

# Assignment-7

Hrithik Raj

**Abstract—**This document contains solution of Problem Ramsey(4.1.4)

Download latex-tikz codes from

[https://github.com/Hrithikraj2/MatrixTheory\\_EE5609/blob/master/Assignment\\_7/A7.tex](https://github.com/Hrithikraj2/MatrixTheory_EE5609/blob/master/Assignment_7/A7.tex)

## 1 PROBLEM

1)Find QR decomposition of matrix

$$\mathbf{V} = \begin{pmatrix} 16 & 12 \\ 12 & 9 \end{pmatrix} \quad (1.0.1)$$

2)Find the vertex  $\mathbf{c}$  of the parabola using SVD for

$$16x^2 + 24xy + 9y^2 - 5x - 10y + 1 = 0 \quad (1.0.2)$$

by changing  $\eta$  to  $\eta/2$  also verify the result using least squares.

## 2 SOLUTION

### 2.1 Part 1:QR Decomposition of V

Let  $\mathbf{x}$  and  $\mathbf{y}$  be the column vectors of the given matrix.

$$\mathbf{x} = \begin{pmatrix} 16 \\ 12 \end{pmatrix} \quad (2.1.1)$$

$$\mathbf{y} = \begin{pmatrix} 12 \\ 9 \end{pmatrix} \quad (2.1.2)$$

The column vectors can be expressed as ,

$$\mathbf{x} = k_1 \mathbf{u}_1 \quad (2.1.3)$$

$$\mathbf{y} = r_1 \mathbf{u}_1 + k_2 \mathbf{u}_2 \quad (2.1.4)$$

$$k_1 = \|\mathbf{x}\| \quad (2.1.5)$$

$$\mathbf{u}_1 = \frac{\mathbf{x}}{k_1} \quad (2.1.6)$$

$$r_1 = \frac{\mathbf{u}_1^T \mathbf{y}}{\|\mathbf{u}_1\|^2} \quad (2.1.7)$$

$$\mathbf{u}_2 = \frac{\mathbf{y} - r_1 \mathbf{u}_1}{\|\mathbf{y} - r_1 \mathbf{u}_1\|} \quad (2.1.8)$$

$$k_2 = \mathbf{u}_2^T \mathbf{y} \quad (2.1.9)$$

The (2.1.3) and (2.1.4) can be written as,

$$\begin{pmatrix} \mathbf{x} & \mathbf{y} \end{pmatrix} = \begin{pmatrix} \mathbf{u}_1 & \mathbf{u}_2 \end{pmatrix} \begin{pmatrix} k_1 & r_1 \\ 0 & k_2 \end{pmatrix} \quad (2.1.10)$$

$$\begin{pmatrix} \mathbf{x} & \mathbf{y} \end{pmatrix} = \mathbf{Q} \mathbf{R} \quad (2.1.11)$$

Now,  $\mathbf{R}$  is an upper triangular matrix and also,

$$\mathbf{Q}^T \mathbf{Q} = \mathbf{I} \quad (2.1.12)$$

Now using equations (2.1.5) to (2.1.9) we get,

$$k_1 = \sqrt{16^2 + 12^2} = 20 \quad (2.1.13)$$

$$\mathbf{u}_1 = \begin{pmatrix} \frac{4}{5} \\ \frac{3}{5} \end{pmatrix} \quad (2.1.14)$$

$$r_1 = \left( \frac{4}{5} \quad \frac{3}{5} \right) \begin{pmatrix} 12 \\ 9 \end{pmatrix} = 15 \quad (2.1.15)$$

$$\mathbf{u}_2 = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad (2.1.16)$$

$$k_2 = \begin{pmatrix} 0 & 0 \end{pmatrix} \begin{pmatrix} 12 \\ 9 \end{pmatrix} = 0 \quad (2.1.17)$$

Thus putting the values from (2.1.13) to (2.1.17) in (2.1.10) we obtain QR decomposition,

$$\begin{pmatrix} 16 & 12 \\ 12 & 9 \end{pmatrix} = \begin{pmatrix} \frac{4}{5} & 0 \\ \frac{3}{5} & 0 \end{pmatrix} \begin{pmatrix} 20 & 15 \\ 0 & 0 \end{pmatrix} \quad (2.1.18)$$

### 2.2 Part 2:Finding Vertex using SVD

$$\mathbf{V} = \begin{pmatrix} 16 & 12 \\ 12 & 9 \end{pmatrix} \quad (2.2.1)$$

$$\mathbf{u} = \begin{pmatrix} -\frac{5}{2} \\ -5 \end{pmatrix} \quad (2.2.2)$$

$$f = 1 \quad (2.2.3)$$

$$\mathbf{P} = \begin{pmatrix} \mathbf{p}_1 & \mathbf{p}_2 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} -3 & 4 \\ 4 & 3 \end{pmatrix} \quad (2.2.4)$$

$$\eta = \mathbf{u}^T \mathbf{p}_1 = -\frac{5}{2} \quad (2.2.5)$$

So the equation of perpendicular line passing through focus and intersecting parabola at vertex  $c$  is given as

