmatrix} \tag{1.1.5.20}$$

$$\mathbf{v}_2 = \begin{pmatrix} -\frac{3}{2} \\ 1 \end{pmatrix} \tag{1.1.5.21}$$

Normalizing the eigen vectors, we obtain V of (1.1.5.11) as follows,

$$\mathbf{V} = \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}$$
 (1.1.5.22)

S of the diagonal matrix of (1.1.5.11) is:

$$\mathbf{S} = \begin{pmatrix} \frac{\sqrt{29}}{4} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.1.5.23}$$

Now, calculating eigen value of  $\mathbf{M}\mathbf{M}^T$ ,

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} \end{pmatrix}$$
 (1.1.5.24)

Eigen values corresponding to  $\mathbf{M}\mathbf{M}^T$  are given by

$$\left| \mathbf{M} \mathbf{M}^T - \lambda \mathbf{I} \right| = 0$$

(1.1.5.25)

$$\implies \left| \begin{pmatrix} 1 - \lambda & 0 & \frac{1}{2} \\ 0 & 1 - \lambda & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} - \lambda \end{pmatrix} \right| = 0$$
(1.1.5.26)

$$\implies \lambda^3 - \frac{45}{16}\lambda^2 + \frac{29}{16}\lambda = 0$$
(1.1.5.27)

Hence eigen values of  $\mathbf{M}^T \mathbf{M}$  are,

$$\lambda_3 = \frac{29}{16} \tag{1.1.5.28}$$

$$\lambda_4 = 1 \tag{1.1.5.29}$$

$$\lambda_5 = 0 \tag{1.1.5.30}$$

Hence we obtain U of (1.1.5.11) as follows,

$$\mathbf{U} = \begin{pmatrix} \frac{8}{\sqrt{377}} & -\frac{3}{\sqrt{13}} & -\frac{2}{\sqrt{29}} \\ \frac{12}{\sqrt{377}} & \frac{2}{\sqrt{13}} & -\frac{3}{29} \\ \sqrt{\frac{13}{29}} & 0 & \frac{4}{\sqrt{29}} \end{pmatrix}$$
(1.1.5.31)

Finally from (1.1.5.11) we get the Singular Value Decomposition of  $\mathbf{M}$  as follows,

$$\mathbf{M} = \begin{pmatrix} \frac{8}{\sqrt{377}} & -\frac{3}{\sqrt{13}} & -\frac{2}{\sqrt{29}} \\ \frac{12}{\sqrt{377}} & \frac{2}{\sqrt{13}} & -\frac{3}{29} \\ \sqrt{\frac{13}{29}} & 0 & \frac{4}{\sqrt{29}} \end{pmatrix} \begin{pmatrix} \frac{\sqrt{29}}{4} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}^{T}$$

$$(1.1.5.32)$$

Now, Moore-Penrose Pseudo inverse of S is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{\sqrt{29}}{4} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.5.33}$$

Substituting the values of (1.1.5.31),(1.1.5.22),(1.1.5.33) in (1.1.5.13) we get,

$$\mathbf{U}^T \mathbf{b} = \begin{pmatrix} 0 \\ -\sqrt{13} \\ 0 \end{pmatrix} \tag{1.1.5.34}$$

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 0\\ -\sqrt{13} \end{pmatrix} \qquad (1.1.5.35)$$

$$\mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 3\\ -2 \end{pmatrix}$$
 (1.1.5.36)

Verifying the solution of (1.1.5.36) using,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.1.5.37}$$

Evaluating the R.H.S in (1.1.5.37) we get,

$$\mathbf{M}^{T}\mathbf{b} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{3}{4} \end{pmatrix} \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix} = \begin{pmatrix} 3 \\ -2 \end{pmatrix} \quad (1.1.5.38)$$

$$\implies \begin{pmatrix} \frac{5}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix} \qquad (1.1.5.39)$$

The augmented matrix of (1.1.5.39) is,

$$\begin{pmatrix} \frac{5}{4} & \frac{3}{8} & 3\\ \frac{3}{8} & \frac{25}{16} & -2 \end{pmatrix} \tag{1.1.5.40}$$

Solving the augmented matrix into Row re-

duced echelon form of (1.1.5.40) we get,

$$\begin{pmatrix}
\frac{5}{4} & \frac{3}{8} & 3 \\
\frac{3}{8} & \frac{25}{16} & -2
\end{pmatrix}
\xrightarrow{R_1 \leftarrow \frac{4}{5}R_1} \begin{pmatrix}
1 & \frac{3}{10} & \frac{1}{5} \\
\frac{3}{8} & \frac{25}{16} & -2
\end{pmatrix}$$

$$(1.1.5.41)$$

$$\stackrel{R_2 \leftarrow R_2 - \frac{3}{8}R_1}{\longleftrightarrow} \begin{pmatrix}
1 & \frac{3}{10} & \frac{1}{5} \\
0 & \frac{29}{20} & -\frac{29}{10}
\end{pmatrix}$$

$$(1.1.5.42)$$

$$\stackrel{R_2 \leftarrow \frac{20}{29}R_2}{\longleftrightarrow} \begin{pmatrix}
1 & \frac{3}{10} & \frac{1}{5} \\
0 & 1 & -2
\end{pmatrix}$$

$$(1.1.5.43)$$

$$\stackrel{R_1 \leftarrow R_1 - \frac{3}{10}R_2}{\longleftrightarrow} \begin{pmatrix}
1 & 0 & 3 \\
0 & 1 & -2
\end{pmatrix}$$

$$(1.1.5.44)$$

Therefore,

$$\mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix} \tag{1.1.5.45}$$

Comparing results of  $\mathbf{x}$  from (1.1.5.36) and (1.1.5.45) we conclude that the solution is verified.

