

Problem 8

$$A = \begin{pmatrix} 10 & 6 \\ 0 & 8 \end{pmatrix} \frac{1}{\sqrt{10}} = \begin{pmatrix} \sqrt{10} & \frac{3\sqrt{2}}{\sqrt{5}} \\ 0 & \frac{4\sqrt{2}}{\sqrt{5}} \end{pmatrix} A = QS; Q = UV^\top; S = V\Sigma V^\top;$$

$$1.S = V\Sigma V^\top :$$

$$A^\top A = \begin{pmatrix} 10 & 6 \\ 6 & 10 \end{pmatrix}$$

$$\text{Eigenvalues}(A^\top A) :$$

$$\det \left(\begin{pmatrix} 10 & 6 \\ 6 & 10 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right) = 0;$$

$$\det \begin{pmatrix} 10 - & 6 \\ 6 & 10 - \end{pmatrix} = 0;$$

$$(10 -)(10 -) - 6 \cdot 6 = 0;$$

$$\lambda^2 - 20\lambda + 64 = 0;$$

$$\lambda_1 = \frac{-(-20) + \sqrt{(-20)^2 - 4 \cdot 1 \cdot 64}}{2 \cdot 1} = 16; \lambda_2 = \frac{-(-20) - \sqrt{(-20)^2 - 4 \cdot 1 \cdot 64}}{2 \cdot 1} = 4. \rightarrow$$

$$\rightarrow \sigma_1 = \sqrt{\lambda_1} = 4; \sigma_2 = \sqrt{\lambda_2} = 2.$$

$$\text{Eigenvector}(\lambda_1 = 16) :$$

$$\begin{pmatrix} 10 & 6 \\ 6 & 10 \end{pmatrix} - 16 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} -6 & 6 \\ 6 & -6 \end{pmatrix} R_2 \leftarrow R_2 + 1 \cdot R_1 = \begin{pmatrix} -6 & 6 \\ 0 & 0 \end{pmatrix} R_1 \leftarrow -\frac{1}{6} \cdot R_1 = \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \rightarrow x = y \rightarrow v_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \rightarrow \frac{v_1}{|v_1|} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\text{Eigenvector}(\lambda_2 = 4) :$$

$$\begin{pmatrix} 10 & 6 \\ 6 & 10 \end{pmatrix} - 4 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 6 & 6 \\ 6 & 6 \end{pmatrix} R_2 \leftarrow R_2 - 1 \cdot R_1 = \begin{pmatrix} 6 & 6 \\ 0 & 0 \end{pmatrix} R_1 \leftarrow \frac{1}{6} \cdot R_1 = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \rightarrow x = -y \rightarrow v_2 = \begin{pmatrix} -1 \\ 1 \end{pmatrix} \rightarrow \frac{v_2}{|v_2|} = \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

Thus,

$$V = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix}; \Sigma = \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix}.$$

$$S = V\Sigma V^\top = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix}$$

$$Q = UV^\top.$$

$$U :$$

$$u_1 = \frac{Av_1}{\sigma_1} = \begin{pmatrix} \sqrt{10} & \frac{3\sqrt{2}}{\sqrt{5}} \\ 0 & \frac{4\sqrt{2}}{\sqrt{5}} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} \frac{1}{4} = \begin{pmatrix} \frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{pmatrix}$$

$$u_2 = \frac{Av_2}{\sigma_2} = \begin{pmatrix} \sqrt{10} & \frac{3\sqrt{2}}{\sqrt{5}} \\ 0 & \frac{4\sqrt{2}}{\sqrt{5}} \end{pmatrix} \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} \frac{1}{2} = \begin{pmatrix} -\frac{1}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} \end{pmatrix}.$$

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

$$Q = UV^\top = \begin{pmatrix} \frac{2}{\sqrt{5}} & -\frac{1}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} \frac{3}{\sqrt{10}} & \frac{1}{\sqrt{10}} \\ -\frac{1}{\sqrt{10}} & \frac{3}{\sqrt{10}} \end{pmatrix}$$

Let's check values of S and Q :

$$A = QS = \begin{pmatrix} \frac{3}{\sqrt{10}} & \frac{1}{\sqrt{10}} \\ -\frac{1}{\sqrt{10}} & \frac{3}{\sqrt{10}} \end{pmatrix} \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} = \begin{pmatrix} \sqrt{10} & \frac{3\sqrt{2}}{\sqrt{5}} \\ 0 & \frac{4\sqrt{2}}{\sqrt{5}} \end{pmatrix} = A.$$

Problem 10

Matrix A

$$A = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix} \rightarrow A^\top A = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix}.$$

$$\text{Eigenvalues}(A^\top A) : \lambda_1 = 4, \lambda_2, \lambda_3, \lambda_4 = 0. \rightarrow \\ \rightarrow \sigma_1 = \sqrt{\lambda_1} = 2, \sigma_2, \sigma_3, \sigma_4 = 0$$

$$\text{Eigenvector}(\lambda_1 = 4) : (1, 1, 1, 1)^\top = v_1 \rightarrow \frac{\vec{v}_1}{\|v_1\|} = \frac{(1,1,1,1)^\top}{2} = \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)^\top$$

$\text{Eigenvectors}(\lambda_2 = 0) :$

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix} ((R_2, R_3, R_4) \leftarrow (R_2, R_3, R_4) - 1 \cdot R_1) = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}.$$

$$(A - 0I) \begin{pmatrix} x \\ y \\ z \\ v \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \\ v \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\rightarrow x + y + z + v = 0;$$

$$x = -y - z - v.$$

$$\rightarrow \begin{pmatrix} -y - z - v \\ y \\ z \\ v \end{pmatrix} = \begin{pmatrix} -y \\ y \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} -z \\ 0 \\ z \\ 0 \end{pmatrix} + \begin{pmatrix} -v \\ 0 \\ 0 \\ v \end{pmatrix} \rightarrow$$

