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EE5609: Matrix Theory Assignment-6

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1 Problem

Find the centre of parabola using SVD and verify through least square method.

$$144y^2 - 120xy + 25x^2 + 619x - 272y + 663 = 0$$
(1.0.1)

2 Solution

The general second degree equation can be expressed as follows,

$$\mathbf{x}^T \mathbf{V} \mathbf{x} + 2\mathbf{u}^T \mathbf{x} + f = 0 \tag{2.0.1}$$

where,

$$\mathbf{V} = \begin{pmatrix} 144 & -60 \\ -60 & 25 \end{pmatrix} \tag{2.0.2}$$

$$\mathbf{u} = \begin{pmatrix} \frac{619}{2} \\ -136 \end{pmatrix} \tag{2.0.3}$$

$$f = 663 (2.0.4)$$

1) Expanding the determinant of V we observe,

$$\begin{vmatrix} 144 & -60 \\ -60 & 25 \end{vmatrix} = 0 \tag{2.0.5}$$

Also

$$\begin{vmatrix} \mathbf{V} & \mathbf{u} \\ \mathbf{u}^T & f \end{vmatrix} = \begin{vmatrix} 144 & -60 & \frac{619}{2} \\ -60 & 25 & -136 \\ \frac{619}{2} & -136 & 663 \end{vmatrix}$$
 (2.0.6)

Hence from (2.0.5) and (2.0.7) we conclude that given equation is an parabola. The characteristic equation of **V** is given as follows,

$$\left| \lambda \mathbf{I} - \mathbf{V} \right| = \begin{vmatrix} \lambda - 144 & 60 \\ 60 & \lambda - 25 \end{vmatrix} = 0$$
 (2.0.8)

$$\implies \lambda^2 - 169\lambda = 0 \qquad (2.0.9)$$

Hence the characteristic equation of V is given by (2.0.9). The roots of (2.0.9) i.e the eigenvalues are given by

$$\lambda_1 = 0, \lambda_2 = 169 \tag{2.0.10}$$

2) For $\lambda_1 = 0$, the eigen vector **p** is given by

$$\mathbf{Vp} = 0 \tag{2.0.11}$$

Row reducing V yields

$$\implies \begin{pmatrix} -144 & 60 \\ 60 & -25 \end{pmatrix} \xrightarrow[R_2=R_2+5R_1]{} \begin{pmatrix} -12 & 5 \\ 0 & 0 \end{pmatrix}$$
(2.0.12)

$$\implies \mathbf{p}_1 = \frac{1}{13} \begin{pmatrix} 5 \\ 12 \end{pmatrix}$$
(2.0.13)

Similarly,

$$\mathbf{p}_2 = \frac{1}{13} \begin{pmatrix} 12 \\ -5 \end{pmatrix} \tag{2.0.14}$$

Thus, the eigenvector rotation matrix and the eigenvalue matrix are

$$\mathbf{P} = \begin{pmatrix} \mathbf{p_1} & \mathbf{p_2} \end{pmatrix} = \frac{1}{13} \begin{pmatrix} 5 & 12 \\ 12 & -5 \end{pmatrix}$$
 (2.0.15)

$$\mathbf{D} = \begin{pmatrix} 0 & 0 \\ 0 & 169 \end{pmatrix} \tag{2.0.16}$$

The focal length of the parabola is given by

$$\frac{\left|2\mathbf{u}^{T}\mathbf{p}_{1}\right|}{\lambda_{2}} = \frac{13}{169} = \frac{1}{13} \tag{2.0.17}$$

and its equation is

$$\mathbf{y}^{\mathbf{T}}\mathbf{D}\mathbf{y} = -\eta \begin{pmatrix} 1 & 0 \end{pmatrix} \mathbf{y} \tag{2.0.18}$$

where

$$\eta = 2\mathbf{u}^T \mathbf{p_1} = -13 \tag{2.0.19}$$

and the vertex c is given by

$$\begin{pmatrix} \mathbf{u}^{\mathrm{T}} + \frac{\eta}{2} \mathbf{p}_{1}^{\mathrm{T}} \\ \mathbf{V} \end{pmatrix} \mathbf{c} = \begin{pmatrix} -f \\ \frac{\eta}{2} \mathbf{p}_{1} - \mathbf{u} \end{pmatrix}$$
 (2.0.20)

using equations (2.0.3),(2.0.4) and (2.0.13)

$$\begin{pmatrix} 307 & -142 \\ 144 & -60 \\ -60 & 25 \end{pmatrix} \mathbf{c} = \begin{pmatrix} -663 \\ -312 \\ 130 \end{pmatrix}$$
 (2.0.21)

