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EE5609: Matrix Theory Assignment-7

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Abstract—This document explains how to find the foot of the perpendicular from a particular point to the given plane using SVD.

Download all latex-tikz codes from

https://github.com/vimalkb007/EE5609/tree/master/ Assignment_7

and all python codes from

https://github.com/vimalkb007/EE5609/tree/master/ Assignment_7/codes

1 Problem

Find the foot of the perpendicular using svd drawn from $\begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix}$ to the plane

$$(2 \quad 3 \quad -4)\mathbf{x} + 5 = 0 \tag{1.0.1}$$

2 EXPLANATION

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

$$\mathbf{m}^{\mathbf{T}}\mathbf{n} = 0 \tag{2.0.1}$$

$$\implies \left(a \quad b \quad c\right) \begin{pmatrix} 2\\3\\-4 \end{pmatrix} = 0 \tag{2.0.2}$$

$$\implies 2a + 3b - 4c = 0 \tag{2.0.3}$$

Let a=1 and b=0 we get,

$$\mathbf{m_1} = \begin{pmatrix} 1\\0\\\frac{1}{2} \end{pmatrix} \tag{2.0.4}$$

Let a=0 and b=1 we get,

$$\mathbf{m_2} = \begin{pmatrix} 0 \\ 1 \\ \frac{3}{4} \end{pmatrix} \tag{2.0.5}$$

Let us solve the equation,

$$\mathbf{M}\mathbf{x} = \mathbf{b} \tag{2.0.6}$$

Substituting (2.0.4) and (2.0.5) in (2.0.6),

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ \frac{1}{2} & \frac{3}{4} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 3 \\ -2 \\ 0 \end{pmatrix} \tag{2.0.7}$$

To solve (2.0.7), we will perform Singular Value Decomposition on \mathbf{M} as follows,

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \tag{2.0.8}$$

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 .

$$\mathbf{M}^T \mathbf{M} = \begin{pmatrix} \frac{5}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix}$$
 (2.0.9)

$$\mathbf{M}\mathbf{M}^{T} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} \end{pmatrix}$$
 (2.0.10)

Substituting (2.0.8) in (2.0.6),

$$\mathbf{USV}^T \mathbf{x} = \mathbf{b} \tag{2.0.11}$$

$$\implies \mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{\mathbf{T}}\mathbf{b} \tag{2.0.12}$$

Where S_+ is Moore-Penrose Pseudo-Inverse of S. Let us calculate eigen values of MM^T ,

$$\left|\mathbf{M}\mathbf{M}^{T} - \lambda \mathbf{I}\right| = 0 \tag{2.0.13}$$

$$\implies \begin{pmatrix} 1 - \lambda & 0 & \frac{1}{2} \\ 0 & 1 - \lambda & \frac{3}{4} \\ \frac{1}{2} & \frac{3}{4} & \frac{13}{16} - \lambda \end{pmatrix} = 0 \qquad (2.0.14)$$

$$\implies \lambda^3 - \frac{45}{16}\lambda^2 + \frac{29}{16}\lambda = 0 \qquad (2.0.15)$$

From equation (2.0.15) eigen values of $\mathbf{M}\mathbf{M}^T$ are,

$$\lambda_1 = \frac{29}{16}$$
 $\lambda_2 = 1$ $\lambda_3 = 0$ (2.0.16)

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

$$\mathbf{u}_{1} = \begin{pmatrix} \frac{8}{13} \\ \frac{12}{13} \\ 1 \end{pmatrix} \quad \mathbf{u}_{2} = \begin{pmatrix} -\frac{3}{2} \\ 1 \\ 0 \end{pmatrix} \quad \mathbf{u}_{3} = \begin{pmatrix} -\frac{1}{2} \\ -\frac{3}{4} \\ 1 \end{pmatrix} \quad (2.0.17)$$

Normalizing the eigen vectors in equation (2.0.17)

$$\mathbf{u}_{1} = \begin{pmatrix} \frac{8}{\sqrt{377}} \\ \frac{12}{\sqrt{377}} \\ \frac{13}{\sqrt{377}} \end{pmatrix} \quad \mathbf{u}_{2} = \begin{pmatrix} -\frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \\ 0 \end{pmatrix} \quad \mathbf{u}_{3} = \begin{pmatrix} -\frac{2}{\sqrt{29}} \\ -\frac{3}{\sqrt{29}} \\ \frac{4}{\sqrt{29}} \end{pmatrix}$$
(2.0.18)

Hence we obtain **U** as follows,

$$\mathbf{U} = \begin{pmatrix} \frac{8}{\sqrt{377}} & -\frac{3}{\sqrt{13}} & -\frac{2}{\sqrt{29}} \\ \frac{12}{\sqrt{377}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{29}} \\ \frac{13}{\sqrt{377}} & 0 & \frac{4}{\sqrt{29}} \end{pmatrix}$$
 (2.0.19)