b)

$$\mathbf{c} = \begin{pmatrix} 3 \\ 3 \\ 5 \end{pmatrix} \tag{1.1.5.46}$$

### **Solution:**

The given equation of plane can be represented as

$$(2 \quad 3 \quad -4) \mathbf{x} = -5$$
 (1.1.5.47)

$$\mathbf{n} = \begin{pmatrix} 2\\3\\-4 \end{pmatrix} \tag{1.1.5.48}$$

We need to find two vectors  $m_1$  and  $m_2$  that are  $\perp$  to n

$$\implies \left(2 \quad 3 \quad -4\right) \begin{pmatrix} a \\ b \\ c \end{pmatrix} = 0 \qquad (1.1.5.49)$$

Put a = 1 and b = 0 in (1.1.5.49), we get,

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ \frac{1}{2} \end{pmatrix} \tag{1.1.5.50}$$

Put a = 0 and b = 1 in (1.1.5.49), we get,

$$\mathbf{m_2} = \begin{pmatrix} 0\\1\\\frac{3}{4} \end{pmatrix} \tag{1.1.5.51}$$

Now, solving the equation

$$\mathbf{M}\mathbf{x} = \mathbf{c} \tag{1.1.5.52}$$

where,

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{1}{2} & \frac{3}{4} \end{pmatrix} \tag{1.1.5.53}$$

$$\mathbf{c} = \begin{pmatrix} 3 \\ 3 \\ 5 \end{pmatrix} \tag{1.1.5.54}$$

Now, to solve equation (1.1.5.52), we perform Singular Value Decomposition on **M** as follows,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.1.5.55}$$

Substituting the value of M from equation (1.1.5.55) to(1.1.5.52),

$$\mathbf{USV}^T \mathbf{x} = \mathbf{c} \tag{1.1.5.56}$$

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{c} \tag{1.1.5.57}$$

Where,  $S_+$  is the Moore-Pen-rose Pseudo-Inverse of S. Columns of V are the eigen vectors of  $M^TM$ , columns of U are the eigen vectors of  $MM^T$  and S is diagonal matrix of singular value of eigenvalues of  $M^TM$ .

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} \frac{5}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix}$$
 (1.1.5.58)

Eigen values corresponding to  $\mathbf{M}^T\mathbf{M}$  are given by,

$$\left| \mathbf{M}^T \mathbf{M} - \lambda \mathbf{I} \right| = 0 \quad (1.1.5.59)$$

$$\implies \left| \begin{pmatrix} \frac{5}{4} - \lambda & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} - \lambda \end{pmatrix} \right| = 0 \quad (1.1.5.60)$$

$$\implies \lambda^2 - \frac{45}{16}\lambda + \frac{29}{16} = 0 \tag{1.1.5.61}$$

Hence eigen values of  $\mathbf{M}^T\mathbf{M}$  are,

$$\lambda_1 = \frac{29}{16} \tag{1.1.5.62}$$

$$\lambda_2 = 1 \tag{1.1.5.63}$$

Hence the eigen vectors of  $\mathbf{M}^T\mathbf{M}$  are,

$$\mathbf{v}_1 = \begin{pmatrix} \frac{2}{3} \\ 1 \end{pmatrix} \tag{1.1.5.64}$$

$$\mathbf{v}_2 = \begin{pmatrix} -\frac{3}{2} \\ 1 \end{pmatrix} \tag{1.1.5.65}$$

Normalizing the eigen vectors, we obtain V of (1.1.5.55) as follows,

$$\mathbf{V} = \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}$$
 (1.1.5.66)

**S** of the diagonal matrix of (1.1.5.55) is:

$$\mathbf{S} = \begin{pmatrix} \frac{\sqrt{29}}{4} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.1.5.67}$$

Now, calculating eigen value of  $\mathbf{M}\mathbf{M}^T$ ,

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} \end{pmatrix}$$
 (1.1.5.68)

Eigen values corresponding to  $\mathbf{M}\mathbf{M}^T$  are given by

$$\left|\mathbf{M}\mathbf{M}^T - \lambda \mathbf{I}\right| = 0$$

(1.1.5.69)

(1.1.5.71)

$$\implies \left| \begin{pmatrix} 1 - \lambda & 0 & \frac{1}{2} \\ 0 & 1 - \lambda & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} - \lambda \end{pmatrix} \right| = 0$$

$$\implies \lambda^3 - \frac{45}{16}\lambda^2 + \frac{29}{16}\lambda = 0$$
(1.1.5.70)

Hence eigen values of  $\mathbf{M}^T \mathbf{M}$  are,

$$\lambda_3 = \frac{29}{16} \tag{1.1.5.72}$$

$$\lambda_4 = 1 \tag{1.1.5.73}$$

$$\lambda_5 = 0 \tag{1.1.5.74}$$

Hence we obtain U of (1.1.5.55) as follows,

$$\mathbf{U} = \begin{pmatrix} \frac{8}{\sqrt{377}} & -\frac{3}{\sqrt{13}} & -\frac{2}{\sqrt{29}} \\ \frac{12}{\sqrt{377}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{29}} \\ \frac{13}{\sqrt{377}} & 0 & \frac{4}{\sqrt{29}} \end{pmatrix}$$
(1.1.5.75)