For $y, z, v = 1 :$

$$\begin{aligned}
v_2 &= \begin{pmatrix} -1 \\ 1 \\ 0 \\ 0 \end{pmatrix} \rightarrow \frac{\vec{v}_2}{\|\vec{v}_2\|} = \frac{(-1,1,0,0)^\top}{\sqrt{2}} = \begin{pmatrix} -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 & 0 \end{pmatrix}^\top \\
v_3 &= \begin{pmatrix} -1 \\ 0 \\ 1 \\ 0 \end{pmatrix} \rightarrow \frac{\vec{v}_3}{\|\vec{v}_3\|} = \frac{(-1,0,1,0)^\top}{\sqrt{2}} = \begin{pmatrix} -\frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} & 0 \end{pmatrix}^\top \\
v_4 &= \begin{pmatrix} -1 \\ 0 \\ 0 \\ 1 \end{pmatrix} \rightarrow \frac{\vec{v}_4}{\|\vec{v}_4\|} = \frac{(-1,0,0,1)^\top}{\sqrt{2}} = \begin{pmatrix} -\frac{1}{\sqrt{2}} & 0 & 0 & \frac{1}{\sqrt{2}} \end{pmatrix}^\top \\
\rightarrow V &= \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 & 0 \\ \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & 0 & 0 & \frac{1}{\sqrt{2}} \end{pmatrix} \\
\rightarrow u_1 &= \frac{Av_1}{\sigma_1} = \frac{1}{\sigma_1} \begin{pmatrix} 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2} \end{pmatrix}^\top = 1. \\
\rightarrow A &= U\Sigma V^\top = [1] \begin{pmatrix} 2 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} & 0 \\ -\frac{1}{\sqrt{2}} & 0 & 0 & \frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 \end{pmatrix}. \\
A^+ &= V\Sigma^+U^\top = \begin{pmatrix} \frac{1}{2} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{2} & \frac{1}{\sqrt{2}} & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{2} & 0 & 0 & \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \frac{1}{2} \\ 0 \\ 0 \\ 0 \end{pmatrix} (1) = \begin{pmatrix} \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \end{pmatrix}
\end{aligned}$$

Matrix B

$$\begin{aligned}
B &= \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \rightarrow B^\top B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \\
\text{Eigenvalues}(B^\top B) : \lambda_1, \lambda_2 &= 1, \lambda_3 = 0. \rightarrow \\
\rightarrow \sigma_1 &= 1, \sigma_2 = 1, \sigma_3 = 0. \\
\text{Eigenvectors}(\lambda_1 = 1) : \\
v_1 &= \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}; v_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}; \\
\text{Eigenvector}(\lambda_2 = 0) : \\
v_3 &= \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}.
\end{aligned}$$

$$V = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

$$u_1 = \frac{Bv_1}{\sigma_1} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}; u_2 = \frac{Bv_2}{\sigma_2} = \begin{pmatrix} 1 \\ 0 \end{pmatrix};$$

$$U = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

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

Thus,

$$B = U\Sigma V^\top = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} = B$$

$$B^+ = V\Sigma^+U^\top = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \\ 0 & 0 \end{pmatrix}$$

Matrix C

$$C = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}.$$

$$C^\top C = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}.$$

$$\text{Eigenvalues}(C^\top C) : \lambda_1 = 2, \lambda_2 = 0 \rightarrow \\ \rightarrow \sigma_1 = \sqrt{2}, \sigma_2 = 0$$

$$\text{Eigenvector}(\lambda_1 = 2) : v_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}; \rightarrow \frac{\vec{v}_1}{\|v_1\|} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}.$$

$$\text{Eigenvector}(\lambda_2 = 0) : v_2 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}. \rightarrow \frac{\vec{v}_2}{\|v_2\|} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$$V = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$$u_1 = \frac{Cv_1}{\sigma_1} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$u_2 = 0 \text{ because } \|Cv_2\| = \sigma_2 = 0.$$

So, we need to find one non-zero u_2 , orthogonal to $u_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \rightarrow u_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$.

$$C = U\Sigma V^\top = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} = C.$$

$$C^+ = V\Sigma^+U^\top = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 \\ \frac{1}{\sqrt{2}} & 0 \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & 0 \\ \frac{1}{2} & 0 \end{pmatrix}.$$

Problem 12

(a)

$$A = \begin{pmatrix} 4 & 2 & 2 \\ 2 & 10 & 10 \\ 2 & 10 & 14 \end{pmatrix}$$

Lets use the leading principal minors of matrix A to determine whether A is a positive definite matrix.

For A to be positive definite all principal minors D_i should be: $D_1 > 0, D_2 > 0, D_3 > 0$:

$$D_1 = 4 \rightarrow D_1 > 0;$$

$$D_2 = \det \begin{pmatrix} 4 & 2 \\ 2 & 10 \end{pmatrix} = 36 \rightarrow D_2 > 0;$$

$$D_3 = \det(A) = 4 \cdot \det \begin{pmatrix} 10 & 10 \\ 10 & 14 \end{pmatrix} - 2 \cdot \det \begin{pmatrix} 2 & 10 \\ 2 & 14 \end{pmatrix} + 2 \cdot \det \begin{pmatrix} 2 & 10 \\ 2 & 10 \end{pmatrix} = 144 \rightarrow D_3 > 0.$$

We have $D_1 > 0, D_2 > 0, D_3 > 0 \rightarrow$ We proved without finding eigenvalues that matrix A is positive definite.

(b)

Matrix A is symmetric as $A = A^T$. That means that in LDL^T decomposition of A we have $U = L^T$. Before finding LDL^T we shall firstly find LU decomposition for A : $A = LU$:

$$U = \begin{pmatrix} 4 & 2 & 2 \\ 2 & 10 & 10 \\ 2 & 10 & 14 \end{pmatrix} R_2 \leftarrow R_2 - \frac{1}{2} \cdot R_1 = \begin{pmatrix} 4 & 2 & 2 \\ 0 & 9 & 9 \\ 2 & 10 & 14 \end{pmatrix} R_3 \leftarrow R_3 - \frac{1}{2} \cdot R_1 =$$

$$\begin{pmatrix} 4 & 2 & 2 \\ 0 & 9 & 9 \\ 0 & 9 & 13 \end{pmatrix} R_3 \leftarrow R_3 - 1 \cdot R_2 = \begin{pmatrix} 4 & 2 & 2 \\ 0 & 9 & 9 \\ 0 & 0 & 4 \end{pmatrix} \rightarrow$$

$$\rightarrow l(2, 1) = -(-\frac{1}{2}) = \frac{1}{2}; l(3, 1) = -(-\frac{1}{2}) = \frac{1}{2}; l(3, 2) = -(-1) = 1.$$

Thus,

$$L = \begin{pmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 1 & 1 \end{pmatrix} \rightarrow$$

$$\rightarrow A = LU = \begin{pmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 1 & 1 \end{pmatrix} \begin{pmatrix} 4 & 2 & 2 \\ 0 & 9 & 9 \\ 0 & 0 & 4 \end{pmatrix} \rightarrow \text{checking} \rightarrow \begin{pmatrix} 4 & 2 & 2 \\ 2 & 10 & 10 \\ 2 & 10 & 14 \end{pmatrix} =$$

$= A.$

$$\text{In } LDL^T \rightarrow D = \begin{pmatrix} 4 & 0 & 0 \\ 0 & 9 & 0 \\ 0 & 0 & 4 \end{pmatrix}; L^T = \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{2} \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\begin{aligned} \text{Thus, } LDL^\top &= \begin{pmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 1 & 1 \end{pmatrix} \begin{pmatrix} 4 & 0 & 0 \\ 0 & 9 & 0 \\ 0 & 0 & 4 \end{pmatrix} \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{2} \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \rightarrow \text{checking} \rightarrow \\ &\rightarrow \begin{pmatrix} 4 & 2 & 2 \\ 2 & 10 & 10 \\ 2 & 10 & 14 \end{pmatrix} = A. \end{aligned}$$

