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(1)	find $\text{rank}(A)$, a basis for the column space of A , $C(A)$, a basis for row space $A \subset C(A^T)$, a basis for the null space of A , $N(A)$ and a basis for the left null space of A , $N(A^T)$.	1
(2)	Construct $B = A^T A$, and find eigenvalues and eigenvectors of B . For positive eigenvalues of B , define $\sigma_i = \sqrt{\lambda_i}$, where $\lambda_1 \geq \lambda_2$. Construct matrix D as follows	6
(3)	If the eigenvectors of are not orthogonal, orthogonalise them. Make a matrix V using the orthonormal vectors you obtained. The ordering of the columns of V should be the same as the ordering of the eigen values, that is $V = [v_1, v_2]$. In other words, v_1 is the eigenvector corresponding λ_1 and λ_2 is the eigenvector corresponding to λ_2 .	9
(4)	find eigenvalues and eigenvectors of the matrix $G = AA^T$, and orthonormalise them. Call them v_1, v_2 , and v_3 according to the order of $\lambda_1 \geq \lambda_2 \geq \lambda_3$ of eigenvalues of G . Construct the matrix $U = [v_1, v_2, v_3]$, and check whether it is an orthogonal matrix.	11
(5)	find three vectors $\{w_1, w_2, w_3\}$ so that	15
	$w_i = \frac{1}{\sigma_i} A v_i, i = 1, 2$	
	$w_3 \perp w_1, w_3 \perp w_2, \text{ and } \ w_3\ = 1$	
	Compare $\{w_1, w_2, w_3\}$ with $\{v_1, v_2, v_3\}$ and comment similar characteristics of these sets.	

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(6)	Compute the product VDV^T	20
(7)	Do the questions (1), (2), (3), (4) for the slightly different version of A $A_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix}$ construct σ_i, u_i, v_i, D_p and U_p . Justify the changes you observe. (In this question you may use decimal numbers.)	23
(8)	Construct a low-rank approximation of A and A_p following this definition using their corresponding eigenvectors. $A \approx \sigma_1 v_1 u_1^T$ $A_p \approx \sigma_{p1} v_{p1} u_{p1}^T$ here σ_{p1}, v_{p1} and u_{p1} are corresponding to the eigenvalues and eigenvectors related to A_p compute these two matrix (justify change observed).	37
(9)	(for D and HD levels) Considered slightly altered version of A	41
(9.1)	Find eigenvalue & $A_\epsilon \equiv \begin{bmatrix} 1 & \epsilon \\ 0 & 1 \\ \epsilon & 0 \end{bmatrix}$ eigenvectors of A_ϵ, A_ϵ^T , and $A_\epsilon^T A_\epsilon$ in term of ϵ . Compute $\sigma_i = \sqrt{\lambda_i}$ in term of ϵ for positive eigen values.	
(9.2)	Construct V_ϵ, D_ϵ , and U_ϵ	
(9.3)	Critically analyse the impact of ϵ on σ_1, σ_2 , eigenvectors, V_ϵ, D_ϵ , and U_ϵ .	

Question 1 Consider the following matrix. $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$
 find $\text{rank}(A)$, a basis for the column space of A , $C(A)$, a basis for the row space of A , $C(A^T)$, a basis for the null space of A , $N(A)$, and a basis for the left null space of A , $N(A^T)$

Hint: for a $m \times n$ matrix A , the column space of A is defined as $C(A) = \{Ax \mid x \in \mathbb{R}^n\}$, and the row space of A is defined as $C(A^T) = \{A^T y \mid y \in \mathbb{R}^m\} \subseteq \mathbb{R}^n$. Also, the nullspace $N(A)$ is defined as $\{x \in \mathbb{R}^n \mid Ax = 0\} \subseteq \mathbb{R}^n$, for the same matrix, the left nullspace $N(A^T)$ is defined as $\{y \in \mathbb{R}^m \mid A^T y = 0\} \subseteq \mathbb{R}^m$. Also we have

$$\dim(C(A)) + \dim(N(A^T)) = m$$

$$\dim(C(A^T)) + \dim(N(A)) = n$$

[Note. You need to use gaussian elimination to answer this question. For all the linear systems in this assignment, you need to use Gaussian elimination to solve them.]

Solution: Given:-

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

This is already in row echelon form (upper triangular form).

Gaussian Elimination:-

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

- Row 1 already has a leading 1
- In Row 2, the leading entry is also 1 and its in a new column
- Row 3 is all zeros $\rightarrow$ nothing to eliminate

1. find Rank(A):

- The rank of A is the number of linearly independent columns (for rows) in A.
- The columns of A are $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$,

which are linearly independent.

- The third row is all zeros, so it does not contribute to the rank.
- The rank is the number of leading 1s (pivots) in the row echelon form of A. matrix A is already in row echelon form.

Answer 1 $\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow \text{Rank}(A) = 2$

2. Basis for the Column Space C(A):

- Column space is spanned by the linearly independent columns of A.

The columns of A are

$$c_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad c_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

These are linearly independent.

