

Homework 1 : Md Mahin, 1900421

1. (Question:10 points) Let B be a 4×4 matrix to which we apply the following operations

- (a) double column 1
- (b) halve row 3
- (c) add row 3 to row 1
- (d) interchange columns 1 and 4
- (e) subtract row 2 from each of the other rows
- (f) replace column 4 by column 3
- (g) delete column 1

- Write the result as a product of eight matrices.
- Write it again as a product ABC of three matrices.

Answer:

Let a 4×4 matrix $B = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix}$

and It's identity matrix $I = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$ as every operation on B

can be represented as multiplication with B , then

(a) **double column 1** $= \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix} \cdot \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

(b) **halve row 3** $= \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/2 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix}$

$$(c) \text{ add row 3 to row 1} = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix}$$

$$(d) \text{ interchange columns 1 and 4} = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix} \cdot \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

$$(e) \text{ subtract row 2 from each of the other rows} = \begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -1 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix}$$

$$(f) \text{ replace column 4 by column 3} = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$(g) \text{ delete column 1} = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix} \cdot \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- (Question) Write the result as a product of eight matrices.

Answer: To write the result as the product of eight matrices, row operations will be in the left of our matrix B and column operations will be in right.

$$\begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -1 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/2 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix} \cdot \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- **(Question)** Write it again as a product ABC of three matrices.

$$\begin{pmatrix} 1 & -1 & 1/2 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -1 & 1/2 & 0 \\ 0 & -1 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 3 & 2 & 9 & 6 \\ 1 & 5 & 8 & 7 \end{pmatrix} \cdot \begin{pmatrix} 0 & 2 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} -3 & -5 & 1/2 \\ 6 & 10 & 7 \\ -5 & -7 & 5/2 \\ -1 & -8 & 1 \end{pmatrix}$$

2. **(4 points) (Question)** What is the rank of the matrix $M = \begin{pmatrix} 1 & 2 & 4 & 4 \\ 3 & 4 & 8 & 0 \end{pmatrix}$?

Answer:

To calculate the rank, first we need to calculate the reduced row echelon form.

First Step: subtracting -3 times row 1 from row 2:

$$\begin{pmatrix} 1 & 2 & 4 & 4 \\ 0 & -2 & -4 & -12 \end{pmatrix}$$

Second Step: adding row 2 with row 1:

$$\begin{pmatrix} 1 & 0 & 0 & -8 \\ 0 & -2 & -4 & -12 \end{pmatrix}$$

Third Step: multiplying row 2 with -1/2:

$$\begin{pmatrix} 1 & 0 & 0 & -8 \\ 0 & 1 & 2 & 6 \end{pmatrix}$$

As there are two non zero diagonal of the matrix, the rank of the matrix is 2.

3. **(Question:)(6 points)** The Pythagorean theorem asserts that for a set of n orthogonal vectors $\{x_i\}$,

$$\left\| \sum_{i=1}^n x_i \right\|^2 = \sum_{i=1}^n \|x_i\|^2$$

- (a) **(Question:)** Prove this in the case $n = 2$ by an explicit computation of $\|x_1 + x_2\|^2$

Answer:

From vector definition we know that,

$\|x\| = \sqrt{x \cdot x}$ So, we can write,

$$\|x_1 + x_2\|^2 = (x_1 + x_2) \cdot (x_1 + x_2)$$

$$\|x_1 + x_2\|^2 = x_1 \cdot x_1 + x_1 \cdot x_2 + x_2 \cdot x_1 + x_2 \cdot x_2$$

$$\|x_1 + x_2\|^2 = \|x_1\|^2 + \|x_2\|^2 + 2(x_1 \cdot x_2)$$

$$\|x_1 + x_2\|^2 = \|x_1\|^2 + \|x_2\|^2 + 2x_1 \cdot x_2 \cos(\theta)$$

Now, as the vectors are orthogonal, $\theta = 90^\circ$ and so, $\cos(\theta) = 0$

So, we can write,

$$\|x_1 + x_2\|^2 = \|x_1\|^2 + \|x_2\|^2$$

(b) **(Question) Show that this computation also establishes the general case, by induction.**