$$\begin{pmatrix} \mathbf{u}^T + \frac{\eta}{2} \mathbf{p}_1^T \\ \mathbf{V} \end{pmatrix} \mathbf{c} = \begin{pmatrix} -f \\ \frac{\eta}{2} \mathbf{p}_1 - \mathbf{u} \end{pmatrix} \quad (2.2.6)$$

using (2.2.1),(2.2.2) ,(2.2.3) and (2.2.4)

$$\begin{pmatrix} -\frac{7}{4} & -6 \\ 16 & 12 \\ 12 & 9 \end{pmatrix} \mathbf{c} = \begin{pmatrix} -1 \\ \frac{13}{4} \\ 4 \end{pmatrix} \quad (2.2.7)$$

$$\mathbf{M}\mathbf{c} = \mathbf{b} \quad (2.2.8)$$

where

$$\mathbf{M} = \begin{pmatrix} -\frac{7}{4} & -6 \\ 16 & 12 \\ 12 & 9 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} -1 \\ \frac{13}{4} \\ 4 \end{pmatrix} \quad (2.2.9)$$

To solve (2.2.8), we perform singular value decomposition on  $\mathbf{M}$  given as

$$\mathbf{M} = \mathbf{U}\mathbf{S}\mathbf{V}^T \quad (2.2.10)$$

Substituting the value of  $\mathbf{M}$  from (2.2.10) in (2.2.8), we get

$$\mathbf{U}\mathbf{S}\mathbf{V}^T \mathbf{c} = \mathbf{b} \quad (2.2.11)$$

$$\Rightarrow \mathbf{c} = \mathbf{V}\mathbf{S}_+ \mathbf{U}^T \mathbf{b} \quad (2.2.12)$$

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

$$\mathbf{M}^T\mathbf{M} = \begin{pmatrix} \frac{6449}{16} & \frac{621}{2} \\ \frac{621}{2} & 261 \end{pmatrix} \quad (2.2.13)$$

$$\mathbf{M}\mathbf{M}^T = \begin{pmatrix} \frac{625}{16} & -100 & -75 \\ -100 & 400 & 300 \\ -75 & -300 & 225 \end{pmatrix} \quad (2.2.14)$$

Eigen values of  $\mathbf{M}^T\mathbf{M}$  can be found out as

$$|\mathbf{M}^T\mathbf{M} - \lambda\mathbf{I}| = 0 \quad (2.2.15)$$

$$\begin{pmatrix} \frac{6449}{16} - \lambda & \frac{621}{2} \\ \frac{621}{2} & 261 - \lambda \end{pmatrix} = 0 \quad (2.2.16)$$

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

$$\lambda_1 = 13.51 \quad (2.2.17)$$

$$\lambda_2 = 650.6 \quad (2.2.18)$$

$$(2.2.19)$$

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

$$\mathbf{v}_1 = \begin{pmatrix} -0.7971 \\ 1 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} 1.255 \\ 1 \end{pmatrix} \quad (2.2.20)$$

Normalizing the eigen vectors we get,

$$\mathbf{v}_1 = \begin{pmatrix} -0.6233 \\ 0.782 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} 0.7816 \\ 0.6228 \end{pmatrix} \quad (2.2.21)$$

Hence we obtain  $\mathbf{V}$  of (2.2.10) as follows,

$$\mathbf{V} = \begin{pmatrix} -0.6233 & 0.7816 \\ 0.782 & 0.6228 \end{pmatrix} \quad (2.2.22)$$

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

$$|\mathbf{M}\mathbf{M}^T - \lambda\mathbf{I}| = 0 \quad (2.2.23)$$

$$\begin{pmatrix} \frac{625}{16} - \lambda & -100 & -75 \\ -100 & 400 - \lambda & 300 \\ -75 & 300 & 225 - \lambda \end{pmatrix} = 0 \quad (2.2.24)$$

$$\lambda_3 = 13.51 \quad (2.2.25)$$

$$\lambda_4 = 650.6 \quad (2.2.26)$$

$$\lambda_5 = 0 \quad (2.2.27)$$

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

$$\mathbf{u}_1 = \begin{pmatrix} 8.153 \\ 1.333 \\ 1 \end{pmatrix}, \mathbf{u}_2 = \begin{pmatrix} -0.3407 \\ 1.333 \\ 1 \end{pmatrix}, \mathbf{u}_3 = \begin{pmatrix} 0 \\ -0.75 \\ 1 \end{pmatrix} \quad (2.2.28)$$

Normalizing the eigen vectors we get,

$$\mathbf{u}_1 = \begin{pmatrix} 0.9798 \\ 0.1601 \\ 0.1201 \end{pmatrix}, \mathbf{u}_2 = \begin{pmatrix} -0.2 \\ 0.7841 \\ 0.5882 \end{pmatrix}, \mathbf{u}_3 = \begin{pmatrix} 0 \\ -0.6 \\ 0.8 \end{pmatrix} \quad (2.2.29)$$