(c)

We can get a Cholesky decomposition by evenly splitting square root of diagonal elements of D between L and L^\top :

For L :

$$\begin{pmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 1 & 1 \end{pmatrix} \begin{pmatrix} \sqrt{4} & 0 & 0 \\ 0 & \sqrt{9} & 0 \\ 0 & 0 & \sqrt{4} \end{pmatrix} = \begin{pmatrix} 2 & 0 & 0 \\ 1 & 3 & 0 \\ 1 & 3 & 2 \end{pmatrix};$$

For L^\top :

$$\begin{pmatrix} \sqrt{4} & 0 & 0 \\ 0 & \sqrt{9} & 0 \\ 0 & 0 & \sqrt{4} \end{pmatrix} \begin{pmatrix} 1 & \frac{1}{2} & \frac{1}{2} \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 1 & 1 \\ 0 & 3 & 3 \\ 0 & 0 & 2 \end{pmatrix}.$$

$$\rightarrow A = LL^\top = \begin{pmatrix} 2 & 0 & 0 \\ 1 & 3 & 0 \\ 1 & 3 & 2 \end{pmatrix} \begin{pmatrix} 2 & 1 & 1 \\ 0 & 3 & 3 \\ 0 & 0 & 2 \end{pmatrix} \rightarrow \text{checking} \rightarrow \begin{pmatrix} 4 & 2 & 2 \\ 2 & 10 & 10 \\ 2 & 10 & 14 \end{pmatrix} = A.$$

(d)

$$B = \begin{pmatrix} 1 & -4 & 2 \\ -4 & 1 & -2 \\ 2 & -2 & -2 \end{pmatrix}.$$

Let's recall matrix definiteness dependence on value of its principal minors and respective criterions:

- 1) For positive definite: $D_1 > 0, D_2 > 0, D_3 > 0$;
- 2) Negative definite: $D_1 < 0, D_2 > 0, D_3 < 0$;
- 3) Positive semidefinite: $D_1 \geq 0, D_2 \geq 0, D_3 \geq 0$ for all principal minors;
- 4) Negative semidefinite: $D_1 \leq 0, D_2 \geq 0, D_3 \leq 0$.

For matrix B its principal minors are:

$D_1 = 1; D_2 = -15$. \rightarrow without computing D_3 we see that combination of D_1, D_2 of matrix B doesn't match any of above-mentioned criterions and this means that matrix B is indefinite.

LU decomposition:

$$\begin{pmatrix} 1 & -4 & 2 \\ -4 & 1 & -2 \\ 2 & -2 & -2 \end{pmatrix} R_2 \leftarrow R_2 + 4 \cdot R_1 = \begin{pmatrix} 1 & -4 & 2 \\ 0 & -15 & 6 \\ 2 & -2 & -2 \end{pmatrix} R_3 \leftarrow R_3 + (-2) \cdot R_1 =$$

$$\begin{aligned}
& \begin{pmatrix} 1 & -4 & 2 \\ 0 & -15 & 6 \\ 0 & 6 & -6 \end{pmatrix} R_3 \leftarrow R_3 + \left(\frac{2}{5}\right) \cdot R_2 = \begin{pmatrix} 1 & -4 & 2 \\ 0 & -15 & 6 \\ 0 & 0 & -\frac{18}{5} \end{pmatrix} = U \rightarrow \\
& \rightarrow L = \begin{pmatrix} 1 & 0 & 0 \\ -4 & 1 & 0 \\ 2 & -\frac{2}{5} & -6 \end{pmatrix}; \\
& LU = \begin{pmatrix} 1 & 0 & 0 \\ -4 & 1 & 0 \\ 2 & -\frac{2}{5} & 1 \end{pmatrix} \begin{pmatrix} 1 & -4 & 2 \\ 0 & -15 & 6 \\ 0 & 0 & -\frac{18}{5} \end{pmatrix} \rightarrow \text{checking} \rightarrow \begin{pmatrix} 1 & -4 & 2 \\ -4 & 1 & -2 \\ 2 & -2 & -2 \end{pmatrix} = \\
& = B \rightarrow D = \begin{pmatrix} 1 & 0 & 0 \\ 0 & -15 & 0 \\ 0 & 0 & -\frac{18}{5} \end{pmatrix} \rightarrow \\
& \rightarrow LDL^T = \begin{pmatrix} 1 & 0 & 0 \\ -4 & 1 & 0 \\ 2 & -\frac{2}{5} & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & -15 & 0 \\ 0 & 0 & -\frac{18}{5} \end{pmatrix} \begin{pmatrix} 1 & -4 & 2 \\ 0 & 1 & -\frac{2}{5} \\ 0 & 0 & 1 \end{pmatrix} \rightarrow \text{checking} \rightarrow \\
& \begin{pmatrix} 1 & -4 & 2 \\ -4 & 1 & -2 \\ 2 & -2 & -2 \end{pmatrix} = B.
\end{aligned}$$

Problem 11

$$A = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix}$$

$$b = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$$

$$A^+ = V\Sigma^+U^T$$

The shortest solution $x' = (V\Sigma^+U^T)b$

$$\rightarrow A^T A = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 3 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}.$$

Eigenvalues($A^T A$):

$$\det \left(\begin{pmatrix} 3 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right) = 0;$$

$$-\lambda^3 + 5\lambda^2 - 4\lambda = 0;$$

$$-\lambda(\lambda-1)(\lambda-4) = 0; \rightarrow \lambda_1 = 0; \lambda_2 = 1; \lambda_3 = 4;$$

Eigenvector($\lambda_1 = 0$):

$$(A - \lambda I) : \begin{pmatrix} 3 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} - 0 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 3 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \rightarrow \text{via-row-reduction} \rightarrow$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow v_1 = (A - 0I) \begin{pmatrix} x \\ y \\ z \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} x = 0 \\ y = -z \end{cases} \rightarrow$$

$$v_1 = \begin{pmatrix} 0 \\ -z \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix}$$

Eigenvector($\lambda_2 = 1$) :

$$\begin{pmatrix} 3 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} - 1 \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} \rightarrow \text{via - row - reduction} \rightarrow$$

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow v_2 = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} x = -z \\ y = z \end{cases} \rightarrow$$

$$v_2 = \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$

Eigenvector(λ_3) :

$$\begin{pmatrix} 3 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} - 4 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} -1 & 1 & 1 \\ 1 & -3 & 1 \\ 1 & 1 & -3 \end{pmatrix} \rightarrow \text{via - row - reduction} \rightarrow$$

$$\begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow v_3 = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} x = 2z \\ y = z \end{cases} \rightarrow v_3 = \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}$$

$$\frac{\vec{v}_1}{\|v_1\|} = \begin{pmatrix} 0 \\ -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\frac{\vec{v}_2}{\|v_2\|} = \begin{pmatrix} -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}$$

$$\frac{\vec{v}_3}{\|v_3\|} = \begin{pmatrix} \sqrt{\frac{2}{3}} \\ \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \end{pmatrix}$$