So

Answer: 2

$$\text{Basis for } C(A) = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$$

3. Basis for the Row Space $C(A^T)$:

Row space of A is the span of the non-zero rows.

non-zero rows:

$$[1, 0], [0, 1]$$

So

Answer 3:

$$\text{Basis for } C(A^T) = \{ [1, 0], [0, 1] \}$$

4. Basis for the null space $N(A)$:

The null space consists of all vectors x such that $Ax = 0$

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

This gives $x_1 = 0$
 $x_2 = 0$

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Thus the only solution is the zero vector, meaning the null space is trivial.

So,

Answer: 4 : Basis of the null space $N(A)$
The only solution is the zero vector,
 $N(A) = \{0\}$; Basis = $\emptyset$

5 Basis for the left null space $N(A^T)$:
The left null space consists of all vectors y such that $A^T y = 0$

$$A^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

This gives,

$$y_1 = 0$$

$$y_2 = 0$$

y_3 is free.

Thus the null space is spanned by $\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$,

Answer 5

Basis for left null space $N(A^T)$:

$$\left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Answer & Solution for Question 1 is

- $\text{rank}(A) = 2$

- Basis for $C(A)$: $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$

- Basis for $C(A^T)$: $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$

- Basis for $C(A^T)$: $\{[1 \ 0], [0 \ 1]\}$

- Basis for $N(A)$: (Only zero vector)

- Basis for $N(A^T)$: $\left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$

The dimensions satisfy

- $\dim(C(A)) + \dim(N(A^T)) = 2 + 1 = 3 = m$

- $\dim(C(A^T)) + \dim(N(A)) = 2 + 0 = 2 = n$

Question no: 2: Construct $B = A^T A$, and find
 (2) eigenvalues and eigenvectors of B . For positive
 eigenvalues of B , define $\sigma_i = \sqrt{\lambda_i}$, where
 $\lambda_1 \geq \lambda_2$. Construct matrix D as follow.

$$D = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \\ 0 & 0 \end{bmatrix}$$

Solution for Question no: 2

given the matrix: $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$

Construct $B = A^T A$
 first compute A^T

$$A^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

now, multiplying A^T by A :

$$B = A^T A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}_{2 \times 3} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}_{3 \times 2}$$

$$= \begin{bmatrix} (1)(1) + (0)(0) + (0)(0) & (1)(0) + (0)(1) + (0)(0) \\ (0)(1) + (1)(0) + (0)(0) & (0)(0) + (1)(1) + (0)(0) \end{bmatrix}$$

$$B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

- Step 2: Find Eigenvalues and Eigenvectors of B

The matrix B is already diagonal, so its eigenvalues are the diagonal entries.

$$\lambda_1 = 1 \quad \lambda_2 = 1$$

The eigenvectors corresponding to $\lambda = 1$ are any non-zero vectors in $\mathbb{R}^2$, since B is the identity matrix. A standard choice for the basis of the eigenspace is.

$$v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

- Step 3 Define $\sigma_i = \sqrt{\lambda_i}$

Since both eigenvalues are positive ($\lambda_1 = 1, \lambda_2 = 1$)

$$\sigma_1 = \sqrt{\lambda_1} = \sqrt{1} = 1$$

$$\sigma_2 = \sqrt{\lambda_2} = \sqrt{1} = 1$$

- Step 4 Construct Matrix D .

The matrix D is constructed as:

$$D = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

So answer for Question 2

- (2) Construct $B = A^T A$, and find eigenvalues and eigenvectors of B . for positive eigenvalues of B , define $\sigma_i = \sqrt{\lambda_i}$, where $\lambda_1 \geq \lambda_2$.
Construct matrix D as follow.

$$D = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \\ 0 & 0 \end{bmatrix}$$

Answer is:

- $B = A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$
- Eigenvalues of $B = \lambda_1 = 1, \lambda_2 = 1$
- Eigenvectors of $B : v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$
- Singular values: $\sigma_1 = 1, \sigma_2 = 1$
- matrix D :

$$D = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Question 3:-

- (3) If the eigenvectors of B are not orthonormal, orthonormalise them. Make a matrix V using the orthonormal vectors you obtained. The ordering of the columns of V should be the same as the ordering of the eigenvalues, that is $V = [V_1, V_2]$. In other words, V_1 is the eigenvector corresponding λ_1 and V_2 is the eigen vector corresponding to λ_2 .

Solution for Question 3.

given the matrix $B = A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, we have already found its eigenvalues and eigenvectors in part (2).

Step 1: Verify Eigenvectors.

The eigenvectors corresponding to the eigenvalues $\lambda_1 = 1$ and $\lambda_2 = 1$ are.

$$V_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, V_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Step 2: Check orthonormality.

We need to verify if these eigenvectors are orthonormal.

1. Orthogonality

$$V_1 \cdot V_2 = (1)(0) + (0)(1) = 0$$

the vectors are orthogonal.

2. Normalization

$$\|v_1\| = \sqrt{1^2 + 0^2} = 1$$

$$\|v_2\| = \sqrt{0^2 + 1^2} = 1$$

both vectors are already normalized.
Since the eigenvectors are already orthonormal, no further orthonormalization is needed.

Step: 3. Construct Matrix V .