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

$$\mathbf{S} = \begin{pmatrix} \frac{\sqrt{29}}{4} & 0\\ 0 & 1\\ 0 & 0 \end{pmatrix} \tag{2.0.20}$$

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

$$|\mathbf{M}^T \mathbf{M} - \lambda \mathbf{I}| = 0 \tag{2.0.21}$$

$$\Longrightarrow \begin{pmatrix} \frac{5}{4} - \lambda & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} - \lambda \end{pmatrix} = 0 \tag{2.0.22}$$

$$\implies \lambda^2 - \frac{45}{16}\lambda + \frac{29}{16} = 0 \tag{2.0.23}$$

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

$$\lambda_1 = \frac{29}{16} \quad \lambda_2 = 1 \tag{2.0.24}$$

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

$$\mathbf{v}_1 = \begin{pmatrix} \frac{2}{3} \\ 1 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} -\frac{3}{2} \\ 1 \end{pmatrix} \tag{2.0.25}$$

Normalizing the eigen vectors,

$$\mathbf{v}_1 = \begin{pmatrix} \frac{2}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} -\frac{3}{\sqrt{13}} \\ \frac{2}{\sqrt{13}} \end{pmatrix}$$
 (2.0.26)

Hence we obtain V as,

$$\mathbf{V} = \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}$$
 (2.0.27)

From (2.0.6), the Singular Value Decomposition of **M** is as follows,

$$\mathbf{M} = \begin{pmatrix} \frac{8}{\sqrt{377}} & -\frac{3}{\sqrt{13}} & -\frac{2}{\sqrt{29}} \\ \frac{12}{\sqrt{377}} & \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{29}} \\ \frac{13}{\sqrt{377}} & 0 & \frac{4}{\sqrt{29}} \end{pmatrix} \begin{pmatrix} \frac{\sqrt{29}}{4} & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{2}{\sqrt{13}} & -\frac{3}{\sqrt{13}} \\ \frac{3}{\sqrt{13}} & \frac{2}{\sqrt{13}} \end{pmatrix}^{T}$$
(2.0.28)

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

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{4}{\sqrt{29}} & 0 & 0\\ 0 & 1 & 0 \end{pmatrix} \tag{2.0.29}$$

From (2.0.12) we get,

$$\mathbf{U}^T \mathbf{b} = \begin{pmatrix} 0 \\ -\sqrt{13} \\ 0 \end{pmatrix} \tag{2.0.30}$$

$$\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 0\\ -\sqrt{13} \end{pmatrix} \tag{2.0.31}$$

$$\mathbf{x} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{T}\mathbf{b} = \begin{pmatrix} 3\\ -2 \end{pmatrix}$$
 (2.0.32)

Verifying the solution of (2.0.32) using,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \mathbf{M}^T \mathbf{b} \tag{2.0.33}$$

Evaluating the R.H.S in (2.0.33) we get,

$$\mathbf{M}^T \mathbf{M} \mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix} \tag{2.0.34}$$

$$\implies \begin{pmatrix} \frac{5}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{25}{16} \end{pmatrix} \mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix}$$
 (2.0.35)

Solving the augmented matrix of (2.0.35) we get,

$$\begin{pmatrix} \frac{5}{4} & \frac{3}{8} & 3\\ \frac{3}{8} & \frac{25}{16} & -2 \end{pmatrix} \xrightarrow{R_1 = \frac{4}{5}R_1} \begin{pmatrix} 1 & \frac{3}{10} & \frac{12}{5}\\ \frac{3}{8} & \frac{29}{16} & -2 \end{pmatrix}$$

$$\leftarrow \begin{pmatrix} R_2 = R_2 - \frac{3}{8}R_1\\ 0 & \frac{29}{20} & -\frac{29}{10} \end{pmatrix}$$

$$\leftarrow \begin{pmatrix} 1 & \frac{3}{10} & \frac{12}{5}\\ 0 & \frac{29}{10} & -\frac{29}{10} \end{pmatrix}$$

$$\leftarrow \begin{pmatrix} 1 & \frac{3}{10} & \frac{12}{5}\\ 0 & 1 & -2 \end{pmatrix}$$

$$(2.0.38)$$

$$\stackrel{R_2 = R_2 - \frac{3}{8}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{3}{10} & \frac{12}{5} \\ 0 & \frac{29}{20} & -\frac{29}{10} \end{pmatrix} \qquad (2.0.37)$$

$$\stackrel{R_2 = \frac{20}{29}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{3}{10} & \frac{12}{5} \\ 0 & 1 & -2 \end{pmatrix} \tag{2.0.38}$$

$$\stackrel{R_1=R_1-\frac{3}{10}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & 0 & 3\\ 0 & 1 & -2 \end{pmatrix} \qquad (2.0.39)$$

From equation (2.0.39), solution is given by,

$$\mathbf{x} = \begin{pmatrix} 3 \\ -2 \end{pmatrix} \tag{2.0.40}$$

Comparing results of \mathbf{x} from (2.0.32) and (2.0.40), we can say that the solution is verified.