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Matrix Theory (EE5609) Assignment-6

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Abstract—This document contains the QR decomposition and SVD followed by Least Square Verification.

Download all python codes from

https://github.com/EE20RESCH11008/Matrix-Theory/tree/master/Assignment-6/Code

and latex-tikz codes from

https://github.com/EE20RESCH11008/Matrix-Theory/tree/master/Assignment-6

1 Problem

1). Find QR decomposition of

$$\mathbf{V} = \begin{pmatrix} 16 & -12 \\ -12 & 9 \end{pmatrix} \tag{1.0.1}$$

2). Find the vertex \mathbf{c} of the parabola using SVD:

$$16x^2 - 24xy + 9y^2 + 32x + 86y - 39 = 0 (1.0.2)$$

also verify the result using least squares.

2 Solution

2.1 QR Decomposition

Let, the column vectors of V be v_1 and v_2 :

$$\mathbf{v_1} = \begin{pmatrix} 16 \\ -12 \end{pmatrix} \tag{2.1.1}$$

$$\mathbf{v_2} = \begin{pmatrix} -12\\9 \end{pmatrix} \tag{2.1.2}$$

To find $\mathbf{Q} = \begin{pmatrix} \mathbf{u_1} & \mathbf{u_2} \end{pmatrix}$, we will orthonormalize the columns of \mathbf{V} using Gram-Schmidt method:

$$\mathbf{u}_1 = \frac{\mathbf{v}_1}{k_1} \tag{2.1.3}$$

$$k_1 = ||\mathbf{v_1}|| = \sqrt{16^2 + (-12)^2} = 20$$
 (2.1.4)

$$\implies \mathbf{u_1} = \frac{1}{20} \begin{pmatrix} 16 \\ -12 \end{pmatrix} = \begin{pmatrix} \frac{4}{5} \\ \frac{-3}{5} \end{pmatrix} \tag{2.1.5}$$

$$\mathbf{u}_2 = \frac{\mathbf{v}_2 - r_1 \mathbf{u}_1}{\|\mathbf{v}_2 - r_1 \mathbf{u}_1\|}$$
 (2.1.6)

$$r_{1} = \frac{\mathbf{u_{1}}^{T} \mathbf{v_{2}}}{\|\mathbf{u_{1}}\|^{2}} = \frac{\left(\frac{4}{5} - \frac{-3}{5}\right) \begin{pmatrix} -12\\9 \end{pmatrix}}{\left(\frac{4}{5}\right)^{2} + \left(\frac{-3}{5}\right)^{2}} = -15 \qquad (2.1.7)$$

From (2.1.2) (2.1.5) and (2.1.7),

$$\mathbf{u}_2 = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \tag{2.1.8}$$

$$k_2 = \mathbf{u_2}^T \mathbf{v_2} = 0 \tag{2.1.9}$$

The QR decomposition is given as:

$$\begin{pmatrix} \mathbf{v_1} & \mathbf{v_2} \end{pmatrix} = \begin{pmatrix} \mathbf{u_1} & \mathbf{u_2} \end{pmatrix} \begin{pmatrix} k_1 & r_1 \\ 0 & k_2 \end{pmatrix}$$
 (2.1.10)

Where,

$$\mathbf{Q} = \begin{pmatrix} \mathbf{u_1} & \mathbf{u_2} \end{pmatrix} \tag{2.1.11}$$

$$\mathbf{R} = \begin{pmatrix} k_1 & r_1 \\ 0 & k_2 \end{pmatrix} \tag{2.1.12}$$

From (2.1.5) (2.1.8) and (2.1.11)

$$\mathbf{Q} = \begin{pmatrix} \frac{4}{5} & 0\\ \frac{-3}{5} & 0 \end{pmatrix} = \begin{pmatrix} \frac{4}{5}\\ \frac{-3}{5} \end{pmatrix} \tag{2.1.13}$$

From (2.1.4) (2.1.7) (2.1.9) and (2.1.12)

$$\mathbf{R} = \begin{pmatrix} 20 & -15 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 20 & -15 \end{pmatrix} \tag{2.1.14}$$

Substituting the values of (2.1.5) (2.1.8) (2.1.4) And the vertex **c** is: (2.1.9) and (2.1.7) in (2.1.10) We get,

$$\mathbf{V} = \begin{pmatrix} \frac{4}{5} & 0\\ \frac{-3}{5} & 0 \end{pmatrix} \begin{pmatrix} 20 & -15\\ 0 & 0 \end{pmatrix} = \begin{pmatrix} \frac{4}{5}\\ \frac{-3}{5} \end{pmatrix} \begin{pmatrix} 20 & -15 \end{pmatrix}$$
(2.1.15)

$$\begin{pmatrix} \mathbf{u}^T + \frac{\eta}{2} \mathbf{p_1}^T \\ \mathbf{V} \end{pmatrix} \mathbf{c} = \begin{pmatrix} -f \\ \frac{\eta}{2} \mathbf{p_1} - \mathbf{u} \end{pmatrix}$$
 (2.2.12)