The matrix V is formed by arranging the orthonormal eigenvectors as columns ordered according to the eigenvalues (which are equal here, so the order is arbitrary).

$$V = [v_1, v_2] = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

So Answer for Question 3 is,

The matrix V is:

$$V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Question (4)

find eigenvalues and eigenvectors of matrix $G = AA^T$, and orthonormalise them. Call them v_1, v_2 , and v_3 according to the order of $\lambda_1 \geq \lambda_2 \geq \lambda_3$ of eigenvalues G . Construct the matrix $U = [v_1, v_2, v_3]$, and check whether it is an orthogonal matrix.

Solution :

given the matrix $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$, we first compute $G = A \cdot A^T$

$$G = A \cdot A^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Step 1: find Eigenvalues of G .

The matrix G is diagonal, so its eigenvalues are the diagonal entries.

$$\lambda_1 = 1, \lambda_2 = 1, \lambda_3 = 0$$

ordering

$$\lambda_1 = 1 \geq \lambda_2 = 1 \geq \lambda_3 = 0.$$

Step 2 Find Eigenvectors of G .

we solve $(G - \lambda I)u = 0$ for each eigenvalue.

1. for $\lambda = 1$:

$$G - I = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

The system $(G - I)u = 0$ gives

$$-u_3 = 0 \Rightarrow u_3 = 0$$

u_1 and u_2 are free. Thus the eigenvectors are,

$$u = \begin{bmatrix} u_1 \\ u_2 \\ 0 \end{bmatrix}$$

A basis for the eigenspace is:

$$u_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, u_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

Step 2. for $\lambda = 0$:
Continue

$$K = 0I = K = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

The system $Ku = 0$ gives

$$u_1 = 0, u_2 = 0$$

u_3 is free, Thus the eigenvectors is:

$$u_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Step 3: Orthonormalize the Eigenvectors
The eigenvectors u_1, u_2, u_3 are already orthogonal (since A is symmetric) and normalised.

$$1. u_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \|u_1\| = 1$$

$$2. u_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \|u_2\| = 1$$

$$3. u_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \|u_3\| = 1$$

no further orthonormalization is needed.

Step 4: Construct Matrix U

Arrange the orthonormal eigenvectors as columns in the order of $\lambda_1 \geq \lambda_2 \geq \lambda_3$:

$$U = [u_1 \ u_2 \ u_3] = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Step 5: Check if U is orthogonal
A matrix U is orthogonal if $U^T U = I$. Compute.

$$U^T U = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I$$

Thus, U is an orthogonal matrix.

Final Answer: Question : 4

- Eigenvalues of A : $\lambda_1 = 1, \lambda_2 = 1, \lambda_3 = 0$
- orthonormal eigenvector of A .

$$u_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, u_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, u_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

• Matrix U :

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

check for orthogonality:

U is orthogonal since $U^T U = I$.

Question: 5

find three vectors $\{w_1, w_2, w_3\}$ so that

$$w_i = \frac{1}{\sigma_i} A v_i, \quad i = 1, 2$$

$w_3 \perp w_1, w_3 \perp w_2$, and $\|w_3\| = 1$.

Compare $\{w_1, w_2, w_3\}$ with $\{u_1, u_2, u_3\}$ and comment similar characteristics of three sets.

Solution 5:

Objectives:

find three vectors $\{w_1, w_2, w_3\}$ such that:

1. $w_i = \frac{1}{\sigma_i} A v_i$ for $i = 1, 2$,

2. w_3 is orthogonal to w_1 and w_2 and has unit norm ($\|w_3\| = 1$).

3. Compare $\{w_1, w_2, w_3\}$ with $\{u_1, u_2, u_3\}$ (from part 4) and comment on their relationship.

Step 1 Recall given quantities:-

• $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$ • $\sigma_1 = 1, \sigma_2 = 1$

(Square roots of eigenvalues of $A^T A$)

• $v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ Eigenvectors of $A^T A$

$$u_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, u_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, u_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Orthogonal eigenvectors of AA^T

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axis

Step 2: Compute w_1 and w_2
using the formula $w_i = \frac{1}{\sigma_i} A v_i$

(i) for $i=1$:

$$w_1 = \frac{1}{1} A v_1 = A \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1.1 + 0.0 \\ 0.1 + 1.0 \\ 0.1 + 0.0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

(ii) for $i=2$:

$$w_2 = \frac{1}{1} A v_2 = A \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1.0 + 0.1 \\ 0.0 + 1.1 \\ 0.0 + 0.1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

observation:

w_1 and w_2 are identical to u_1 and u_2 , respectively.

Step 3: Compute w_3 (Orthogonal to w_1 and w_2)

We need w_3 such that.

$$1. w_3 \perp w_1 \text{ (i.e., } w_3 \cdot w_1 = 0)$$

$$2. w_3 \perp w_2 \text{ (i.e., } w_3 \cdot w_2 = 0)$$

$$3. \|w_3\| = 1.$$

Approach

- w_1 and w_2 span the xy -plane in $\mathbb{R}^3$
- The orthogonal complement is the z -axis so we choose.

$$w_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

Verification

- $w_3 \cdot w_1 = 0 \cdot 1 + 0 \cdot 0 + 1 \cdot 0 = 0$
- $w_3 \cdot w_2 = 0 \cdot 0 + 0 \cdot 1 + 1 \cdot 0 = 0$
- $\|w_3\| = \sqrt{0^2 + 0^2 + 1^2} = 1$

Observation:

w_3 is identical to u_3 .