Finally from (1.1.5.55) we get the Singular

Value Decomposition of **M** as follows,

$$\mathbf{M} = \begin{pmatrix} \frac{8}{\sqrt{377}} & -\frac{3}{\sqrt{13}} & -\frac{2}{\sqrt{29}} \\ \frac{12}{\sqrt{377}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{29}} \\ \frac{13}{\sqrt{377}} & 0 & \frac{4}{\sqrt{29}} \end{pmatrix} \begin{pmatrix} \frac{\sqrt{29}}{4} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}^{T}$$

$$(1.1.5.76)$$

Now, Moore-Penrose Pseudo inverse of **S** is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{4}{\sqrt{29}} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.1.5.77}$$

Substituting the values of (1.1.5.75),(1.1.5.66),(1.1.5.77) in (1.1.5.57) we get,

$$\mathbf{U}^{T}\mathbf{c} = \begin{pmatrix} \frac{125}{\sqrt{377}} \\ \frac{-3}{\sqrt{13}} \\ \frac{5}{\sqrt{29}} \end{pmatrix}$$
 (1.1.5.78)

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{c} = \begin{pmatrix} \frac{500}{29\sqrt{13}} \\ -\frac{3}{\sqrt{13}} \end{pmatrix}$$
 (1.1.5.79)

$$\mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{c} = \begin{pmatrix} \frac{97}{29} \\ \frac{102}{29} \end{pmatrix}$$
 (1.1.5.80)

Verifying the solution of (1.1.5.80) using,

$$\implies \mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{c} \tag{1.1.5.81}$$

Evaluating the R.H.S in (1.1.5.81) we get,

$$\mathbf{M}^{T}\mathbf{c} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{3}{4} \end{pmatrix} \begin{pmatrix} 3 \\ 3 \\ 5 \end{pmatrix} = \begin{pmatrix} \frac{11}{2} \\ \frac{27}{4} \end{pmatrix}$$
 (1.1.5.82)

$$\implies \begin{pmatrix} \frac{5}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix} \mathbf{x} = \begin{pmatrix} \frac{11}{2} \\ \frac{27}{4} \end{pmatrix} \qquad (1.1.5.83)$$

The augmented matrix of (1.1.5.83) is,

$$\begin{pmatrix} \frac{5}{4} & \frac{3}{8} & \frac{11}{2} \\ \frac{3}{8} & \frac{25}{16} & \frac{27}{4} \end{pmatrix} \tag{1.1.5.84}$$

Solving the augmented matrix into Row re-

duced echelon form of (1.1.5.84) we get,

$$\begin{pmatrix} \frac{5}{4} & \frac{3}{8} & \frac{11}{2} \\ \frac{3}{8} & \frac{25}{16} & \frac{27}{4} \end{pmatrix} \xrightarrow{R_1 \leftarrow \frac{4}{5}R_1} \begin{pmatrix} 1 & \frac{3}{10} & \frac{22}{5} \\ \frac{3}{8} & \frac{25}{16} & \frac{27}{4} \end{pmatrix} (1.1.5.85)$$

$$\xrightarrow{R_2 \leftarrow R_2 - \frac{3}{8}R_1} \begin{pmatrix} 1 & \frac{3}{10} & \frac{22}{5} \\ 0 & \frac{29}{20} & \frac{51}{10} \end{pmatrix} (1.1.5.86)$$

$$\xrightarrow{R_2 \leftarrow \frac{20}{29}R_2} \begin{pmatrix} 1 & \frac{3}{10} & \frac{22}{5} \\ 0 & 1 & \frac{102}{29} \end{pmatrix} (1.1.5.87)$$

$$\xrightarrow{R_1 \leftarrow R_1 - \frac{3}{10}R_2} \begin{pmatrix} 1 & 0 & \frac{97}{29} \\ 0 & 1 & \frac{102}{29} \end{pmatrix} (1.1.5.88)$$

Therefore,

$$\mathbf{x} = \begin{pmatrix} \frac{97}{29} \\ \frac{102}{29} \end{pmatrix} \tag{1.1.5.89}$$

Comparing results of  $\mathbf{x}$  from (1.1.5.80) and (1.1.5.89), Hence, the solution is verified.

- 1.2 The Pencil of Planes. The Bundle of Planes
- 1.2.1. Write the equation of a plane through the point A (-3, 4, -1) and perpendicular to the line

$$\frac{x+2}{-3} = \frac{y-2}{1} = \frac{z-4}{2}$$
 (1.2.1.1)

### **Solution:**

Let the equation of plane is

$$ax + by + cz + d = 0$$
 (1.2.1.2)