Answer: Previously we have proved, for $n = 2$, $\{x_i\}$,

$$\left\| \sum_{i=1}^2 x_i \right\|^2 = \sum_{i=1}^2 \|x_i\|^2$$

So, for $n = 3$,

$$\left\| \sum_{i=1}^3 x_i \right\|^2 = \sum_{i=1}^3 \|x_i\|^2$$

or,

$$\|x_1 + x_2 + x_3\|^2 = \|x_1\|^2 + \|x_2\|^2 + \|x_3\|^2$$

we can write,

$$\|x_1 + x_2 + x_3\|^2 = (x_1 + x_2 + x_3) \cdot (x_1 + x_2 + x_3)$$

$$\|x_1 + x_2 + x_3\|^2 = x_1 \cdot x_1 + x_1 \cdot x_2 + x_1 \cdot x_3 + x_2 \cdot x_1 + x_2 \cdot x_2 + x_2 \cdot x_3 + x_3 \cdot x_1 + x_3 \cdot x_2 + x_3 \cdot x_3$$

$$\|x_1 + x_2 + x_3\|^2 = \|x_1\|^2 + \|x_2\|^2 + \|x_3\|^2 + 2(x_1 \cdot x_2) + 2(x_2 \cdot x_3) + 2(x_3 \cdot x_1)$$

$$\|x_1 + x_2 + x_3\|^2 = \|x_1\|^2 + \|x_2\|^2 + \|x_3\|^2$$

So, it is proven that if the equation is true for 2 and 3. Now, let's assume the equation is true for $n = n - 1$. It is,

$$\left\| \sum_{i=1}^{n-1} x_i \right\|^2 = \sum_{i=1}^{n-1} \|x_i\|^2$$

for x_1, \dots, x_{n-1} orthogonal vectors. Now, for another orthogonal

vector x_n to the vectors x_1, \dots, x_{n-1} , let assume, $x_1 = \left\| \sum_{i=1}^{n-1} x_i \right\|^2$

and $x_2 = x_n$, then,

$$\|x_1 + x_2\|^2 = \|x_1\|^2 + \|x_2\|^2$$

or,

$$\|x_1 + x_2\|^2 = \left\| \sum_{i=1}^{n-1} x_i \right\|^2 + \|x_n\|^2$$

So, we can write,

$$\left\| \sum_{i=1}^n x_i \right\|^2 = \sum_{i=1}^n \|x_i\|^2$$

[Proved]

4. **(Question)(6 points) Let $A \in \mathbb{R}^{m \times m}$ be symmetric. An eigenvector of A is a nonzero vector $x \in \mathbb{R}^m$ such that $Ax = \lambda x$ for some $\lambda \in \mathbb{R}$, the corresponding eigenvalue.**

- (a) **(Question:) Prove that all eigenvalues are real.**

Answer:

Here given that, $A \in \mathbb{R}^{m \times m}$ is a symmetric matrix and $Ax = \lambda x$ for some $\lambda \in \mathbb{R}$. For symmetric matrix we know, $A = A^T = A^*$

Given,

$$Ax = \lambda x$$

$$Ax^* = \lambda^* x^* \text{ [Complex conjugating both sides]}$$

$$x^* A^* = \lambda^* x^*$$

$$x^* A = \lambda^* x^* \text{ [As } A = A^*]$$

$$x^* Ax = \lambda^* x^* x \text{ [Multiplying x in both sides]}$$

$$x^* \lambda x = \lambda^* x^* x \text{ [As, } Ax = \lambda x]$$

$$x^* \lambda x - \lambda^* x^* x = 0$$

$$(\lambda - \lambda^*) x^* x = 0$$

Now, as, $x^* x$ can not become zero, as x is a non zero vector,

$$(\lambda - \lambda^*) = 0$$

So, we can write,

$$\lambda = \lambda^*$$

So, hence all eigenvalues are real. [Proved]

- (b) **(Question:) Prove that if x and y are eigenvectors corresponding to distinct eigenvalues, then x and y are orthogonal.**

Answer:

Let's, for two distinct eigenvectors x and y

$$Ax = \lambda_1 x \tag{1}$$

and

$$Ay = \lambda_2 y \tag{2}$$

where, λ_1 and λ_2 are distinct eigenvalues for two vectors.