Substitute values of (2.2.2) (2.2.7) and (2.2.9) in (2.2.12) we get,

$$\begin{pmatrix} \frac{212}{5} & \frac{391}{5} \\ 16 & -12 \\ -12 & 9 \end{pmatrix} \mathbf{c} = \begin{pmatrix} 39 \\ \frac{52}{5} \\ \frac{-39}{5} \end{pmatrix}$$
 (2.2.13)

2.2 Vertex of Parabola:

The general second degree equation can be expressed as:

$$\mathbf{x}^T \mathbf{x} + 2\mathbf{u}^T \mathbf{x} + f = 0 \tag{2.2.1}$$

From (1.0.2) and (2.2.1),

$$\mathbf{V} = \begin{pmatrix} 16 & -12 \\ -12 & 9 \end{pmatrix} \quad \mathbf{u} = \begin{pmatrix} 16 \\ 43 \end{pmatrix}, \quad f = -39 \quad (2.2.2) \quad \text{where}$$

Eigen Values of V is given as

$$\left| \lambda \mathbf{I} - \mathbf{V} \right| = 0 \tag{2.2.3}$$

$$\implies \begin{vmatrix} \lambda - 16 & 12 \\ 12 & \lambda - 9 \end{vmatrix} = 0 \tag{2.2.4}$$

$$\implies \lambda^2 - 25\lambda = 0 \tag{2.2.5}$$

Hence,

$$\lambda_1 = 0, \quad \lambda_2 = 25$$
 (2.2.6)

Eigen-vector corresponding to $\lambda_1 = 0$,

$$\implies \mathbf{p_1} = \frac{1}{5} \begin{pmatrix} 3 \\ 4 \end{pmatrix} \tag{2.2.7}$$

Eigen-vector corresponding to $\lambda_2 = 25$,

$$\implies \mathbf{p_2} = \frac{1}{5} \begin{pmatrix} -4 \\ 3 \end{pmatrix} \tag{2.2.8}$$

From (2.2.2) and (2.2.7)

$$\eta = 2\mathbf{p_1}^T \mathbf{u} = 88 \tag{2.2.9}$$

Using (2.2.6) and (2.2.9), Focal length of the parabola is given by:

$$\left| \frac{\eta}{\lambda_2} \right| = \left| \frac{88}{25} \right| = 3.52$$
 (2.2.10)

The standard equation of parabola is given by:

$$\mathbf{y}^T \mathbf{D} \mathbf{y} = -\eta \begin{pmatrix} 1 & 0 \end{pmatrix} \mathbf{y} \tag{2.2.11}$$

2.3 Singular Value Decomposition:

$$\mathbf{Mc} = \mathbf{b} \tag{2.3.1}$$

$$\mathbf{M} = \begin{pmatrix} \frac{212}{5} & \frac{391}{5} \\ 16 & -12 \\ -12 & 9 \end{pmatrix}, b = \begin{pmatrix} 39 \\ \frac{52}{5} \\ -\frac{39}{5} \end{pmatrix}$$
 (2.3.2)

To solve (2.3.1), we perform singular value decomposition on M given as

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^{\mathbf{T}} \tag{2.3.3}$$

Substituting the value of M from (2.3.3) in (2.3.1), we get

$$\mathbf{USV}^{\mathbf{T}}\mathbf{c} = \mathbf{b} \tag{2.3.4}$$

$$\Longrightarrow \mathbf{c} = \mathbf{V}\mathbf{S}_{+}\mathbf{U}^{\mathbf{T}}\mathbf{b} \tag{2.3.5}$$

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

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

$$\mathbf{U} = \begin{pmatrix} \frac{99}{100} & \frac{11}{250} & 0\\ \frac{-7}{200} & \frac{799}{1000} & \frac{3}{5}\\ \frac{26}{1000} & \frac{-599}{1000} & \frac{4}{10} \end{pmatrix}$$
(2.3.6)

$$\mathbf{S} = \begin{pmatrix} \frac{8903}{100} & 0\\ 0 & \frac{2471}{100}\\ 0 & 0 \end{pmatrix} \tag{2.3.7}$$

$$\mathbf{V} = \begin{pmatrix} \frac{93}{200} & \frac{221}{250} \\ \frac{221}{250} & \frac{293}{200} \end{pmatrix} \tag{2.3.8}$$

Now, More-Pen-Rose Pseudo inverse of S is given Solving the augmented matrix, we get by,

$$\mathbf{S}_{+} = \begin{pmatrix} \frac{100}{8903} & 0 & 0\\ 0 & \frac{100}{2471} & 0 \end{pmatrix} \tag{2.3.9}$$

Hence, we get singular value decomposition of M as,

$$\mathbf{M} = \begin{pmatrix} \frac{99}{100} & \frac{11}{250} & 0\\ \frac{-7}{200} & \frac{799}{1000} & \frac{3}{5}\\ \frac{26}{1000} & \frac{-599}{1000} & \frac{4}{10} \end{pmatrix} \begin{pmatrix} \frac{8903}{100} & 0\\ 0 & \frac{2471}{100}\\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{93}{200} & \frac{221}{250}\\ \frac{221}{250} & \frac{-93}{200} \end{pmatrix}$$
(2.3.10)