Direction ratio of the line (1.2.1.1) is given as

$$\mathbf{D} = \begin{pmatrix} -3\\1\\2 \end{pmatrix} \tag{1.2.1.3}$$

Now let consider

$$\mathbf{A} = \begin{pmatrix} -3 & 4 & -1 \end{pmatrix} \tag{1.2.1.4}$$

Since plane is passing through the point A (-3, 4, -1) and perpendicular to the line (1.2.1.1), hence

$$\mathbf{AD} + d = 0 \tag{1.2.1.5}$$

$$\implies d = -11 \tag{1.2.1.6}$$

Hence equation of the plane is

$$-3x + y + 2z - 11 = 0 (1.2.1.7)$$

$$\implies$$
  $-3x + y + 2z = 11$  (1.2.1.8)

equation (1.2.1.8) can written as:

$$(-3 \ 1 \ 2)\mathbf{x} = 11$$
 (1.2.1.9)

For foot perpendicular we need to find the distance between the plane and point P (-2, 2, 4). First we find orthogonal vectors  $\mathbf{m_1}$  and  $\mathbf{m_2}$  to

the given normal vector **n**. Let,  $\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$ , then

$$\mathbf{m}^{\mathbf{T}}\mathbf{n} = 0$$
 (1.2.1.10)

$$\implies (a \quad b \quad c) \begin{pmatrix} -3\\1\\2 \end{pmatrix} = 0 \tag{1.2.1.11}$$

$$\implies$$
  $-3a + b + 2c = 0$  (1.2.1.12)

Putting a=1 and b=0 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ \frac{3}{2} \end{pmatrix} \tag{1.2.1.13}$$

Putting a=0 and b=1 we get,

$$\mathbf{m_2} = \begin{pmatrix} 0\\1\\-\frac{1}{2} \end{pmatrix} \qquad (1.2.1.14)$$

Now we solve the equation,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.2.1.15}$$

Putting values in (1.2.1.15),

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{3}{2} & -\frac{1}{2} \end{pmatrix} \mathbf{x} = \begin{pmatrix} -2 \\ 2 \\ 4 \end{pmatrix}$$
 (1.2.1.16)

Now, to solve (1.2.1.16), we perform Singular Value Decomposition on **M** as follows,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{1.2.1.17}$$

Where the columns of V are the eigen vectors of  $M^TM$ , the columns of U are the eigen vectors of  $MM^T$  and S is diagonal matrix of singular value of eigenvalues of  $M^TM$ .

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} \frac{13}{4} & -\frac{3}{4} \\ -\frac{3}{4} & \frac{5}{4} \end{pmatrix}$$
 (1.2.1.18)

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 & \frac{3}{2} \\ 0 & 1 & -\frac{1}{2} \\ \frac{3}{2} & -\frac{1}{2} & \frac{5}{2} \end{pmatrix}$$
 (1.2.1.19)

From (1.2.1.15) putting (1.2.1.17) we get,

$$\mathbf{USV}^T \mathbf{x} = \mathbf{b} \tag{1.2.1.20}$$

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{\mathbf{T}}\mathbf{b} \tag{1.2.1.21}$$

Where  $S_+$  is Moore-Penrose Pseudo-Inverse of **S**.Now, calculating eigen value of  $\mathbf{M}\mathbf{M}^T$ ,

$$\left| \mathbf{M} \mathbf{M}^T - \lambda \mathbf{I} \right| = 0 \quad (1.2.1.22)$$

$$\implies \begin{pmatrix} 1 - \lambda & 0 & \frac{3}{2} \\ 0 & 1 - \lambda & -\frac{1}{2} \\ \frac{3}{2} & -\frac{1}{2} & \frac{5}{2} - \lambda \end{pmatrix} = 0 \quad (1.2.1.23)$$

$$\implies \lambda(\lambda - 1)(\lambda - \frac{7}{2}) = 0 \quad (1.2.1.24)$$

Hence eigen values of  $\mathbf{M}\mathbf{M}^T$  are,

$$\lambda_1 = \frac{7}{2} \tag{1.2.1.25}$$

$$\lambda_2 = 1 \tag{1.2.1.26}$$

$$\lambda_1 = 0 \tag{1.2.1.27}$$

$$\lambda_3 = 0$$
 (1.2.1.27)

Hence the eigen vectors of  $\mathbf{M}\mathbf{M}^T$  are,

$$\mathbf{u}_{1} = \begin{pmatrix} \frac{3}{5} \\ -\frac{1}{5} \\ 1 \end{pmatrix}, \mathbf{u}_{2} = \begin{pmatrix} \frac{1}{3} \\ 1 \\ 0 \end{pmatrix}, \mathbf{u}_{3} = \begin{pmatrix} -\frac{3}{2} \\ \frac{1}{2} \\ 1 \end{pmatrix},$$
(1.2.1.28)

Normalizing the eigen vectors we get,

$$\mathbf{u}_{1} = \begin{pmatrix} \frac{3}{\sqrt{35}} \\ -\frac{1}{\sqrt{35}} \\ \frac{5}{\sqrt{35}} \end{pmatrix}, \mathbf{u}_{2} = \begin{pmatrix} \frac{1}{\sqrt{10}} \\ \frac{3}{\sqrt{10}} \\ 0 \end{pmatrix}, \mathbf{u}_{3} = \begin{pmatrix} -\frac{3}{\sqrt{14}} \\ \frac{1}{\sqrt{14}} \\ \frac{2}{\sqrt{14}} \\ \frac{2}{\sqrt{14}} \end{pmatrix}$$

$$(1.2.1.29)$$

Hence we obtain U of (1.2.1.17) as follows,

$$\mathbf{U} = \begin{pmatrix} \frac{3}{\sqrt{35}} & \frac{1}{\sqrt{10}} & -\frac{3}{\sqrt{14}} \\ -\frac{1}{\sqrt{35}} & \frac{3}{\sqrt{10}} & \frac{1}{\sqrt{14}} \\ \frac{5}{\sqrt{35}} & 0 & \frac{2}{\sqrt{14}} \end{pmatrix}$$
 (1.2.1.30)