Let's sort eigenvalues from in descending order and arrange respective eigenvectors, so that λ_3 interchange with λ_1 and the same goes for eigenvectors v_1 and v_3 :

$$\lambda_1 = 4; \lambda_2 = 1; \lambda_3 = 0; \rightarrow \sigma_1 = 2; \sigma_2 = 1; \sigma_3 = 0.$$

$$V = \begin{pmatrix} \sqrt{\frac{2}{3}} & -\frac{1}{\sqrt{3}} & 0 \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$u_1 = \frac{Av_1}{\sigma_1} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} \sqrt{\frac{2}{3}} \\ \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \end{pmatrix} \frac{1}{2} = \begin{pmatrix} \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \\ \sqrt{\frac{2}{3}} \end{pmatrix}$$

$$u_2 = \frac{Av_2}{\sigma_2} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix} = \begin{pmatrix} -\frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}.$$

We need to find vector u_3 orthogonal to both u_1 and u_2 to form a matrix U of SVD decomposition. To satisfy these two orthogonality conditions, the vector u_3 must be a solution of the homogeneous linear system:

$$\begin{pmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} & \sqrt{\frac{2}{3}} \\ -\frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow \text{reduced - row - echelon - form} \rightarrow$$

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow x_3 = 0; x_1 = -x_2 \rightarrow u_3 = \begin{pmatrix} -x_2 \\ x_2 \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} \rightarrow$$

$$\frac{\vec{u}_3}{\|\vec{u}_3\|} = \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix}$$

Let's check orthogonality:

$$u_2 \cdot u_3 = \begin{pmatrix} -\frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \end{pmatrix} \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix} = (0) = 0$$

$$u_1 \cdot u_3 = \begin{pmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} & \sqrt{\frac{2}{3}} \end{pmatrix} \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix} = 0 \rightarrow$$

$$\rightarrow u_3 \text{ is orthogonal to } u_1 \text{ and } u_2. \rightarrow U = \begin{pmatrix} \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \\ \sqrt{\frac{2}{3}} & \frac{1}{\sqrt{3}} & 0 \end{pmatrix}$$

$$\Sigma = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\text{Let's check} \rightarrow A = U\Sigma V^T = \begin{pmatrix} \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \\ \sqrt{\frac{2}{3}} & \frac{1}{\sqrt{3}} & 0 \end{pmatrix} \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} \sqrt{\frac{2}{3}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} \\ -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ 0 & -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} =$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix} = A \rightarrow \text{we performed SVD rightly. Now let's find pseudo-inverse}$$

in order to find minimum length solution:

$$\begin{aligned}
A^+ &= V\Sigma^+U^\top = \begin{pmatrix} \sqrt{\frac{2}{3}} & -\frac{1}{\sqrt{3}} & 0 \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} & \sqrt{\frac{2}{3}} \\ -\frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \end{pmatrix} = \\
&\begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ -\frac{1}{4} & -\frac{1}{4} & \frac{1}{2} \\ -\frac{1}{4} & -\frac{1}{4} & \frac{1}{2} \end{pmatrix} \cdot \rightarrow x' = (V\Sigma^+U^\top)b = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ -\frac{1}{4} & -\frac{1}{4} & \frac{1}{2} \\ -\frac{1}{4} & -\frac{1}{4} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \\ \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \rightarrow \\
&\rightarrow \text{minimum-length least square solution.}
\end{aligned}$$

Problem 3

(a)

$$A = U\Sigma V^\top \rightarrow A^\top = (U\Sigma V^\top)^\top = V\Sigma^\top U^\top.$$

After transposing Σ , in Σ^\top we have that main diagonal remains the same as in Σ , while other entries are zero. We also know that matrices V and U are orthogonal. As a conclusion, $V\Sigma^\top U^\top$ is an SVD of A .

(b)

$$\begin{aligned}
A &= uv^\top = (\|u\| \frac{u}{\|u\|})(\|v\| \frac{v}{\|v\|})^\top = (\|u\|\|v\|) \frac{u}{\|u\|} \frac{v}{\|v\|}^\top = \sigma \hat{u} \hat{v}^\top \\
&\text{, where} \\
\sigma &= \|u\|\|v\|; \\
\hat{u} &= \frac{u}{\|u\|}; \\
\hat{v} &= \frac{v}{\|v\|}.
\end{aligned}$$

(c)

(i)

$$A = \begin{pmatrix} 0 & 1 & 2 \end{pmatrix} \rightarrow A = U\Sigma V^\top.$$

$$A^\top A = \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix} \begin{pmatrix} 0 & 1 & 2 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 4 \end{pmatrix}$$

$$\text{Eigenvalues } (A^\top A): \det \left(\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 4 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right) = 0;$$

$$\det \begin{pmatrix} -\lambda & 0 & 0 \\ 0 & 1-\lambda & 2 \\ 0 & 2 & 4-\lambda \end{pmatrix} = 0;$$

$$- \det \begin{pmatrix} 1- & 2 \\ 2 & 4- \end{pmatrix} - 0 \cdot \det \begin{pmatrix} 0 & 2 \\ 0 & 4- \end{pmatrix} + 0 \cdot \det \begin{pmatrix} 0 & 1- \\ 0 & 2 \end{pmatrix} = 0;$$

$$- \lambda (\lambda^2 - 5\lambda) = 0;$$

$$- \lambda^2 (\lambda - 5) = 0 \rightarrow \lambda_1 = 5; \lambda_2, \lambda_3 = 0. \rightarrow$$

$$\rightarrow \sigma_1 = \sqrt{5}; \sigma_2 = 0; \sigma_3 = 0.$$

Eigenvector($\lambda_1 = 5$) :

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 4 \end{pmatrix} - 5 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} -5 & 0 & 0 \\ 0 & -4 & 2 \\ 0 & 2 & -1 \end{pmatrix} R_3 \leftarrow R_3 + \frac{1}{2} R_2 = \begin{pmatrix} -5 & 0 & 0 \\ 0 & -4 & 2 \\ 0 & 0 & 0 \end{pmatrix} R_2 \leftarrow$$

$$-\frac{1}{4} \cdot R_2 = \begin{pmatrix} -5 & 0 & 0 \\ 0 & 1 & -\frac{1}{2} \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow -\frac{1}{5} \cdot R_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & -\frac{1}{2} \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{cases} x = 0 \\ y = \frac{1}{2}z \end{cases} \rightarrow$$

$$v_1 = \begin{pmatrix} 0 \\ \frac{1}{2} \\ 1 \end{pmatrix} \rightarrow \frac{v_1}{|v_1|} = \begin{pmatrix} 0 \\ \frac{1}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} \end{pmatrix}$$

Eigenvectors($\lambda_2 = 0$) :