Step 4: Compare $\{w_1, w_2, w_3\}$ and $\{u_1, u_2, u_3\}$

The two sets are:-

$$\{w_1, w_2, w_3\} = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\},$$

$$\{u_1, u_2, u_3\} = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}.$$

• Key Observations:

1. Identical Sets:

both sets are the same, consisting of the standard orthonormal basis of $\mathbb{R}^3$

2. Geometric Interpretation:

- w_1 and w_2 are the images of the right singular vectors $(v_1 \text{ and } v_2)$ under A , scaled by $\frac{1}{\sigma_i}$,
- u_1 and u_2 are the left singular vectors (eigenvectors of AA^T)
- The equality $w_i = u_i$ reflects the fact that A maps the standard basis vectors directly to the standard basis in $\mathbb{R}^3$

3. Orthonormality:

both sets are orthonormal:

- $w_i \cdot w_j = \delta_{ij}$ (Kronecker delta)
- $u_i \cdot u_j = \delta_{ij}$

Step 5: Connection to singular value decomposition (SVD)
The SVD of A is given by:

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

where :

- $U = [u_1, u_2, u_3] = I_3$ (identity matrix)
- $\Sigma = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$,
- $V = [v_1, v_2] = I_2$

Implications:

- The equality $w_i = u_i$ arises because A is already in a "diagonal" form with respect to the standard basis.
- Vectors w_3 and u_3 span the null space of A^T , which is orthogonal to the column space of A .

Final Answer for Question 5

$$w_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, w_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, w_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix},$$

comparison : $\{w_1, w_2, w_3\}$ and $\{u_1, u_2, u_3\}$ are identical.

Both sets form the standard orthonormal basis for $\mathbb{R}^3$ and represents the left singular vectors of A .

Question 6:

Compute the product UDV^T

Solution:

Objective :-

Compute the product UDV^T using the matrices obtained from previous parts and verify that it reconstructs the original matrix A .

Step 1: Recall the matrices from previous parts. From part (3), (2), and (4) we have:

1. U : Matrix of left singular vectors (from AA^T)

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

2. D : Diagonal matrix of singular values

$$D = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

3. V : matrix of right singular vectors (from $A^T A$)

$$V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Thus, the transpose V^T is

$$V^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

- Step 2: Compute the product UD
first multiply U and D .

$$UD = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

- The product is computed as follows
• first column of UD :

$$\textcircled{u} \quad \begin{bmatrix} 1 \cdot 1 + 0 \cdot 0 + 0 \cdot 0 \\ 0 \cdot 1 + 1 \cdot 0 + 0 \cdot 0 \\ 0 \cdot 1 + 0 \cdot 0 + 1 \cdot 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

- Second column of UD :

$$\begin{bmatrix} 1 \cdot 0 + 0 \cdot 1 + 0 \cdot 0 \\ 0 \cdot 0 + 1 \cdot 1 + 0 \cdot 0 \\ 0 \cdot 0 + 0 \cdot 1 + 1 \cdot 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

Thus:

$$UD = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

- Step 3: Compute the final product UDV^T
now, multiply UD by V^T

$$UDV^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

- The product is computed as follows
- first column of VDV^T :

$$\begin{bmatrix} 1 \cdot 1 + 0 \cdot 0 \\ 0 \cdot 1 + 1 \cdot 0 \\ 0 \cdot 1 + 0 \cdot 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

- Second column of VDV^T :

$$\begin{bmatrix} 1 \cdot 0 + 0 \cdot 1 \\ 0 \cdot 0 + 1 \cdot 1 \\ 0 \cdot 0 + 0 \cdot 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

Thus

$$VDV^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Step: 4 **Verification:**
The original matrix A is.

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

we observe that: $VDV^T = A$,

Final Answer:

$$VDV^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} = A$$

Conclusion:

The product VDV^T successfully reconstructs the original matrix A , demonstrating the singular value decomposition (SVD) of A .

Question 7: Do the questions (1), (2), (3), (4) for the slightly different version of A:

$$A_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix}$$

Construct G_p , U_p , μ_p , V_p , D_p , and V_p . Justify the changes you observe (In this question you may use decimal numbers).

Solution: for matrix $A_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix}$

We will systematically compute the requested quantities using gaussian elimination.

(i) Rank of A_p .

The rank is the number of linearly independent columns (or rows).

• Columns

$$c_1 = \begin{bmatrix} 1 \\ 0 \\ 0.1 \end{bmatrix}, \quad c_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

These are linearly independent (neither is a scalar multiple of the other).

- Rows: The first two are linearly independent, and the third row is multiple of the first row

Conclusion $\text{rank}(A_p) = 2$.

2. Basis for the column Space $C(A_p)$

The column space is spanned by the linearly independent columns of A_p .

$$\text{Basis for } C(A_p) = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0.1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$$

3. Basis for the Row Space $C(A_p^T)$

The row space is spanned by the non-zero rows of A_p , after Gaussian elimination.

Gaussian Elimination:-

$$A_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix} \quad R_3 \leftarrow R_3 - 0.1R_1 \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

The non-zero rows first rows of original matrix

Basis:

$$\text{Basis for } C(A_p^T) = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$$

4. Basis for the Null Space $N(A_P)$.

The null space consists of all $x \in \mathbb{R}^2$ such that $A_P x = 0$.