Hence we obtain  $\mathbf{U}$  of (2.2.10) as follows,

$$\mathbf{U} = \begin{pmatrix} 0.9798 & -0.2 & 0 \\ 0.1601 & 0.7841 & -0.6 \\ 0.1201 & 0.5882 & 0.8 \end{pmatrix} \quad (2.2.30)$$

After computing the singular values from eigen values  $\lambda_3, \lambda_4, \lambda_5$  we get  $\mathbf{S}$  of (2.2.10) as follows,

$$\mathbf{S} = \begin{pmatrix} 3.675 & 0 \\ 0 & 25.5 \\ 0 & 0 \end{pmatrix} \quad (2.2.31)$$

From (2.2.10) we get the Singular Value Decompo-

sition of  $\mathbf{M}$ ,

$$\mathbf{M} = \begin{pmatrix} 0.9798 & -0.2 & 0 \\ 0.1601 & 0.7841 & -0.6 \\ 0.1201 & 0.5882 & 0.8 \end{pmatrix} \begin{pmatrix} 3.675 & 0 \\ 0 & 25.5 \\ 0 & 0 \end{pmatrix} \quad (2.2.32)$$

$$\begin{pmatrix} -0.6233 & 0.7816 \\ 0.782 & 0.6228 \end{pmatrix}^T \quad (2.2.33)$$

$$= \begin{pmatrix} -1.75 & -6 \\ 16 & 12 \\ 12 & 9 \end{pmatrix} \quad (2.2.34)$$

Moore-Penrose Pseudo inverse of  $\mathbf{S}$  is given by,

$$\mathbf{S}_+ = \begin{pmatrix} 0.272 & 0 & 0 \\ 0 & 0.0391 & 0 \end{pmatrix} \quad (2.2.35)$$

From (2.2.12) we get,

$$\mathbf{U}^T \mathbf{b} = \begin{pmatrix} 0.02093 \\ 5.101 \\ 1.25 \end{pmatrix} \quad (2.2.36)$$

$$\mathbf{S}_+ \mathbf{U}^T \mathbf{b} = \begin{pmatrix} 0.005692 \\ 0.1995 \end{pmatrix} \quad (2.2.37)$$

$$\mathbf{c} = \mathbf{V} \mathbf{S}_+ \mathbf{U}^T \mathbf{b} = \begin{pmatrix} 0.16 \\ 0.12 \end{pmatrix} \quad (2.2.38)$$

$$\Rightarrow \mathbf{c} = \begin{pmatrix} 0.16 \\ 0.12 \end{pmatrix} \quad (2.2.39)$$

Verifying the solution of (2.2.39) using,

$$\mathbf{M}^T \mathbf{M} \mathbf{c} = \mathbf{M}^T \mathbf{b} \quad (2.2.40)$$

Evaluating the R.H.S in (2.2.40) we get,

$$\mathbf{M}^T \mathbf{M} \mathbf{c} = \begin{pmatrix} \frac{1018}{10} \\ 81 \end{pmatrix} \quad (2.2.41)$$

$$\Rightarrow \begin{pmatrix} \frac{6449}{16} & \frac{621}{2} \\ \frac{621}{2} & 261 \end{pmatrix} \mathbf{c} = \begin{pmatrix} \frac{1018}{10} \\ 81 \end{pmatrix} \quad (2.2.42)$$

Solving the augmented matrix of (2.2.42) we get,

$$\begin{pmatrix} \frac{6449}{16} & \frac{621}{2} & \frac{1018}{10} \\ \frac{621}{2} & 261 & 81 \end{pmatrix} \xleftrightarrow{R_1 = \frac{16}{6449} R_1} \begin{pmatrix} 1 & \frac{4968}{6449} & \frac{8144}{32245} \\ \frac{621}{2} & 261 & 81 \end{pmatrix} \quad (2.2.43)$$

$$\xleftrightarrow{R_2 = R_2 - \frac{621}{2} R_1} \begin{pmatrix} 1 & \frac{4968}{6449} & \frac{8144}{32245} \\ 0 & \frac{140625}{6449} & \frac{83133}{32245} \end{pmatrix} \quad (2.2.44)$$

$$\xleftrightarrow{R_2 = \frac{6449}{140625} R_2} \begin{pmatrix} 1 & \frac{4968}{6449} & \frac{8144}{32245} \\ 0 & 1 & \frac{3}{25} \end{pmatrix} \quad (2.2.45)$$

$$\xleftrightarrow{R_1 = R_1 - \frac{4968}{6449} R_2} \begin{pmatrix} 1 & 0 & \frac{197}{1250} \\ 0 & 1 & \frac{3}{25} \end{pmatrix} \quad (2.2.46)$$

From equation (2.2.46), solution is given by,

$$\mathbf{c} = \begin{pmatrix} \frac{197}{1250} \\ \frac{3}{25} \end{pmatrix} \quad (2.2.47)$$

$$\mathbf{c} = \begin{pmatrix} 0.16 \\ 0.12 \end{pmatrix} \quad (2.2.48)$$

Comparing results of  $\mathbf{c}$  from (2.2.39) and (2.2.48), we can say that the solution is verified.