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 4 \end{pmatrix} - 0 \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 4 \end{pmatrix} R_2 \leftrightarrow R_3 = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 2 & 4 \\ 0 & 1 & 2 \end{pmatrix} R_3 \leftarrow$$

$$R_3 - \frac{1}{2} \cdot R_2 = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 2 & 4 \\ 0 & 0 & 0 \end{pmatrix} R_2 \leftarrow \frac{1}{2} \cdot R_2 = \begin{pmatrix} 0 & 1 & 2 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow y = -2z$$

$$\rightarrow v_2 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}; \rightarrow \frac{v_2}{|v_2|} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}; v_3 = \begin{pmatrix} 0 \\ -2 \\ 1 \end{pmatrix} \rightarrow \frac{v_3}{|v_3|} = \begin{pmatrix} 0 \\ -\frac{2}{\sqrt{5}} \\ \frac{1}{\sqrt{5}} \end{pmatrix}$$

$$u_1 = \frac{Av_1}{\sigma_1} = (0 \quad 1 \quad 2) \begin{pmatrix} 0 \\ \frac{1}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} \end{pmatrix} \frac{1}{\sqrt{5}} = 1 = U.$$

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

$$A = U\Sigma V^\top = 1 \begin{pmatrix} \sqrt{5} & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \\ 1 & 0 & 0 \\ 0 & -\frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{pmatrix} = (0 \quad 1 \quad 2) = A.$$

(ii)

$$A = (0 \quad 1 \quad 2)^\top$$

Matrix from (ii) is a transpose matrix of matrix from (i). As we know: $A^\top = (U\Sigma V^\top)^\top = V\Sigma^\top U^\top$. \rightarrow we can conclude that SVD of (ii) is:

$$(1 \cdot (\sqrt{5} \quad 0 \quad 0) \begin{pmatrix} 0 & \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \\ 1 & 0 & 0 \\ 0 & -\frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{pmatrix})^\top = \begin{pmatrix} 0 & 1 & 0 \\ \frac{1}{\sqrt{5}} & 0 & -\frac{2}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} & 0 & \frac{1}{\sqrt{5}} \end{pmatrix} \begin{pmatrix} \sqrt{5} \\ 0 \\ 0 \end{pmatrix} \cdot 1 = \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix} = A$$

(iii)

$$A = \begin{pmatrix} 2 & 1 & -2 \\ -2 & -1 & 2 \end{pmatrix} \rightarrow A^\top A = \begin{pmatrix} 8 & 4 & -8 \\ 4 & 2 & -4 \\ -8 & -4 & 8 \end{pmatrix}$$

Eigenvalues($A^\top A$):

$$\det \left(\begin{pmatrix} 8 & 4 & -8 \\ 4 & 2 & -4 \\ -8 & -4 & 8 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right) = 0;$$

$$\det \begin{pmatrix} 8-\lambda & 4 & -8 \\ 4 & 2-\lambda & -4 \\ -8 & -4 & 8-\lambda \end{pmatrix} = 0;$$

$$(8-\lambda)(\lambda^2 - 10\lambda) - 4(-4\lambda) - 8(-8\lambda) = 0;$$

$$-\lambda^2(\lambda - 18) = 0; \rightarrow \lambda_1 = 18, \lambda_2, \lambda_3 = 0; \rightarrow$$

$$\rightarrow \sigma_1 = \sqrt{18}; \sigma_2 = 0; \sigma_3 = 0.$$

$$\text{Eigenvector}(\lambda_1 = 18): \begin{pmatrix} 8 & 4 & -8 \\ 4 & 2 & -4 \\ -8 & -4 & 8 \end{pmatrix} - 18 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} -10 & 4 & -8 \\ 4 & -16 & -4 \\ -8 & -4 & -10 \end{pmatrix} R_2 \leftarrow$$

$$R_2 + \frac{2}{5} \cdot R_1 = \begin{pmatrix} -10 & 4 & -8 \\ 0 & -\frac{72}{5} & -\frac{36}{5} \\ -8 & -4 & -10 \end{pmatrix} R_3 \leftarrow R_3 - \frac{4}{5} \cdot R_1 = \begin{pmatrix} -10 & 4 & -8 \\ 0 & -\frac{72}{5} & -\frac{36}{5} \\ 0 & -\frac{36}{5} & -\frac{18}{5} \end{pmatrix} R_3 \leftarrow$$

$$R_3 - \frac{1}{2} \cdot R_2 = \begin{pmatrix} -10 & 4 & -8 \\ 0 & -\frac{72}{5} & -\frac{36}{5} \\ 0 & 0 & 0 \end{pmatrix} R_2 \leftarrow -\frac{5}{72} \cdot R_2 = \begin{pmatrix} -10 & 4 & -8 \\ 0 & 1 & \frac{1}{2} \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow$$

$$R_1 - 4 \cdot R_2 = \begin{pmatrix} -10 & 0 & -10 \\ 0 & 1 & \frac{1}{2} \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow -\frac{1}{10} \cdot R_1 = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & \frac{1}{2} \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{cases} x = -z \\ y = -\frac{1}{2}z \end{cases} \rightarrow$$

$$v_1 = \begin{pmatrix} -1 \\ -\frac{1}{2} \\ 1 \end{pmatrix} \rightarrow \frac{v_1}{|v_1|} = \begin{pmatrix} -\frac{2}{3} \\ -\frac{1}{3} \\ \frac{2}{3} \end{pmatrix}$$

Eigenvectors($\lambda_2 = 0$):

$$\begin{pmatrix} 8 & 4 & -8 \\ 4 & 2 & -4 \\ -8 & -4 & 8 \end{pmatrix} - 0 \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 8 & 4 & -8 \\ 4 & 2 & -4 \\ -8 & -4 & 8 \end{pmatrix} \rightarrow \text{via row-reduction}$$

$$\rightarrow \begin{pmatrix} 1 & \frac{1}{2} & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow x + \frac{1}{2}y - z = 0 \rightarrow \begin{pmatrix} -\frac{1}{2}y \\ y \\ 0 \end{pmatrix} + \begin{pmatrix} z \\ 0 \\ z \end{pmatrix} \rightarrow v_2 = \begin{pmatrix} -\frac{1}{2} \\ 1 \\ 0 \end{pmatrix} \rightarrow$$

$$\frac{v_2}{|v_2|} = \begin{pmatrix} -\frac{1}{\sqrt{5}} \\ \frac{2}{\sqrt{5}} \\ 0 \end{pmatrix} v_3 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \rightarrow \frac{v_3}{|v_3|} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\begin{aligned}
V &= \begin{pmatrix} -\frac{2}{3} & -\frac{1}{\sqrt{5}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{3} & \frac{2}{\sqrt{5}} & 0 \\ \frac{2}{3} & 0 & \frac{1}{\sqrt{2}} \end{pmatrix} \\
\Sigma &= \begin{pmatrix} 5 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \\
u_1 = \frac{Av_1}{\sigma_1} &= \begin{pmatrix} 2 & 1 & -2 \\ -2 & -1 & 2 \end{pmatrix} \begin{pmatrix} -\frac{2}{3} \\ -\frac{1}{3} \\ \frac{2}{3} \end{pmatrix} \frac{1}{\sqrt{18}} = \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} \\
u_2 \text{ should be orthogonal to } u_1: u_2 &= \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} \\
U &= \begin{pmatrix} -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} \\
A = U\Sigma V^\top &= \begin{pmatrix} -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \sqrt{18} & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} -\frac{2}{3} & -\frac{1}{\sqrt{5}} & \frac{2}{3} \\ -\frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} & 0 \\ \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} -3 & 0 & 0 \\ 3 & 0 & 0 \end{pmatrix} \begin{pmatrix} -\frac{2}{3} & -\frac{1}{\sqrt{5}} & \frac{2}{3} \\ -\frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} & 0 \\ \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \end{pmatrix} = \\
&= \begin{pmatrix} 2 & 1 & -2 \\ -2 & -1 & 2 \end{pmatrix} = A.
\end{aligned}$$