Solve $A_P x = 0$.

from the reduced row echelon form (RREF) of A_P .

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow x_1 = 0, x_2 = 0.$$

Conclusion

The null space is trivial

Basis for $N(A_P) = \emptyset$ (no basis vectors)

5. Basis for the left Null Space $N(A_P^T)$.

The left Null space consists of all $y \in \mathbb{R}^3$ such that $A_P^T y = 0$.

Solve $A_P^T y = 0$

$$A_P^T = \begin{bmatrix} 1 & 0 & 0.1 \\ 0 & 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0.1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

This gives

$$1. y_1 + 0.1 y_3 = 0 \Rightarrow y_1 = -0.1 y_3,$$

$$2. y_2 = 0$$

The solution is $y = y_3 \begin{bmatrix} -0.1 \\ 0 \\ 1 \end{bmatrix}, y_3 \in \mathbb{R}.$

Basis:

$$\text{Basis for } N(A_P^T) = \left\{ \begin{bmatrix} -0.1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Verification of Dimensions

1. Column Space + left Null Space

$$1. \dim(C(A_P)) + \dim(N(A_P^T)) = 2 + 1 = 3 = m$$

2. Row Space + Null space

$$2. \dim(C(A_P^T)) + \dim(N(A_P)) = 2 + 0 = 2 = n.$$

Both conditions are satisfied.

Findings & Observation: 1

$$\text{rank}(A_P) = 2,$$

$$\text{Basic for } C(A_P) = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0.1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\} \quad \text{12 3p}$$

$$\text{Basic for } C(A_P^T) = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$$

$$\text{Basic for } N(A_P) = \emptyset,$$

$$\text{Basic for } N(A_P^T) = \left\{ \begin{bmatrix} -0.1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Solution for Matrix $A_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix}$

Step 1: Compute $B_p = A_p^T A_p$

First, find $A_p^T = \begin{bmatrix} 1 & 0 & 0.1 \\ 0 & 1 & 0 \end{bmatrix}$

Now multiply A_p^T by A_p :

$$B_p = A_p^T A_p = \begin{bmatrix} 1.1 + 0.0 + 0.1 \cdot 0.1 & 1.0 + 0.1 + 0.1 \cdot 0 \\ 0.1 + 1.0 + 0.0 \cdot 1 & 0.0 + 1.1 + 0.0 \end{bmatrix}$$

Observation 1 $B_p = \begin{bmatrix} 1.01 & 0 \\ 0 & 1 \end{bmatrix}$

Key observation

The perturbation 0.1 introduces a slight change to the diagonal entry B_{11} .

Step 2: Find Eigenvalues (λ) and eigenvectors (v) of B_p

Solve $\det(B_p - \lambda I) = 0$

$$\det \begin{bmatrix} 1.01 - \lambda & 0 \\ 0 & 1 - \lambda \end{bmatrix} = (1.01 - \lambda)(1 - \lambda) = 0$$

• Eigenvalues

$$\lambda_1 = 1.01, \lambda_2 = 1.$$

- Eigenvectors:

- for $\lambda_1 = 1.01$

$$(B_p - 1.01I)v_1 = 0 \Rightarrow \begin{bmatrix} 0 & 0 \\ 0 & -0.01 \end{bmatrix} v_1 = 0 \Rightarrow$$

$$v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

- for $\lambda_2 = 1$:

$$(B_p - I)v_2 = 0 \Rightarrow \begin{bmatrix} 0.01 & 0 \\ 0 & 0 \end{bmatrix} v_2 = 0 \Rightarrow v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

observation:

Eigenvectors remain the standard basis, but λ_1 increases slightly due to perturbation.

Step 3. Compute singular values (σ_i)

$$\sigma_1 = \sqrt{\lambda_1} = \sqrt{1.01} \approx 1.005 \quad \sigma_2 = \sqrt{\lambda_2} = 1.$$

Step 4. Construct Matrix D_p .

$$D_p = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1.005 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

1,
A_p
≈ 1.005

Verification: The structure matches A_p , with third row zeroed out.

get,
the

the

Observation:

$$B_p = A_p^T A_p = \begin{bmatrix} 1.01 & 0 \\ 0 & 1 \end{bmatrix}$$

Eigenvalues of B_p : $\lambda_1 = 1.01$, $\lambda_2 = 1$,

Eigenvectors of B_p : $v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$,

Singular values: $\sigma_1 \approx 1.005$, $\sigma_2 = 1$

$$D_p = \begin{bmatrix} 1.005 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Key observations.

1. Impact of perturbation

- The eigenvalue λ_1 increases from 1 to 1.01, reflecting the added 0.1 perturbation in A_p
- Singular value adjust accordingly, with $\sigma_1 \approx 1.005$

2. Stability of Eigenvectors:

The eigenvectors v_1 and v_2 remain unchanged, indicating the perturbation does not alter the principal directions of B_p .

3. Matrix D_p :

The non-zero singular value σ_1 capture the stretch introduced by the perturbation.

This decomposition is critical for understanding how small perturbations affect the singular value decomposition (SVD) of A_p .

now: perturbed matrix $A_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix}$

Step 1. Compute $B_p = A_p^T A_p$

$$B_p = \begin{bmatrix} 1.01 & 0 \\ 0 & 1 \end{bmatrix}$$

Step 2. Eigenvalues and Eigenvectors of B_p .