After computing the singular values from eigen values  $\lambda_1, \lambda_2, \lambda_3$  we get **S** of (1.2.1.17) as follows,

$$\mathbf{S} = \begin{pmatrix} \sqrt{\frac{7}{2}} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.2.1.31}$$

Now, calculating eigen value of  $\mathbf{M}^T\mathbf{M}$ ,

$$\left|\mathbf{M}^{T}\mathbf{M} - \lambda \mathbf{I}\right| = 0 \qquad (1.2.1.32)$$

$$\implies \begin{pmatrix} \frac{13}{4} - \lambda & -\frac{3}{4} \\ -\frac{3}{4} & \frac{5}{4} - \lambda \end{pmatrix} = 0 \qquad (1.2.1.33)$$

$$\implies \lambda^2 - \frac{9}{2}\lambda + \frac{7}{2} = 0 \qquad (1.2.1.34)$$

Hence eigen values of  $\mathbf{M}^T\mathbf{M}$  are,

$$\lambda_4 = \frac{7}{2} \tag{1.2.1.35}$$

$$\lambda_5 = 1$$
 (1.2.1.36)

Hence the eigen vectors of  $\mathbf{M}^T \mathbf{M}$  are,

$$\mathbf{v}_1 = \begin{pmatrix} -3\\1 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} \frac{1}{3}\\1 \end{pmatrix} \tag{1.2.1.37}$$

Normalizing the eigen vectors we get,

$$\mathbf{v}_1 = \begin{pmatrix} -\frac{3}{\sqrt{10}} \\ \frac{1}{\sqrt{10}} \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} \frac{1}{\sqrt{10}} \\ \frac{3}{\sqrt{10}} \end{pmatrix}$$
 (1.2.1.38)

Hence we obtain V of (1.2.1.17) as follows,

$$\mathbf{V} = \begin{pmatrix} -\frac{3}{\sqrt{10}} & \frac{1}{\sqrt{10}} \\ \frac{1}{\sqrt{10}} & \frac{3}{\sqrt{10}} \end{pmatrix}$$
 (1.2.1.39)

Finally from (1.2.1.17) we get the Singualr Value Decomposition of M as follows,

$$\mathbf{M} = \begin{pmatrix} \frac{3}{\sqrt{35}} & \frac{1}{\sqrt{10}} & -\frac{3}{\sqrt{14}} \\ -\frac{1}{\sqrt{35}} & \frac{3}{\sqrt{10}} & \frac{1}{\sqrt{14}} \\ \frac{5}{\sqrt{35}} & 0 & \frac{2}{\sqrt{14}} \end{pmatrix} \begin{pmatrix} \sqrt{\frac{7}{2}} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{3}{\sqrt{10}} & \frac{1}{\sqrt{10}} \\ -\frac{1}{\sqrt{10}} & \frac{3}{\sqrt{10}} \end{pmatrix}^{T}$$

$$(1.2.1.40)$$

Now, Moore-Penrose Pseudo inverse of S is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \sqrt{\frac{2}{7}} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \tag{1.2.1.41}$$

From (1.2.1.21) we get,

$$\mathbf{U}^T \mathbf{b} = \begin{pmatrix} \frac{12}{\sqrt{35}} \\ \frac{2\sqrt{2}}{\sqrt{5}} \\ \frac{8\sqrt{2}}{\sqrt{7}} \end{pmatrix}$$
 (1.2.1.42)

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{12\sqrt{10}}{35} \\ \frac{2\sqrt{10}}{5} \end{pmatrix}$$
 (1.2.1.43)

$$\mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} \frac{10}{7} \\ \frac{6}{7} \end{pmatrix}$$
 (1.2.1.44)

Verifying the solution of (1.2.1.44) using,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{1.2.1.45}$$

Evaluating the R.H.S in (1.2.1.45) we get,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \begin{pmatrix} -7 \\ \frac{1}{2} \end{pmatrix} \tag{1.2.1.46}$$

$$\implies \begin{pmatrix} \frac{13}{4} & -\frac{3}{4} \\ -\frac{3}{4} & \frac{5}{4} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 4 \\ 0 \end{pmatrix} \tag{1.2.1.47}$$

Solving the augmented matrix of (1.2.1.47) we get,

$$\begin{pmatrix} \frac{13}{4} & -\frac{3}{4} & 4 \\ -\frac{3}{4} & \frac{5}{4} & 0 \end{pmatrix} \stackrel{R_1 = \frac{4}{13}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & -\frac{3}{13} & \frac{16}{13} \\ -\frac{3}{4} & \frac{5}{4} & 0 \end{pmatrix}$$

$$(1.2.1.48)$$

$$\stackrel{R_2 = R_2 + \frac{3}{4}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & -\frac{3}{13} & \frac{16}{13} \\ 0 & \frac{14}{13} & \frac{12}{13} \end{pmatrix}$$

$$(1.2.1.49)$$

$$\stackrel{R_2 = \frac{13}{14}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & -\frac{3}{13} & \frac{16}{13} \\ 0 & 1 & \frac{6}{7} \end{pmatrix}$$

$$(1.2.1.50)$$

$$\stackrel{R_1 = R_1 + \frac{3}{13}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & \frac{10}{7} \\ 0 & 1 & \frac{6}{7} \end{pmatrix}$$

$$(1.2.1.51)$$

Hence, Solution of (1.2.1.45) is given by,

$$\mathbf{x} = \begin{pmatrix} \frac{10}{7} \\ \frac{6}{7} \end{pmatrix} \tag{1.2.1.52}$$

Comparing results of  $\mathbf{x}$  from (1.2.1.44) and (1.2.1.52) we conclude that the solution is verified.