Multiplying y^* on the left, first equation becomes,

$$y^*Ax = y^*\lambda_1x$$

$$y^*Ax = \lambda_1y^*x$$

Taking, complex conjugated of the equation,

$$y^*Ax^* = \lambda_1y^*x^*$$

$$x^*A^*y = \lambda_1x^*y$$

So,

$$x^*Ay = \lambda_1x^*y \quad (3)$$

Multiplying x^* on the left, second equation becomes,

$$x^*Ay = x^*\lambda_2y$$

So,

$$x^*Ay = \lambda_2x^*y \quad (4)$$

From equation three and four,

$$\lambda_1x^*y = \lambda_2x^*y$$

$$\lambda_1x^*y - \lambda_2x^*y = 0$$

$$(\lambda_1 - \lambda_2)x^*y = 0$$

Since, $(\lambda_1 \neq \lambda_2)$, so, $x^*y = 0$, and thus they are orthogonal.

[Proved]

5. (Question)6 points) If u and v are m -vectors, the matrix $A = I + uv^*$ is known as a *rank-one perturbation of the identity*. Show that if A is nonsingular, then its inverse has the form $A^{-1} = I + \alpha uv^*$ for some scalar α , and give an expression for α . For what u and v is A singular? If it is singular, what is $\text{null}(A)$?

Answer:

Given that A is non singular and $A = I + uv^*$. If, A is non singular then A^{-1} exists and the determinate of $A \neq 0$.

Lets consider, $A^{-1} = I + \alpha uv^*$

Now, we know, $AA^{-1} = I$ [Where, I is the identity matrix]

$$\text{So, } (I + uv^*)(I + \alpha uv^*) = I$$

$$\text{or, } I + uv^* + \alpha uv^* + uv^*\alpha uv^* = I$$

$$\text{or, } uv^* + \alpha uv^* + \alpha u(v^*u)v^* = 0$$

$$\text{or, } uv^* + \alpha uv^* + \alpha u(v^*u)v^* = 0$$

$$\text{or, } uv^*(1 + \alpha + \alpha v^*u) = 0$$

That implies that, $(1 + \alpha + \alpha v^*u) = 0$

or, $\alpha = -\frac{1}{1+v^*u}$ [Answer 1]

Here, we can see that, if A^{-1} exists, there exists a constant α . However, if $v^*u = -1$, α becomes undefined. So, at that point A becomes singular. So, A is singular if, $v^*u = -1$. [Answer 2]

Now, if A is singular, let for the vector u ,

$$Au = (I + uv^*)u$$

$$\text{or, } Au = Iu + uv^*u$$

$$\text{or, } Au = u + u(-1)$$

$$\text{or, } Au = u - u$$

$$\text{or, } Au = 0$$

Thus, u vector is the null space of A if it is singular [Answer 3].

6. (Answer 6:)(5 points) Let $\|\cdot\|_w$ denote any norm on \mathbb{R}^m and also the induced matrix norm on $\mathbb{R}^{m \times m}$. Show that $\rho(A) \leq \|A\|$, where $\rho(A)$ is the spectral radius of A , that is largest absolute value of $|\lambda|$ of an eigenvalue λ of A .

Answer:

We know, $Ax = \lambda x$

$$\text{or, } \|Ax\| = \|\lambda x\|$$

We know, from the properties of metrics norm,

$$\|\lambda A\| = |\lambda| \|A\|$$

and

$$\|Ax\| \leq \|A\| \|x\|$$

from here we can write,

$$\|\lambda x\| \leq \|A\| \|x\|$$

$$\text{or, } |\lambda| \|x\| \leq \|A\| \|x\|$$

as, x is eigenvector, $x \neq 0$,

$$\text{So, } |\lambda| \leq \|A\|$$

Given, $\rho(A) = \lambda$

$$\text{or, } \rho(A) \leq \|A\| \text{ [Proved]}$$

7. (Question)(15 points) Determine SVDs of the following matrices (by hand calculations)

$$(a) \text{SVD of : } \begin{pmatrix} 3 & 0 \\ 0 & -2 \end{pmatrix}$$

Answer:

$$\text{Let, } A = \begin{pmatrix} 3 & 0 \\ 0 & -2 \end{pmatrix}$$

For SVD, we know, $A = U\Sigma V^T$

$$\begin{aligned}
\text{Now, } A^T A &= V \Sigma^T \Sigma V \\
&= \begin{pmatrix} 3 & 0 \\ 0 & -2 \end{pmatrix} \cdot \begin{pmatrix} 3 & 0 \\ 0 & -2 \end{pmatrix} \\
&= \begin{pmatrix} 9 & 0 \\ 0 & 4 \end{pmatrix}
\end{aligned}$$