• Eigenvalues

$$\lambda = 1.01, \lambda_2 = 1$$

• Eigenvectors

$$v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Step 3. Orthonormalization.

The eigenvectors v_1 and v_2 are already orthonormal.

$$\bullet v_1 \cdot v_2 = 0$$

$$\bullet \|v_1\| = \|v_2\| = 1$$

Step: 4 Construct V_p :

$$V = [v_1, v_2] = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Observation: 4

Observation for A_p :

Despite the perturbation, the eigenvectors remain orthonormal, and V_p is identical to V . The perturbation only affects the eigenvalues (λ_1 increases to 1.01).

Comparison and Justification of changes

Property	Original A	perturbed A_p	Justification.
eigenvalues (λ)	$\lambda_1 = 1, \lambda_2 = 1$	$\lambda_1 = 1.01$ $\lambda_2 = 1$	Perturbation increases λ_1 but leaves λ_2 unchanged.
Eigenvectors (CV)	$v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ $v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$	same as A	Perturbation does not alter eigenvector direction.
Matrix V	$V = I$	$V_p = I$	Eigenvectors remain orthonormal in both cases.

Key: Insights:

1. Stability of Eigenvectors:

The eigenvectors v_1 and v_2 are inherently orthonormal for both A and A_p , so no additional orthonormalization is required.

2. Impact of Perturbation.

The perturbation 0.1 in A_p increases λ_1 but not affect the eigenvector direction. This shows robustness in the principal directions of B_p .

3. Matrix V

V is the identity matrix in both cases indicating that the row space of A and A_p is perfectly aligned with the standard basis.

Answer

$$V_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \text{ eigenvalues}$$

$$\lambda_1 = 1.01$$

$$\lambda_2 = 1$$

The perturbation in A_p , increases λ_1 but leaves V_p unchanged, demonstrating that small perturbations do not necessarily disrupt eigenvector orthonormality.

Now: matrix $A_p = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix}$

Step 1: Compute $G_p = A_p A_p^T$

$$G_p = A_p A_p^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0.1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0.1 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0.1 \\ 0 & 1 & 0 \\ 0.1 & 0 & 0.01 \end{bmatrix}$$

Step 2: find Eigenvalues (λ) and eigenvectors (u) of G_p

- Eigenvalues: .
Solve $\det(G_p - \lambda I) = 0$

$$\det \begin{bmatrix} 1-\lambda & 0 & 0.1 \\ 0 & 1-\lambda & 0 \\ 0.1 & 0 & 0.01-\lambda \end{bmatrix} = (1-\lambda)[(1-\lambda)(0.01-\lambda)-0.01] = 0$$

$$\lambda_1 \approx 1.01, \lambda_2 = 1, \lambda_3 \approx 0$$

- Eigenvectors:

- for $\lambda_1 \approx 1.01$ $u_1 \approx \begin{bmatrix} 0.995 \\ 0 \\ 0.1 \end{bmatrix}$

- for $\lambda_2 \approx 1$

$$u_2 \approx \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

for $\lambda_3 \approx 0$ $u_3 \approx$

$$\begin{bmatrix} -0.1 \\ 0 \\ 0.995 \end{bmatrix}$$

Step 3 Orthogonalize Eigenvectors.

Normalize u_1 and u_3

$$u_1 = \frac{1}{\sqrt{1+e^2}} \begin{bmatrix} 1 \\ 0 \\ e \end{bmatrix} \approx \begin{bmatrix} 0.995 \\ 0 \\ 0.1 \end{bmatrix}$$

$$u_3 = \frac{1}{\sqrt{1+e^2}} \begin{bmatrix} -e \\ 0 \\ 1 \end{bmatrix} \approx \begin{bmatrix} -0.1 \\ 0 \\ 0.995 \end{bmatrix}$$

Step 4 Construct V_p , and verify orthogonality

$$V_p = [u_1 \ u_2 \ u_3] \approx \begin{bmatrix} 0.995 & 0 & -0.1 \\ 0 & 1 & 0 \\ 0.1 & 0 & 0.995 \end{bmatrix}$$

• check orthogonality

 $V_p^T V_p \approx I$, so V_p is orthogonal.

Comparison and Justification of changes

Observation : 5

Property	Original A	perturbed A_p	Justification
Eigenvalue(s)	$\lambda_1 = 1,$ $\lambda_2 = 1,$ $\lambda_3 = 0$	$\lambda_1 \approx 1.01,$ $\lambda_2 = 1,$ $\lambda_3 \approx 0$	Perturbation increases λ_1 due to added ϵ
Eigenvector(s)	$u_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix},$ $u_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix},$ $u_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$	$u_1 \approx \begin{bmatrix} 0.995 \\ 0 \\ 0.1 \end{bmatrix},$ $u_2 \approx \begin{bmatrix} -0.1 \\ 0 \\ 0.995 \end{bmatrix}$	Perturbation tilts, u_1 and u_3 away from standard basis.
Matrix U	$U = I$ (Identity)	$U_p \approx \begin{bmatrix} 0.995 & 0 & -0.1 \\ 0 & 1 & 0 \\ 0.1 & 0 & 0.995 \end{bmatrix}$	U_p becomes a rotation matrix due to the perturbation

Key observations:

1. perturbation effect:

- The eigenvalue λ_1 increases slightly, reflecting the added energy in the perturbed direction.
- The eigenvectors u_1 and u_3 tilt to align with the perturbation, while u_2 remains unchanged.