Problem 2

(a)

From Problem 3(i) we already concluded that $V\Sigma^\top U^\top$ is an SVD of A^\top . Since after transposing A , we have that Σ^\top contains the same main diagonal as Σ , non-zero singular values of A and A^\top are the same.

(b.i)

From Problem 3(i) we conclude that singular values of (i) are:
 $\sigma_1 = \sqrt{5}; \sigma_2 = 0; \sigma_3 = 0$.

(b.ii)

From (a) and Problem 3(i) we concluded that non-zero singular values of (i) and (ii) are the same: $\sigma_1 = \sqrt{5}$.

However here we wouldn't have 3 by 3 matrix from $A^T A$ as in (b.ii), but only 1 by 1 from AA^T . That's why we don't have zero singular values here.

(b.iii)

$$A = \begin{pmatrix} 1 & 2 & 2 \\ 1 & 3 & 3 \end{pmatrix} \rightarrow A^\top A = \begin{pmatrix} 2 & 5 & 5 \\ 5 & 13 & 13 \\ 5 & 13 & 13 \end{pmatrix}$$

Eigenvalues($A^\top A$):

$$\det \left(\begin{pmatrix} 2 & 5 & 5 \\ 5 & 13 & 13 \\ 5 & 13 & 13 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right) = 0;$$

$$\det \begin{pmatrix} 2-\lambda & 5 & 5 \\ 5 & 13-\lambda & 13 \\ 5 & 13 & 13-\lambda \end{pmatrix} = 0;$$

$$(2-\lambda)(\lambda^2 - 26\lambda) - 5(-5\lambda) + 5 \cdot 5\lambda = 0;$$

$$-\lambda^3 + 28\lambda^2 - 2\lambda = 0;$$

$$-\lambda(\lambda^2 - 28\lambda + 2) = 0 = 0;$$

$$\rightarrow \lambda^2 - 28\lambda + 2 = 0 : \quad \lambda = 14 + \sqrt{194}, \lambda = 14 - \sqrt{194}$$

Thus,

$$\lambda_1 = 14 + \sqrt{194}; \lambda_2 = 14 - \sqrt{194}; \lambda_3 = 0$$

Thus, singular values are equal $\sigma_1 = \sqrt{14 + \sqrt{194}}; \sigma_2 = \sqrt{14 - \sqrt{194}}; \sigma_3 = 0$.

(b.iv)

$$A = \begin{pmatrix} 2 & 0 \\ 1 & 2 \end{pmatrix} \rightarrow A^\top A = \begin{pmatrix} 5 & 2 \\ 2 & 4 \end{pmatrix}$$

$$\text{Eigenvalues}(A^\top A) : \det \left(\begin{pmatrix} 5 & 2 \\ 2 & 4 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right) = 0$$

$$\begin{pmatrix} 5-\lambda & 2 \\ 2 & 4-\lambda \end{pmatrix} = 0$$

$$(5-\lambda)(4-\lambda) - 2 \cdot 2 = 0$$

$$\lambda^2 - 9\lambda + 16 = 0 \rightarrow \lambda_1 = \frac{9+\sqrt{17}}{2}, \lambda_2 = \frac{9-\sqrt{17}}{2}.$$

$$\rightarrow \sigma_1 = \sqrt{\frac{9+\sqrt{17}}{2}}; \sigma_2 = \sqrt{\frac{9-\sqrt{17}}{2}}.$$

Problem 4

$$A = \begin{pmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{pmatrix}$$

(a) and (b)

Given that $A^\top A$ is 3 by 3 matrix, we will have 3 singular values. Number of non-zero singular values of matrix A equals to the rank of the matrix A , which can be determined by number of linearly independent columns. In any 2 by 3 matrix such number at maximum equals 2. In matrix A we have exactly 2 linearly independent vectors. It means that $\text{rank}(A) = 2 \rightarrow A$ has 2 non-zero singular values.

Non-zero singular values of A are determined by non-zero eigenvalues of $A^\top A$:

$$A^\top A = \begin{pmatrix} 3 & 2 \\ 2 & 3 \\ 2 & -2 \end{pmatrix} \begin{pmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{pmatrix} = \begin{pmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{pmatrix} \rightarrow$$

$$\rightarrow \text{Eigenvalues}(A^\top A) :$$

$$\begin{aligned}
& \det \left(\begin{pmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \right) = 0; \\
& \det \begin{pmatrix} 13 - \lambda & 12 & 2 \\ 12 & 13 - \lambda & -2 \\ 2 & -2 & 8 - \lambda \end{pmatrix} = 0; \\
& (13 - \lambda)(\lambda^2 - 21\lambda + 100) - 12(-12\lambda + 100) + 2(2\lambda - 50) = 0; \\
& -\lambda^3 + 34\lambda^2 - 225\lambda = 0; \\
& -\lambda(\lambda - 9)(\lambda - 25) = 0; \rightarrow \lambda_1 = 25; \lambda_2 = 9; \lambda_3 = 0; \rightarrow \\
& \rightarrow \sigma_1 = \sqrt{25} = 5; \sigma_2 = \sqrt{9} = 3.
\end{aligned}$$

Right-singular vectors of A are orthonormal eigenvectors of $A^\top A$:

Eigenvector($\lambda_1 = 25$):

$$\begin{aligned}
& \begin{pmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{pmatrix} - 25 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} -12 & 12 & 2 \\ 12 & -12 & -2 \\ 2 & -2 & -17 \end{pmatrix} R_2 \leftarrow R_2 + 1 \cdot R_1 = \\
& \begin{pmatrix} -12 & 12 & 2 \\ 0 & 0 & 0 \\ 2 & -2 & -17 \end{pmatrix} R_3 \leftarrow R_3 + \frac{1}{6} \cdot R_1 = \begin{pmatrix} -12 & 12 & 2 \\ 0 & 0 & 0 \\ 0 & 0 & -\frac{50}{3} \end{pmatrix} R_3 \leftarrow -\frac{3}{50} \cdot \\
& R_3 = \begin{pmatrix} -12 & 12 & 2 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} R_1 \leftarrow R_1 - 2 \cdot R_3 = \begin{pmatrix} -12 & 12 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} R_2 \leftrightarrow R_3 = \\
& \begin{pmatrix} -12 & 12 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow -\frac{1}{12} \cdot R_1 = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \\
& \rightarrow (A - 25I) \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{cases} z = 0 \\ x = y \end{cases} \rightarrow v_1 = \\
& = \begin{pmatrix} y \\ y \\ 0 \end{pmatrix} \rightarrow v_1 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \rightarrow \frac{v_1}{\|v_1\|} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix}
\end{aligned}$$