Now,

$$\det(A^T A - \lambda I) = \begin{vmatrix} 9 - \lambda & 0 \\ 0 & 4 - \lambda \end{vmatrix}$$

We can write,

$$(9 - \lambda)(4 - \lambda) = 0$$

or, $\lambda = 9$ or $\lambda = 4$

For, $\lambda = 4$

$$A^T A - 4I = \begin{pmatrix} 5 & 0 \\ 0 & 0 \end{pmatrix} = 0$$

So, $5x_1 + 0x_2 = 0$

$$\text{So, } v_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

For, $\lambda = 9$

$$A^T A - 9I = \begin{pmatrix} 0 & 0 \\ 0 & -5 \end{pmatrix} = 0$$

So, $0x_1 - 5x_2 = 0$

$$\text{So, } v_2 = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$

$$\text{So, our } V = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = V^*$$

Now,

$$\Sigma = \begin{pmatrix} \sqrt{9} & 0 \\ 0 & \sqrt{4} \end{pmatrix} = \begin{pmatrix} 3 & 0 \\ 0 & 2 \end{pmatrix}$$

Again, $AV = U\Sigma$

$$AV = \begin{pmatrix} 3 & 0 \\ 0 & -2 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 3 & 0 \\ 0 & 2 \end{pmatrix}$$

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

So, Finally,

$$A = U\Sigma V^T = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 3 & 0 \\ 0 & 2 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

(b) **SVD of** : $\begin{pmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$

Answer:

Let, $A = \begin{pmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$

For SVD, we know, $A = U\Sigma V^T$

Now, $A^T A = V\Sigma^T \Sigma V$

$$= \begin{pmatrix} 0 & 0 & 2 \\ 0 & 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 0 & 0 \\ 0 & 4 \end{pmatrix}$$

Now,

$$\det(A^T A - \lambda I) = \begin{vmatrix} -\lambda & 0 \\ 0 & 4 - \lambda \end{vmatrix}$$

We can write,

$$(-\lambda)(4 - \lambda) = 0$$

or, $\lambda = 0$ or $\lambda = 4$

For, $\lambda = 4$

$$A^T A - 4I = \begin{pmatrix} -4 & 0 \\ 0 & 0 \end{pmatrix} = 0$$

So, $-4x_1 + 0x_2 = 0$

$$\text{So, } v_1 = \begin{pmatrix} -1 \\ 0 \end{pmatrix}$$

For, $\lambda = 0$

$$A^T A - 0I = \begin{pmatrix} 0 & 0 \\ 0 & 4 \end{pmatrix} = 0$$

So, $0x_1 + 4x_2 = 0$

$$\text{So, } v_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$\text{So, } v = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\text{and, } v^T = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

Now,

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

Again, $AV = U\Sigma$

$$AV = \begin{pmatrix} 0 & -1 \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 2 & 0 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$AV = \begin{pmatrix} 2 & 0 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$$

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

So, Finally,

$$A = U\Sigma V^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 2 & 0 \\ 0 & 0 \\ 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

(c) **SVD of:** $\begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$

Answer:

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

For SVD, we know, $A = U\Sigma V^T$

Now, $A^T A = V\Sigma^T \Sigma V$

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

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

Now,

$$\det(A^T A - \lambda I) = \begin{vmatrix} 1-\lambda & 1 \\ 1 & 1-\lambda \end{vmatrix}$$

We can write,

$$(1-\lambda)(1-\lambda) - 1 = 0$$

$$1 - 2\lambda + \lambda^2 - 1 = 0$$

$$\lambda(\lambda - 2) = 0$$

$$\text{or, } \lambda = 0 \text{ or } \lambda = 2$$

$$\text{For, } \lambda = 0$$

$$A^T A - 0I = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} = 0$$

$$\text{Using row echelon form: } = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$$

$$\text{So, } x_1 + x_2 = 0$$

$$\text{or, } x_1 = -x_2$$

$$\text{So, } v_1 = \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\text{For, } \lambda = 2$$

$$A^T A - 2I = \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} = 0$$

$$\text{Using row echelon form: } = \begin{pmatrix} -1 & 1 \\ 0 & 0 \end{pmatrix}$$

$$\text{So, } -x_1 + x_2 = 0$$

$$\text{or, } x_1 = x_2$$

$$\text{So, } v_1 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

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

Now,

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

Again, $AV = U\Sigma$

$$AV = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$= \begin{pmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{pmatrix} \text{ So, } U = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