2. Orthogonality preservation:

both U and U_p are orthogonal, confirming that the SVD structure is maintained despite the perturbation.

3. Geometric Interpretation:

The perturbation introduces a slight rotation in the column space of A_p , captured by U_p .

for A_p .

(i) Eigenvalues of A_p : $\lambda_1 \approx 1.01$, $\lambda_2 = 1$, $\lambda_3 \approx 0$

(ii) Eigenvectors $u_1 \approx \begin{bmatrix} 0.995 \\ 0 \\ 0.1 \end{bmatrix}$, $u_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$, $u_3 \approx \begin{bmatrix} -0.1 \\ 0 \\ 0.995 \end{bmatrix}$

(iii) $U_p \approx \begin{bmatrix} 0.995 & 0 & -0.1 \\ 0 & 1 & 0 \\ 0.1 & 0 & 0.995 \end{bmatrix}$, orthogonal: Yes.

Questions:-

- (8) Construct a low-rank approximation of A and A_p following this definition using their corresponding eigenvectors:

$$A \approx \sigma_1 v_1 u_1^T$$

$$A_p \approx \sigma_{p1} v_{p1} u_{p1}^T$$

here σ_{p1} , v_{p1} , and u_{p1} are corresponding to the eigenvalue and eigenvectors related to A_p . Compare these two matrices and justify the changes if you observe any.

Solution:-

Low-Rank approximation of A and A_p .

1. Low-Rank Approximation of original matrix A from previous parts we have.

$$\bullet \sigma_1 = 1, \sigma_2 = 1$$

$$\bullet v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, u_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

Rank-1 Approximation of A :

$$A \approx \sigma_1 v_1 u_1^T = 1 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Observation:

This approximation captures only the first column of A , losing the second column entirely

2. Low-Rank Approximation of perturbed matrix A_p .

from part (7), we have

- $\sigma_{p1} \approx 1.005$

- $V_{p1} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

- $U_{p1} \approx \begin{bmatrix} 0.995 \\ 0 \\ 0.1 \end{bmatrix}$

Rank-1 Approximation of A_p :

$$A_p \approx \sigma_{p1} V_{p1} U_{p1}^T$$

$$= 1.005 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 0.995 & 0 & 0.1 \end{bmatrix} = \begin{bmatrix} 1.005 \cdot 0.995 & 0 & 1.005 \cdot 0.1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\approx \begin{bmatrix} 1 & 0 & 0.1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Observation

The approximation now includes the perturbation term (0.1) in the third row, the modified structure of A_p

Comparison and Justification of changes
 (a) end of key insight,

next part,

Key insight:

The perturbation in A_p introduces a slight tilt in the dominant singular vector (u_{p1}), allowing the rank-1 approximation to partially capture the modified structure. The original A lacks this flexibility, leading to a coarse approximation.

Final Answer:

Rank-1 approximation of A : $\begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$.

Rank-1 approximation of $A_p \approx \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0.1 & 0 \end{bmatrix}$

Changes.

1. A_p 's approximation retains the perturbation (0.1) while A 's does not.
2. The singular value σ_{p1} increases slightly reflecting the perturbation's effect.

Comparison and Justification of Changes,

Property	Original A	Perturbed A_p	Justification,
① Singular Value (G)	1	≈ 1.005	Perturbation increases the "strength" of the dominant singular value.
② Approximation Quality	Loses Second. Column entirely.	Capture the perturbation in the third row.	The perturbation introduces a new direction in A_p preserved in the approximation.
③ Error.	$\ A - \text{Approx}\ _F = 1$	$\ A_p - \text{Approx}\ _F \approx 1$	Both approximations discard the second singular component but A_p 's perturbation is retained.
④ Visualization,	$\begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0.1 & 0 \end{bmatrix}$	Blue highlights the retained perturbation.

Question:

(9) (For D and 11D levels) consider the slightly altered version of A :

$$A_\epsilon = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ \epsilon & 0 \end{bmatrix}$$

- (9.1) find eigenvalues and eigenvectors of $A_\epsilon A_\epsilon^T$ and $A_\epsilon^T A_\epsilon$ in terms of ϵ . Compute $\sigma_i = \sqrt{\lambda_i}$ in terms of ϵ for positive eigenvalues
- (9.2) Construct V_ϵ , D_ϵ , and U_ϵ
- (9.3) Critically analyse the impact of ϵ on the σ_i , λ_i , eigenvectors, V_ϵ , D_ϵ and U_ϵ

Solution for part (9): Analysis of A_ϵ

given matrix:

$$A_\epsilon = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ \epsilon & 0 \end{bmatrix}$$

- (9.1) Eigenvalues and eigenvectors of $A_\epsilon A_\epsilon^T$ and $A_\epsilon^T A_\epsilon$

Step 1: Compute $A_\epsilon^T A_\epsilon$

$$A_\epsilon^T A_\epsilon = \begin{bmatrix} 1 & 0 & \epsilon \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ \epsilon & 0 \end{bmatrix} = \begin{bmatrix} 1+\epsilon^2 & 0 \\ 0 & 1 \end{bmatrix}$$