1.2.2. Write the equation of the line through  $A = \begin{pmatrix} 3 \\ 4 \\ -1 \end{pmatrix}$  and perpendicular to the plane 2x - y + 2z - 5 = 0. Determine the coordinates of the point in which the plane is met by this line.

**Solution:** Given a point  $\mathbf{A} = \begin{pmatrix} 3 \\ 4 \\ -1 \end{pmatrix}$  and a plane

(2 -1 2)x = 5. We know that the equation of a plane is given by

$$\mathbf{n}^{\mathbf{T}}\mathbf{x} = c \tag{1.2.2.1}$$

Hence, normal vector  $\mathbf{n}$  is given by

$$\mathbf{n} = \begin{pmatrix} 2 \\ -1 \\ 2 \end{pmatrix} \tag{1.2.2.2}$$

Let  $\mathbf{m_1}$  and  $\mathbf{m_2}$  be two vectors that are normal to normal vector  $\mathbf{n}$ . Let  $\mathbf{m} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$ , then if

$$\mathbf{n}^{\mathbf{T}}\mathbf{m} = 0 \tag{1.2.2.3}$$

$$(2 -1 2) \begin{pmatrix} a \\ b \\ c \end{pmatrix} = 0$$
 (1.2.2.4)

Taking a = 1, b = 0, we get c = -1, and hence

$$\mathbf{m_1} = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \tag{1.2.2.5}$$

Take a = 0 and b = 1, we get  $c = \frac{1}{2}$ , and hence

$$\mathbf{m_2} = \begin{pmatrix} 0\\1\\\frac{1}{2} \end{pmatrix} \tag{1.2.2.6}$$

Since foot of perpendicular is the point where the plane is met by a line perpendicular to the same plane. So, to get foot of perpendicular, we solve

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{1.2.2.7}$$

where

$$\mathbf{M} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -1 & \frac{1}{2} \end{pmatrix}, b = \begin{pmatrix} 3 \\ 4 \\ -1 \end{pmatrix}$$
 (1.2.2.8)

To solve (1.2.2.7), we perform singular value decomposition on  $\mathbf{M}$  given as

$$\mathbf{M} = \mathbf{USV}^{\mathbf{T}} \tag{1.2.2.9}$$

Substituting the value of M from (1.2.2.9) in (1.2.2.7), we get

$$\mathbf{USV}^{\mathrm{T}}\mathbf{x} = \mathbf{b} \tag{1.2.2.10}$$

$$\Longrightarrow \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{\mathrm{T}}\mathbf{b} \tag{1.2.2.11}$$

where,  $S_+$  is Moore-Pen-rose Pseudo-Inverse of S. Columns of U are eigen-vectors of  $MM^T$ , columns of V are eigenvectors of  $M^TM$  and S is diagonal matrix of singular value of eigenvalues of  $M^TM$ . First calculating the eigenvec-

tors corresponding to  $\mathbf{M}^{\mathbf{T}}\mathbf{M}$ .

$$\mathbf{M}^{\mathbf{T}}\mathbf{M} = \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -1 & \frac{1}{2} \end{pmatrix} = \begin{pmatrix} 2 & \frac{-1}{2} \\ \frac{-1}{2} & \frac{5}{4} \end{pmatrix}$$
(1.2.2.12)

Eigen values of  $M^{T}M$  can be found out as

$$\left|\mathbf{M}^{\mathbf{T}}\mathbf{M} - \lambda \mathbf{I}\right| = 0 \tag{1.2.2.13}$$

$$\begin{vmatrix} 2 - \lambda & \frac{-1}{2} \\ \frac{-1}{2} & \frac{5}{4} - \lambda \end{vmatrix} = 0 \tag{1.2.2.14}$$

$$\left(\frac{5}{4} - \lambda\right)(2 - \lambda) - \frac{1}{4} = 0 \tag{1.2.2.15}$$

$$\left(\lambda - \frac{9}{4}\right)(\lambda - 1) = 0 \tag{1.2.2.16}$$

Hence,

$$\lambda_1 = \frac{9}{4}, \lambda_2 = 1 \tag{1.2.2.17}$$

Eigen-vector corresponding to  $\lambda = \frac{9}{4}$ ,

$$\mathbf{v_1} = \begin{pmatrix} 2 \\ -1 \end{pmatrix} \tag{1.2.2.18}$$

Eigen-vector corresponding to  $\lambda = 1$ ,

$$\mathbf{v_2} = \begin{pmatrix} 1\\2 \end{pmatrix} \tag{1.2.2.19}$$

Normalizing, the eigen vectors  $v_1$  and  $v_2$ , we get

$$\mathbf{v_1} = \frac{1}{\sqrt{5}} \begin{pmatrix} 2 \\ -1 \end{pmatrix} \tag{1.2.2.20}$$

$$\mathbf{v_2} = \frac{1}{\sqrt{5}} \begin{pmatrix} 1\\2 \end{pmatrix} \tag{1.2.2.21}$$