Eigenvector($\lambda_1 = 9$):

$$\begin{aligned}
& \begin{pmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{pmatrix} - 9 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 4 & 12 & 2 \\ 12 & 4 & -2 \\ 2 & -2 & -1 \end{pmatrix} R_1 \leftrightarrow R_2 = \begin{pmatrix} 12 & 4 & -2 \\ 4 & 12 & 2 \\ 2 & -2 & -1 \end{pmatrix} R_2 \leftarrow \\
& R_2 - \frac{1}{3} \cdot R_1 = \begin{pmatrix} 12 & 4 & -2 \\ 0 & \frac{32}{3} & \frac{8}{3} \\ 2 & -2 & -1 \end{pmatrix} R_3 \leftarrow R_3 - \frac{1}{6} \cdot R_1 = \begin{pmatrix} 12 & 4 & -2 \\ 0 & \frac{32}{3} & \frac{8}{3} \\ 0 & -\frac{8}{3} & -\frac{2}{3} \end{pmatrix} R_3 \leftarrow
\end{aligned}$$

$$\begin{aligned}
R_3 + \frac{1}{4} \cdot R_2 &= \begin{pmatrix} 12 & 4 & -2 \\ 0 & \frac{32}{3} & \frac{8}{3} \\ 0 & 0 & 0 \end{pmatrix} R_2 \leftarrow \frac{3}{32} \cdot R_2 = \begin{pmatrix} 12 & 4 & -2 \\ 0 & 1 & \frac{1}{4} \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow R_1 - 4 \cdot R_2 = \\
&\begin{pmatrix} 12 & 0 & -3 \\ 0 & 1 & \frac{1}{4} \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow \frac{1}{12} \cdot R_1 = \begin{pmatrix} 1 & 0 & -\frac{1}{4} \\ 0 & 1 & \frac{1}{4} \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \\
\rightarrow (A - 9I) \begin{pmatrix} x \\ y \\ z \end{pmatrix} &= \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & -\frac{1}{4} \\ 0 & 1 & \frac{1}{4} \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \left\{ \begin{array}{l} x = \frac{1}{4}z \\ y = -\frac{1}{4}z \end{array} \right\} \rightarrow \\
\rightarrow v_2 &= \begin{pmatrix} \frac{1}{4} \\ -\frac{1}{4} \\ 1 \end{pmatrix} \rightarrow \frac{v_2}{\|v_2\|} = \begin{pmatrix} \frac{1}{3\sqrt{2}} \\ -\frac{1}{3\sqrt{2}} \\ \frac{2\sqrt{2}}{3} \end{pmatrix}
\end{aligned}$$

Eigenvector($\lambda_3 = 0$) :

$$\begin{aligned}
&\begin{pmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{pmatrix} - 0 \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{pmatrix} R_2 \leftarrow R_2 - \frac{12}{13} \cdot R_1 = \\
&\begin{pmatrix} 13 & 12 & 2 \\ 0 & \frac{25}{13} & -\frac{50}{13} \\ 2 & -2 & 8 \end{pmatrix} R_3 \leftarrow R_3 - \frac{2}{13} \cdot R_1 = \begin{pmatrix} 13 & 12 & 2 \\ 0 & \frac{25}{13} & -\frac{50}{13} \\ 0 & -\frac{50}{13} & \frac{100}{13} \end{pmatrix} R_2 \leftrightarrow R_3 = \\
&\begin{pmatrix} 13 & 12 & 2 \\ 0 & -\frac{50}{13} & \frac{100}{13} \\ 0 & \frac{25}{13} & -\frac{50}{13} \end{pmatrix} R_3 \leftarrow R_3 + \frac{1}{2} \cdot R_2 = \begin{pmatrix} 13 & 12 & 2 \\ 0 & -\frac{50}{13} & \frac{100}{13} \\ 0 & 0 & 0 \end{pmatrix} R_2 \leftarrow -\frac{13}{50} \cdot \\
&R_2 = \begin{pmatrix} 13 & 12 & 2 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow R_1 - 12 \cdot R_2 = \begin{pmatrix} 13 & 0 & 26 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{pmatrix} R_1 \leftarrow \frac{1}{13} \cdot \\
&R_1 = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \left\{ \begin{array}{l} x = -2z \\ y = 2z \end{array} \right\} \rightarrow v_3 = \\
&\begin{pmatrix} -2 \\ 2 \\ 1 \end{pmatrix} \rightarrow \frac{v_3}{\|v_3\|} = \begin{pmatrix} -\frac{2}{3} \\ \frac{2}{3} \\ \frac{1}{3} \end{pmatrix}
\end{aligned}$$

Thus, right singular vectors are:

$$v_1 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix}; v_2 = \begin{pmatrix} \frac{1}{3\sqrt{2}} \\ -\frac{1}{3\sqrt{2}} \\ \frac{2\sqrt{2}}{3} \end{pmatrix}; v_3 = \begin{pmatrix} -\frac{2}{3} \\ \frac{2}{3} \\ \frac{1}{3} \end{pmatrix}.$$

(c)

Left singular vectors are:

$$u_1 = \frac{Av_1}{\sigma_1} = \frac{Av_1}{\sqrt{\lambda_1}} = \begin{pmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix} \frac{1}{5} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$u_2 = \frac{Av_2}{\sqrt{\lambda_2}} = \begin{pmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{pmatrix} \begin{pmatrix} \frac{1}{3\sqrt{2}} \\ -\frac{1}{3\sqrt{2}} \\ \frac{2\sqrt{2}}{3} \end{pmatrix} \frac{1}{3} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

(d)

$$U = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$$V = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{3\sqrt{2}} & -\frac{2}{3} \\ \frac{1}{\sqrt{2}} & -\frac{1}{3\sqrt{2}} & \frac{2}{3} \\ 0 & \frac{2\sqrt{2}}{3} & \frac{1}{3} \end{pmatrix}$$

$$\Sigma = \begin{pmatrix} 5 & 0 & 0 \\ 0 & 3 & 0 \end{pmatrix}$$

$$A = U\Sigma V^\top = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 5 & 0 & 0 \\ 0 & 3 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{3\sqrt{2}} & -\frac{1}{3\sqrt{2}} & \frac{2\sqrt{2}}{3} \\ -\frac{2}{3} & \frac{2}{3} & \frac{1}{3} \end{pmatrix} = \begin{pmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{pmatrix} =$$

$$= A.$$

Problem 1

(a)

Let $A = [a_1, \dots, a_n]$, where a_i is a column vector of A .