From (2.3.2) and (2.3.6)

$$\mathbf{U}^{\mathbf{T}}\mathbf{b} = \begin{pmatrix} \frac{99}{100} & \frac{-7}{200} & \frac{26}{1000} \\ \frac{11}{250} & \frac{799}{1000} & \frac{-599}{1000} \\ 0 & \frac{3}{5} & \frac{4}{10} \end{pmatrix} \begin{pmatrix} \frac{39}{52} \\ \frac{52}{5} \\ -\frac{39}{5} \end{pmatrix} = \begin{pmatrix} \frac{3838}{100} \\ \frac{1472}{100} \\ 0 \end{pmatrix} \quad (2.3.11)$$

From (2.3.8) and (2.3.9)

$$\mathbf{VS}_{+} = \begin{pmatrix} \frac{93}{200} & \frac{221}{250} \\ \frac{293}{250} & \frac{-93}{200} \end{pmatrix} \begin{pmatrix} \frac{100}{8903} & 0 & 0 \\ 0 & \frac{100}{2471} & 0 \end{pmatrix} = \begin{pmatrix} \frac{52}{10000} & \frac{358}{10000} & 0 \\ \frac{99}{10000} & \frac{-188}{10000} & 0 \end{pmatrix}$$

$$(2.3.12)$$

Substitute (2.3.12) and (2.3.11) in (2.3.5) we get,

$$\mathbf{c} = \begin{pmatrix} \frac{52}{10000} & \frac{358}{10000} & 0\\ \frac{9}{10000} & \frac{-188}{10000} & 0 \end{pmatrix} \begin{pmatrix} \frac{3838}{100} \\ \frac{1472}{100} \\ 0 \end{pmatrix}$$
 (2.3.13)

$$\implies \left| \mathbf{c} = \left(\frac{\frac{728}{1000}}{\frac{104}{1000}} \right) = \begin{pmatrix} 0.728\\0.104 \end{pmatrix} \right| \tag{2.3.14}$$

$$\mathbf{S} = \begin{pmatrix} \frac{100}{8903} & 0 & 0 \end{pmatrix} \tag{2.3.0}$$

$$\begin{pmatrix} \frac{54944}{25} & \frac{75392}{25} & \frac{9568}{5} \\ \frac{75392}{25} & \frac{158506}{25} & \frac{14274}{5} \end{pmatrix} \xrightarrow{R_1 = \frac{25}{54944}} R_1 \begin{pmatrix} 1 & \frac{75392}{54944} & \frac{47840}{54944} \\ \frac{75392}{25} & \frac{158506}{25} & \frac{14274}{5} \\ & & & & & & & \\ & & & & & \\ & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & & \\ & & & \\ & & & & \\ & & & \\ & & & & \\ & & & \\ & & & & \\$$

$$\stackrel{R_2=R_2-\frac{75392}{25}R_1}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{75392}{54944} & \frac{47840}{54944} \\ 0 & \frac{8809}{4} & \frac{22903}{100} \end{pmatrix} (2.4.5)$$

$$\stackrel{R_2 = \frac{4}{8809}R_2}{\longleftrightarrow} \begin{pmatrix} 1 & \frac{75392}{54944} & \frac{47840}{54944} \\ 0 & 1 & \frac{104}{1000} \end{pmatrix}$$
(2.4.6)

$$\xrightarrow{R_1 = R_1 - \frac{75392}{54944} R_2} \begin{pmatrix} 1 & 0 & \frac{728}{1000} \\ 0 & 1 & \frac{104}{1000} \end{pmatrix}$$
(2.4.7)

Thus,

$$\implies \boxed{\mathbf{c} = \left(\frac{\frac{728}{1000}}{\frac{104}{1000}}\right) = \left(\frac{0.728}{0.104}\right)} \tag{2.4.8}$$

Hence, verified the result from SVD.

2.4 Least Square Verification

Now, verify our solution using,

$$\mathbf{M}^{\mathbf{T}}\mathbf{M}\mathbf{c} = \mathbf{M}^{\mathbf{T}}\mathbf{b} \tag{2.4.1}$$

From (2.3.2),

$$\begin{pmatrix} \frac{212}{5} & 16 & -12\\ \frac{391}{5} & -12 & 9 \end{pmatrix} \begin{pmatrix} \frac{212}{5} & \frac{391}{5}\\ 16 & -12\\ -12 & 9 \end{pmatrix} \mathbf{c} = \begin{pmatrix} \frac{212}{5} & 16 & -12\\ \frac{391}{5} & -12 & 9 \end{pmatrix} \begin{pmatrix} \frac{39}{52}\\ \frac{52}{5}\\ -\frac{39}{5} \end{pmatrix}$$
(2.4.2)

$$\Longrightarrow \begin{pmatrix} \frac{54944}{25} & \frac{75392}{25} \\ \frac{75392}{25} & \frac{158506}{25} \end{pmatrix} \mathbf{c} = \begin{pmatrix} \frac{9568}{14274} \\ \frac{14274}{5} \end{pmatrix}$$
 (2.4.3)