Hence,

$$\mathbf{V} = \frac{1}{\sqrt{5}} \begin{pmatrix} 2 & 1 \\ -1 & 2 \end{pmatrix} \tag{1.2.2.22}$$

Now calculating the eigenvectors corresponding to  $\mathbf{M}\mathbf{M}^T$ 

$$\mathbf{M}\mathbf{M}^{\mathbf{T}} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -1 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & \frac{1}{2} \end{pmatrix}$$
$$= \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & \frac{1}{2} \\ -1 & \frac{1}{2} & \frac{5}{4} \end{pmatrix} \quad (1.2.2.23)$$

Eigen values of MM<sup>T</sup> can be found out as

$$\left|\mathbf{M}\mathbf{M}^{\mathrm{T}} - \lambda \mathbf{I}\right| = 0 \tag{1.2.2.24}$$

$$\begin{vmatrix} 1 - \lambda & 0 & -1 \\ 0 & 1 - \lambda & \frac{1}{2} \\ -1 & \frac{1}{2} & \frac{5}{4} - \lambda \end{vmatrix} = 0$$
 (1.2.2.25)

$$(1 - \lambda)\left((1 - \lambda)\left(\frac{5}{4} - \lambda\right) - \frac{1}{4}\right) - 1 + \lambda = 0$$

$$(1.2.2.26)$$

$$\lambda \left(\lambda - \frac{9}{4}\right)(\lambda - 1) = 0 \tag{1.2.2.27}$$

Hence,

$$\lambda_3 = 0, \lambda_4 = 1, \lambda_5 = \frac{9}{4}$$
 (1.2.2.28)

Eigen-vector corresponding to  $\lambda = 0$ ,

$$\mathbf{v_3} = \begin{pmatrix} 2 \\ -1 \\ 2 \end{pmatrix} \tag{1.2.2.29}$$

Eigen-vector corresponding to  $\lambda = 1$ ,

$$\mathbf{v_4} = \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} \tag{1.2.2.30}$$

Eigen-vector corresponding to  $\lambda = \frac{9}{4}$ ,

$$\mathbf{v_5} = \begin{pmatrix} 4 \\ -2 \\ -5 \end{pmatrix} \tag{1.2.2.31}$$

Normalizing, the eigen vectors  $v_3$ ,  $v_4$  and  $v_5$ , we get

$$\mathbf{v_3} = \frac{1}{3} \begin{pmatrix} 2 \\ -1 \\ 2 \end{pmatrix} = \begin{pmatrix} \frac{2}{3} \\ \frac{-1}{3} \\ \frac{2}{3} \end{pmatrix}$$
 (1.2.2.32)

$$\mathbf{v_4} = \frac{1}{\sqrt{5}} \begin{pmatrix} 1\\2\\0 \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{5}}\\\frac{2}{\sqrt{5}}\\0 \end{pmatrix}$$
 (1.2.2.33)

$$\mathbf{v_5} = \frac{1}{3\sqrt{5}} \begin{pmatrix} 4\\-2\\-5 \end{pmatrix} = \begin{pmatrix} \frac{4}{3\sqrt{5}}\\ \frac{-2}{3\sqrt{5}}\\ \frac{-5}{3\sqrt{5}} \end{pmatrix}$$
(1.2.2.34)

Hence,

$$\mathbf{U} = \begin{pmatrix} \frac{4}{3\sqrt{5}} & \frac{1}{\sqrt{5}} & \frac{2}{3} \\ \frac{-2}{3\sqrt{5}} & \frac{2}{\sqrt{5}} & \frac{-1}{3} \\ \frac{-5}{3\sqrt{5}} & 0 & \frac{2}{3} \end{pmatrix}$$
 (1.2.2.35)

Now **S** corresponding to eigenvalues  $\lambda_5$ ,  $\lambda_4$  and  $\lambda_3$  is as follows,

$$\mathbf{S} = \begin{pmatrix} \frac{3}{2} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{1.2.2.36}$$

Now, Moore-Pen-Rose Pseudo inverse of **S** is given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{2}{3} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{1.2.2.37}$$

Hence, we get singular value decomposition of **M** as,

$$\mathbf{M} = \begin{pmatrix} \frac{4}{3\sqrt{5}} & \frac{1}{\sqrt{5}} & \frac{2}{3} \\ \frac{-2}{3\sqrt{5}} & \frac{2}{\sqrt{5}} & \frac{-1}{3} \\ \frac{-5}{3\sqrt{5}} & 0 & \frac{2}{3} \end{pmatrix} \begin{pmatrix} \frac{3}{2} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \frac{1}{\sqrt{5}} \begin{pmatrix} 2 & -1 \\ 1 & 2 \end{pmatrix}$$
(1.2.2.38)