For any A with rank r there exists a set of vectors, which comprises a basis of column space of A :

$$(\forall A, \exists \{u_1, \dots, u_r\} \mid \{u_1, \dots, u_r\} \text{ is a basis for } C(A)),$$

where r - rank of A .

It means that every column ^{vector} of A can be created as a linear combination of $\{u_1, \dots, u_r\}$:

$$a_i = v_1 u_1 + \dots + v_r u_r = \sum_{j=1}^r v_j^{(i)} u_j, \text{ where}$$

- $v_j^{(i)}$ is a scalar

for certain basis vector

- v_j is a vector of scalars ^{u_i} for a col. vector a_i
- i is an index of the column vector of A .

Thus,

$$\begin{aligned} A &= [a_1, \dots, a_n] = [(v_1^{(1)} u_1 + \dots + v_r^{(1)} u_r) \dots (v_1^{(n)} u_1 + \dots + v_r^{(n)} u_r)] \\ &= \left[\sum_{j=1}^r v_j^{(1)} u_j \dots \sum_{j=1}^r v_j^{(n)} u_j \right] = \sum_{j=1}^r u_j v_j^T \end{aligned}$$

Given that there is an infinite ways to construct matrix A from different sets of basis vectors, we can conclude that representation of A as $\sum u_i v_i^T$ of rank 4 is not unique.

(b)

$$A = \begin{pmatrix} 1 & -2 & 3 & -4 \\ -2 & 4 & -6 & 8 \\ 3 & -6 & 9 & -12 \\ -4 & 8 & -12 & 16 \end{pmatrix} \rightarrow \text{we see that all columns of } A \text{ are just multiples of first column.}$$

It means, that A has rank of 1 with $u = \begin{pmatrix} 1 \\ -2 \\ 3 \\ -4 \end{pmatrix}$ being a basis.

Thus, we need only to find a vector v , such that $A = u v^T$.

We see that vector v of coefficients is $\begin{pmatrix} 1 \\ -2 \\ 3 \\ -4 \end{pmatrix}$ because first column is just a multiple of itself $a_1 = 1u$, while other cols are $a_2 = -2u$, $a_3 = 3u$, $a_4 = -4u$.

$$\text{So, } u v^T = \begin{pmatrix} 1 \\ -2 \\ 3 \\ -4 \end{pmatrix} \begin{pmatrix} 1 & -2 & 3 & -4 \end{pmatrix} = \begin{pmatrix} 1 & -2 & 3 & -4 \\ -2 & 4 & -6 & 8 \\ 3 & -6 & 9 & -12 \\ -4 & 8 & -12 & 16 \end{pmatrix} = A.$$

$$B = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \\ 4 & 5 & 6 & 7 \end{pmatrix}$$

It seems like we can construct matrix B from 2 basis vectors: $(1 \ 1 \ 1 \ 1)^T$ and first column vector of B , that is, use their linear combination.

Thus, let $\text{Basis}_{\text{of Column space of } B} = \{u_1, u_2\} = \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix} \right\}.$

Then: $0u_1 + 1u_2 = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix};$

$1u_1 + 1u_2 = \begin{pmatrix} 2 \\ 3 \\ 4 \\ 5 \end{pmatrix};$

$2u_1 + 1u_2 = \begin{pmatrix} 3 \\ 4 \\ 5 \\ 6 \end{pmatrix};$

$3u_1 + 1u_2 = \begin{pmatrix} 4 \\ 5 \\ 6 \\ 7 \end{pmatrix}.$

Thus, for v_1 we will use $\begin{pmatrix} 0 \\ 1 \\ 2 \\ 3 \end{pmatrix}$ and for $v_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}.$

Therefore $B = u_1 v_1^T + u_2 v_2^T$, where

$$u_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, u_2 = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix}, v_1 = \begin{pmatrix} 0 \\ 1 \\ 2 \\ 3 \end{pmatrix}, v_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}.$$

Problem 9

(a)

$$\text{Let } A = U \Sigma V^T$$

$$\begin{aligned} A^+ &= (A^T A)^{-1} A^T = ((V \Sigma^T U^T)(U \Sigma V^T))^{-1} V \Sigma U^T = \\ &= (V \Sigma^2 V^T)^{-1} V \Sigma U^T = (V^T)^{-1} \Sigma^{-2} V^{-1} V \Sigma U^T = V \Sigma^{-2} \Sigma U^T = \\ &= V \Sigma^{-1} U^T = V \Sigma^+ U^T \rightarrow (V \Sigma^+ U^T) A = V \Sigma^+ U^T U \Sigma V^T = \\ &= V \Sigma^+ \Sigma V^T = V V^T = I \rightarrow A^+ \text{ is a left inverse of } A. \end{aligned}$$

$$A A^+ = (U \Sigma V^T)(V \Sigma^+ U^T) = U \Sigma \Sigma^+ U^T$$

(b)(1).

$$\begin{aligned} A A^+ A &= (U \Sigma V^T)(V \Sigma^+ U^T)(U \Sigma V^T) = \\ &= U \Sigma V^T V \Sigma^+ \Sigma V^T = U \Sigma V^T = A. \end{aligned}$$

(b)(2).

$$\begin{aligned} A^+ A A^+ &= (V \Sigma^+ U^T)(U \Sigma V^T)(V \Sigma^+ U^T) = \\ &= V \Sigma^+ \Sigma \Sigma^+ U^T = V \Sigma^+ U^T = A^+ \end{aligned}$$

(b)(3)

$$(AA^+)^T = ((U\Sigma V^T)(V\Sigma^+U^T))^T =$$

$$= (U\Sigma\Sigma^+U^T)^T = U(\Sigma\Sigma^+)^T U^T = U\Sigma\Sigma^+U^T =$$

$$= AA^+ \rightarrow AA^+ \text{ is symmetric.}$$

$$(A^+A)^T = ((V\Sigma^+U^T)(U\Sigma V^T))^T = (V\Sigma^+\Sigma V^T)^T =$$

$$= V(\Sigma^+\Sigma)^T V^T = V\Sigma^+\Sigma V^T = A^+A \rightarrow$$

$$\rightarrow A^+A \text{ is symmetric.}$$

(c)

$$\text{Let } A = QR \text{ and } A^+ = (A^+A)^{-1}A^T \rightarrow$$

\rightarrow Then

$$A^+ = ((QR)^T(QR))^{-1}(QR)^T = (R^T Q^T Q R)^{-1} R^T Q^T =$$

$$\stackrel{\text{by } (Q^T Q) = I}{=} (R^T R)^{-1} R^T Q^T = R^{-1} (R^T)^{-1} R^T Q^T = R^{-1} I Q^T =$$

$$= R^{-1} Q^T.$$