2.1 Singular Value Decomposition:

$$\mathbf{Mc} = \mathbf{b} \tag{2.1.1}$$

where

$$\mathbf{M} = \begin{pmatrix} 307 & -142 \\ 144 & -60 \\ -60 & 25 \end{pmatrix}, b = \begin{pmatrix} -663 \\ -312 \\ 130 \end{pmatrix}$$
 (2.1.2)

To solve (2.1.1), we perform singular value decomposition on \mathbf{M} given as

$$\mathbf{M} = \mathbf{USV^{T}} \tag{2.1.3}$$

Substituting the value of M from (2.1.3) in (2.1.1), we get

$$\mathbf{USV}^{\mathrm{T}}\mathbf{c} = \mathbf{b} \tag{2.1.4}$$

$$\Longrightarrow \mathbf{c} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{\mathrm{T}}\mathbf{b} \tag{2.1.5}$$

where, S_+ is Moore-Pen-rose Pseudo-Inverse of S. Columns of U are eigen-vectors 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 $\mathbf{M}^T\mathbf{M}$.

Using Python, U, S and V for M is given by:

$$\mathbf{U} = \begin{pmatrix} -\frac{8946}{10000} & \frac{4468}{10000} & 0\\ -\frac{4124}{10000} & -\frac{8258}{10000} & \frac{3846}{10000}\\ \frac{1718}{10000} & \frac{3441}{10000} & \frac{9231}{10000} \end{pmatrix}$$
(2.1.6)

$$\mathbf{S} = \begin{pmatrix} \frac{3780744}{10000} & 0\\ 0 & \frac{58110}{10000}\\ 0 & 0 \end{pmatrix} \tag{2.1.7}$$

$$\mathbf{V} = \begin{pmatrix} \frac{-9108}{10000} & \frac{-4128}{10000} \\ \frac{4128}{10000} & \frac{-9108}{10000} \end{pmatrix} \tag{2.1.8}$$

Now, More-Pen-Rose Pseudo inverse of S is

given by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{10000}{3780744} & 0 & 0\\ 0 & \frac{10000}{58110} & 0 \end{pmatrix} \tag{2.1.9}$$

Hence, we get singular value decomposition of **M** as,

$$\mathbf{M} = \begin{pmatrix} -\frac{8946}{10000} & \frac{4468}{10000} & 0\\ \frac{-4124}{10000} & \frac{-8258}{10000} & \frac{3846}{10000} \\ \frac{1718}{10000} & \frac{3446}{10000} & \frac{9231}{10000} \end{pmatrix} \begin{pmatrix} \frac{3780744}{10000} & 0\\ 0 & \frac{58110}{10000} \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{-9108}{10000} & \frac{4128}{10000} \\ \frac{-4128}{10000} & \frac{-9108}{10000} \end{pmatrix}$$

$$(2.1.10)$$

From (2.1.2) and (2.1.6)

$$\mathbf{U}^{\mathbf{T}}\mathbf{b} = \begin{pmatrix} -\frac{8946}{10000} & \frac{-4124}{10000} & 0\\ \frac{-4468}{10000} & \frac{-8258}{10000} & \frac{3846}{10000} \\ \frac{1718}{10000} & \frac{3441}{10000} & \frac{9231}{10000} \end{pmatrix} \begin{pmatrix} -663\\ -312\\ 130 \end{pmatrix} = \begin{pmatrix} \frac{7441606}{10000} \\ \frac{61658}{10000} \\ 0 \end{pmatrix}$$