Substituting values of (1.2.2.8), (1.2.2.22), (1.2.2.35) and (1.2.2.36) into (1.2.2.11), we get

$$\mathbf{U}^{\mathbf{T}}\mathbf{b} = \begin{pmatrix} \frac{4}{3\sqrt{5}} & \frac{-2}{3\sqrt{5}} & \frac{-5}{3\sqrt{5}} \\ \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} & 0 \\ \frac{2}{3} & \frac{-1}{3} & \frac{2}{3} \end{pmatrix} \begin{pmatrix} 3\\4\\-1 \end{pmatrix}$$
 (1.2.2.39)

$$\implies \mathbf{U}^{\mathbf{T}}\mathbf{b} = \begin{pmatrix} \frac{3}{\sqrt{5}} \\ \frac{11}{\sqrt{5}} \\ 0 \end{pmatrix} \qquad (1.2.2.40)$$

Now,

$$\mathbf{VS}_{+} = \frac{1}{\sqrt{5}} \begin{pmatrix} 2 & 1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} \frac{2}{3} & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \qquad (1.2.2.41)$$

$$\implies \mathbf{VS}_{+} = \frac{1}{\sqrt{5}} \begin{pmatrix} \frac{4}{3} & 1 & 0 \\ \frac{-2}{3} & 2 & 0 \end{pmatrix} \qquad (1.2.2.42)$$

Now, by (1.2.2.11), we have

$$\mathbf{x} = \frac{1}{\sqrt{5}} \begin{pmatrix} \frac{4}{3} & 1 & 0 \\ \frac{-2}{3} & 2 & 0 \end{pmatrix} \begin{pmatrix} \frac{3}{\sqrt{5}} \\ \frac{11}{\sqrt{5}} \\ 0 \end{pmatrix}$$
 (1.2.2.43)

$$\implies \mathbf{x} = \begin{pmatrix} 3 \\ 4 \end{pmatrix} \tag{1.2.2.44}$$

Now, we verify our solution using

$$\mathbf{M}^{\mathbf{T}}\mathbf{M}\mathbf{x} = \mathbf{M}^{\mathbf{T}}\mathbf{b} \qquad (1.2.2.45)$$

$$\implies \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -1 & \frac{1}{2} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 3 \\ 4 \\ -1 \end{pmatrix} \qquad (1.2.2.46)$$

$$\Longrightarrow \begin{pmatrix} 2 & \frac{-1}{2} \\ \frac{-1}{2} & \frac{5}{4} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 4 \\ \frac{7}{2} \end{pmatrix} \tag{1.2.2.47}$$

Solving the augumented matrix, we get

$$\begin{pmatrix}
2 & \frac{-1}{2} & 4 \\
\frac{-1}{2} & \frac{5}{4} & \frac{7}{2}
\end{pmatrix}
\qquad \stackrel{r_1=(1/2)*(r_1)}{\longleftarrow} \begin{pmatrix}
1 & \frac{-1}{4} & 2 \\
\frac{-1}{2} & \frac{5}{4} & \frac{7}{2}
\end{pmatrix}
\qquad (1.2.2.48)$$

$$\begin{pmatrix}
1 & \frac{-1}{4} & 2 \\
\frac{-1}{2} & \frac{5}{4} & \frac{7}{2}
\end{pmatrix}
\qquad \stackrel{r_2=r_2+(1/2)*(r_1)}{\longleftarrow} \begin{pmatrix}
1 & \frac{-1}{4} & 2 \\
0 & \frac{9}{8} & \frac{9}{2}
\end{pmatrix}
\qquad (1.2.2.49)$$

$$\begin{pmatrix}
1 & \frac{-1}{4} & 2 \\
0 & \frac{9}{8} & \frac{9}{2}
\end{pmatrix}
\qquad \stackrel{r_2=(8/9)*(r_2)}{\longleftarrow} \begin{pmatrix}
1 & \frac{-1}{4} & 2 \\
0 & 1 & 4
\end{pmatrix}
\qquad (1.2.2.50)$$

$$\begin{pmatrix}
1 & \frac{-1}{4} & 2 \\
0 & 1 & 4
\end{pmatrix}
\qquad (1.2.2.51)$$

Thus,

$$\mathbf{x} = \begin{pmatrix} 3 \\ 4 \end{pmatrix} \tag{1.2.2.52}$$

verifying the result from SVD.

Now, we solve for third coordinate of foot of perpendicular by,

$$\mathbf{n}^{\mathbf{T}}\mathbf{x} = 5 \tag{1.2.2.53}$$

$$(2 -1 2)\begin{pmatrix} 3 \\ -4 \\ z \end{pmatrix} = 5$$
 (1.2.2.54)

$$z = \frac{-5}{2} \tag{1.2.2.55}$$

Normalizing z, we get

$$z = \frac{\left(\frac{-5}{2}\right)}{3} \implies z = \frac{-5}{6}$$
 (1.2.2.56)

Hence, coordinate of foot of perpendicular is

$$\mathbf{x} = \begin{pmatrix} 3\\4\\\frac{-5}{6} \end{pmatrix} \tag{1.2.2.57}$$

Now, we try to find equation of straight line through  $\mathbf{P} = \begin{pmatrix} 3 \\ 4 \\ -1 \end{pmatrix}$  and having direction cosines as  $\mathbf{Q} = \begin{pmatrix} 2 \\ -1 \\ 2 \end{pmatrix}$ 

$$L_1: \mathbf{x} = \begin{pmatrix} 3\\4\\-1 \end{pmatrix} + \lambda \begin{pmatrix} 2\\-1\\2 \end{pmatrix} \tag{1.2.2.58}$$