• Eigenvalues (λ_i):

$$\det(A_e^T A_e - \lambda I) = (1 + e^2 - \lambda)(1 - \lambda) = 0$$

$$\lambda_1 = 1 + e^2, \lambda_2 = 1$$

• Eigenvectors (v_i):

• For $\lambda_1 = 1 + e^2$:

$$\begin{bmatrix} 0 & 0 \\ 0 & -e^2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = 0 \Rightarrow v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

• For $\lambda_2 = 1$

$$\begin{bmatrix} e^2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = 0 \Rightarrow v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Singular Value (σ_i):

$$\sigma_1 = \sqrt{1 + e^2}, \sigma_2 = 1$$

Step 2: Compute $A_e A_e^T$

$$A_e A_e^T = \begin{bmatrix} 1 & 0 & e \\ 0 & 1 & 0 \end{bmatrix}^T \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ e & 0 \end{bmatrix} = \begin{bmatrix} 1 + e^2 & 0 & e \\ 0 & 1 & 0 \\ e & 0 & e^2 \end{bmatrix}$$

Eigenvalues (λ_i):

$$\det(Ae Ae^T - \lambda I) = -\lambda(1-\lambda)(e^2-\lambda) + e^2(1-\lambda) = 0$$

$$\lambda_1 = 1+e^2, \lambda_2 = 1, \lambda_3 = 0$$

Eigenvectors (μ_i):

• For $\lambda_1 = 1+e^2$

$$\mu_1 = \frac{1}{\sqrt{1+e^2}} \begin{bmatrix} 1 \\ 0 \\ e \end{bmatrix}$$

• For $\lambda_2 = 1$:

$$\mu_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

• For $\lambda_3 = 0$

$$\mu_3 = \frac{1}{\sqrt{1+e^2}} \begin{bmatrix} -e \\ 0 \\ 1 \end{bmatrix}$$

(9.2) Construct V_e , D_e and V_e

Matrix V_e :

$$V_e = [v_1 \ v_2] = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Matrix D_ϵ :

$$D_\epsilon = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} \sqrt{1+\epsilon^2} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Matrix U_ϵ :

$$U_\epsilon = [u_1 \ u_2 \ u_3] = \begin{bmatrix} \frac{1}{\sqrt{1+\epsilon^2}} & 0 & \frac{\epsilon}{\sqrt{1+\epsilon^2}} \\ 0 & 1 & 0 \\ \frac{\epsilon}{\sqrt{1+\epsilon^2}} & 0 & \frac{1}{\sqrt{1+\epsilon^2}} \end{bmatrix}$$

(9.3) Impact of ϵ on Components

Component	Impact of ϵ	Critical Analysis
① Singular values σ_i	$\sigma_1 = \sqrt{1+\epsilon^2}$ increase with ϵ ; $\sigma_2 = 1$ unchanged.	Measure the "stretch" along principle axes. perturbation amplifies the first singular value.
② Eigen values λ_i	$\lambda_1 = 1+\epsilon^2$ grows quadratically; others unchanged.	Reflects the energy in the perturbed direction.
③ Eigenvectors u_i	u_1 and u_3 tilt by $\arctan(\epsilon)$; u_2 unchanged.	Tilt captures the perturbation's direction. For $\epsilon \rightarrow 0$, $U_\epsilon \rightarrow I$.

Components	Impact of ϵ	Critical Analysis
• Matrix V_e	Unchanged (ϵ Independent)	Right singular vectors depend only on row space, unaffected by the perturbation.
• Matrix D_e	First diagonal entry scales with ϵ	Directly reflects the amplification of the first singular value.
• Matrix V_e	Becomes a rotation matrix for the first and third axes, as ϵ increases.	Encodes the perturbation's geometric effect on the column space.

Key observations -

- For $\epsilon \rightarrow 0$:
 - $A_e \rightarrow A$, $\sigma_1 \rightarrow 1$, $U_e \rightarrow U$.
- For $\epsilon \gg 0$:
 - $\sigma_1 \propto \epsilon$, U_e approaches a 90° rotation.
- Stability of V_e :
 - The right singular vectors remain stable, showing robustness in vector row space.

Final Answer:-

(Q.1) Eigen values and singular values

$$\lambda_1 = 1 + \epsilon^2, \lambda_2 = 1, \lambda_3 = 0; \sigma_1 = \sqrt{1 + \epsilon^2}, \sigma_2 = 1$$

(Q.2) Matrices,

$$V_\epsilon = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, D_\epsilon = \begin{bmatrix} \sqrt{1 + \epsilon^2} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$U_\epsilon = \begin{bmatrix} \frac{1}{\sqrt{1 + \epsilon^2}} & 0 & \frac{-\epsilon}{\sqrt{1 + \epsilon^2}} \\ 0 & 1 & 0 \\ \frac{\epsilon}{\sqrt{1 + \epsilon^2}} & 0 & \frac{1}{\sqrt{1 + \epsilon^2}} \end{bmatrix}$$

(Q.3) Impact of ϵ :

σ_1 and λ_1 grow with ϵ , U_ϵ tilts, V_ϵ remains stable.