From (2.1.8) and (2.1.9)

$$\mathbf{VS}_{+} = \begin{pmatrix} \frac{-9108}{10000} & \frac{-4128}{10000} \\ \frac{4128}{10000} & \frac{-9108}{10000} \end{pmatrix} \begin{pmatrix} \frac{10000}{3780744} & 0 & 0 \\ 0 & \frac{10000}{58110} & 0 \end{pmatrix} = \begin{pmatrix} \frac{-24}{10000} & \frac{-710}{10000} \\ \frac{11}{10000} & \frac{-1567}{10000} \end{pmatrix}$$

$$(2.1.12)$$

Substitute (2.1.12) and (2.1.11) in (2.1.5) we get,

$$\mathbf{c} = \begin{pmatrix} \frac{-24}{10000} & \frac{-710}{10000} & 0\\ \frac{11}{10000} & \frac{-1567}{10000} & 0 \end{pmatrix} \begin{pmatrix} \frac{7441606}{10000}\\ \frac{61658}{10000}\\ 0 \end{pmatrix}$$
 (2.1.13)

$$\implies \boxed{\mathbf{c} = \begin{pmatrix} \frac{-22308}{10000} \\ \frac{-1538}{10000} \end{pmatrix} = \begin{pmatrix} -2.23 \\ -0.15 \end{pmatrix}}$$
 (2.1.14)

2.2 Least Square Verification

Now, verify our solution using,

$$\mathbf{M}^{\mathbf{T}}\mathbf{M}\mathbf{c} = \mathbf{M}^{\mathbf{T}}\mathbf{b} \tag{2.2.1}$$

From (2.1.2).

$$\begin{pmatrix} 307 & 144 & -60 \\ -142 & -60 & 25 \end{pmatrix} \begin{pmatrix} 307 & -142 \\ 144 & -60 \\ -60 & 25 \end{pmatrix} \mathbf{c} \quad (2.2.2)$$

$$= \begin{pmatrix} 307 & 144 & -60 \\ -142 & -60 & 25 \end{pmatrix} \begin{pmatrix} -663 \\ -312 \\ 130 \end{pmatrix}$$
 (2.2.3)

$$\implies \begin{pmatrix} 118585 & -53734 \\ -53734 & 24389 \end{pmatrix} \mathbf{c} = \begin{pmatrix} -256269 \\ 116116 \end{pmatrix}$$
(2.2.4)

Solving the augmented matrix, we get

$$\begin{pmatrix} 118585 & -53734 & -256269 \\ -53734 & 24389 & 116116 \end{pmatrix} \xleftarrow{R_1 = \frac{R_1}{118585}} \begin{pmatrix} 1 & \frac{-53734}{118585} & \frac{-256269}{118585} \\ -53734 & 24389 & 116116 \end{pmatrix}$$

$$(2.2.5)$$

$$\xleftarrow{R_2 = R_2 + 53734R_1} \begin{pmatrix} 1 & \frac{-53734}{118585} & \frac{-256269}{118585} \\ 0 & \frac{4826809}{118585} & \frac{-742586}{118585} \end{pmatrix}$$

$$(2.2.6)$$

$$\xleftarrow{R_2 = \frac{118585}{4826809}R_2} \begin{pmatrix} 1 & \frac{-53734}{118585} & \frac{-256269}{118585} \\ 0 & 1 & \frac{-742586}{4826809} \end{pmatrix}$$

$$(2.2.7)$$

$$\xleftarrow{R_1 = R_1 + \frac{53734}{118585}R_2} \begin{pmatrix} 1 & 0 & \frac{-11970539}{5723871} \\ 0 & 1 & \frac{-742586}{4826809} \end{pmatrix}$$

$$(2.2.8)$$

Thus,

$$\implies \mathbf{c} = \begin{pmatrix} \frac{-11970539}{5723871} \\ \frac{-742386}{4826809} \end{pmatrix} = \begin{pmatrix} -2.23 \\ -0.15 \end{pmatrix}$$
 (2.2.9)

Hence, verified the result from SVD.