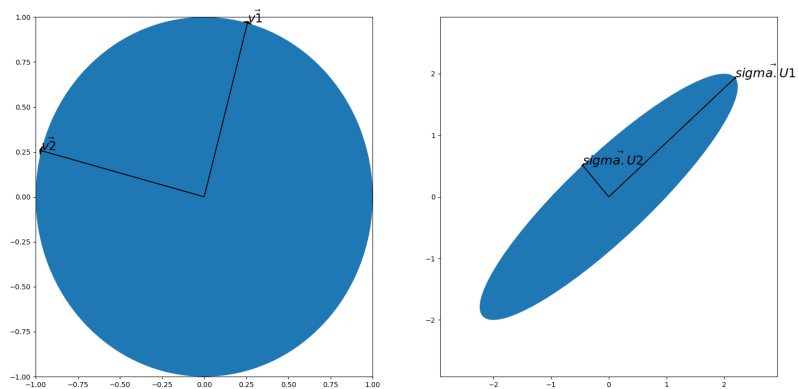
So, Finally,

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

8. (Question:)(12 points) Write a Matlab program which, given a real 2×2 matrix A , plots the right singular vectors v_1 and v_2 in the unit circle and also the left singular vectors u_1 and u_2 in the appropriate ellipse, as in Figure 4.1. Apply your program to the matrix $\begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix}$ and also to the 2×2 matrices of Exercise 7.

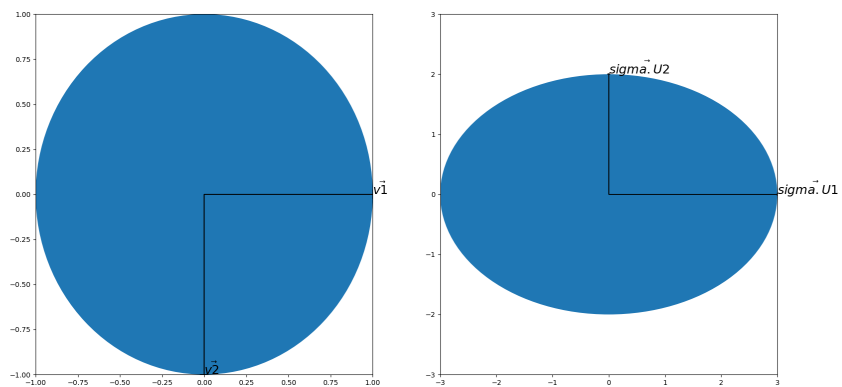
Answer:

For Matrix: $\begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix}$



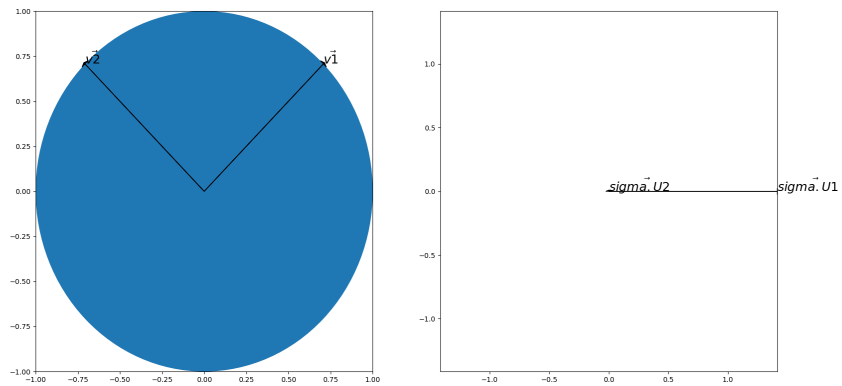
Answer:

For Matrix: $\begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix}$



Answer:

For Matrix: $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$



9. (Answer)(6 points) Consider the matrix

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

Using the SVD, work out (on paper) the exact values of $\sigma_{\min}(A)$ and $\sigma_{\max}(A)$ for this matrix.

Answer:

Given,

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

So,

$$A^T A = \begin{pmatrix} 1 & 0 \\ 2 & 2 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 \\ 0 & 2 \end{pmatrix}$$

So,

$$A^T A = \begin{pmatrix} 1 & 2 \\ 4 & 8 \end{pmatrix}$$

$$\text{Now, } \det(A^T A - \lambda I) = \begin{vmatrix} 1 - \lambda & 2 \\ 4 & 8 - \lambda \end{vmatrix}$$

$$\text{or, } (1 - \lambda)(8 - \lambda) - 4 = 0$$

$$\text{or, } 8 - 8\lambda - \lambda + \lambda^2 - 4 = 0$$

$$\text{or, } \lambda^2 - 9\lambda + 4 = 0$$

$$\text{or, } \lambda = \frac{9+\sqrt{65}}{2} \text{ or } \frac{9-\sqrt{65}}{2}$$

So, Largest value $\sigma_{\max}(A)$ is $\frac{9+\sqrt{65}}{2}$ and the smallest value $\sigma_{\min}(A)$ is $\frac{9-\sqrt{65}}{2}$.

10. **(Question:)(8 points)** Suppose $A \in \mathbb{R}^{m \times m}$ has an SVD $A = U\Sigma V^*$. find an eigenvalue decomposition of the $2m \times 2m$ symmetric matrix

$$\begin{pmatrix} 0 & A^* \\ A & 0 \end{pmatrix}$$

Please note: If the columns of a matrix $X \in \mathbb{R}^{m \times m}$ contain linearly independent eigen vectors of $A \in \mathbb{R}^{m \times m}$, the eigenvalue decomposition of A is

$$A = X\Sigma X^{-1}$$

where Σ is an $m \times m$ diagonal matrix whose entries are the eigenvalue of A.

Answer:

Given, $A = U\Sigma V^*$.

So, we can write, $AV = U\Sigma$

and $A^*U = V\Sigma$

By solving the linear equation, we can write,

$$\begin{pmatrix} 0 & A^* \\ A & 0 \end{pmatrix} \cdot \begin{pmatrix} U \\ V \end{pmatrix} = \begin{pmatrix} U\Sigma \\ V\Sigma \end{pmatrix}$$

Same way, it can be written that,

$$\begin{pmatrix} 0 & A^* \\ A & 0 \end{pmatrix} \cdot \begin{pmatrix} U \\ -V \end{pmatrix} = \begin{pmatrix} -U\Sigma \\ V\Sigma \end{pmatrix}$$

From above two equation we can write,

$$\begin{pmatrix} 0 & A^* \\ A & 0 \end{pmatrix} = \begin{pmatrix} U & -U \\ V & V \end{pmatrix}$$

$$\text{or, } \begin{pmatrix} U\Sigma & U\Sigma \\ V\Sigma & -V\Sigma \end{pmatrix} = \begin{pmatrix} U & -U \\ V & V \end{pmatrix} \cdot \begin{pmatrix} \Sigma & 0 \\ 0 & -\Sigma \end{pmatrix}$$

So, the EVD of A is,

$$\begin{pmatrix} 0 & A^* \\ A & 0 \end{pmatrix} = \begin{pmatrix} U & -U \\ V & V \end{pmatrix} \cdot \begin{pmatrix} \Sigma & 0 \\ 0 & -\Sigma \end{pmatrix} \begin{pmatrix} U & -U \\ V & V \end{pmatrix}^{-1}$$

Here, $X = \begin{pmatrix} U & -U \\ V & V \end{pmatrix}$ and $XX^T = 2I$

So, $\frac{1}{\sqrt{2}}X$ is unitary.

So, Eigen decomposition is,

$$\begin{pmatrix} 0 & A^* \\ A & 0 \end{pmatrix} = \frac{1}{\sqrt{2}}X \begin{pmatrix} \Sigma & 0 \\ 0 & -\Sigma \end{pmatrix} \frac{1}{\sqrt{2}}X^T$$

11. (Question:)(8 points) Consider the matrices

$$A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Answer the following questions by hand calculation.

(Question:)What is the orthogonal projector P onto range(A), and what is the image under P of the vector (1, 2, 3)*?

Answer:

Given,

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

The orthogonal projector P onto range(A) is,

$$P = A(A^T A)^{-1}A$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \left(\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \right)^{-1} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$$

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

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

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

$$\begin{aligned}
&= \begin{pmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \\ \frac{1}{2} & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \\
&= \begin{pmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 1 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \end{pmatrix} \\
\text{So, } \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} &= \begin{pmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 1 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 2 \\ 2 \\ 2 \end{pmatrix}
\end{aligned}$$

(Question:) What is the orthogonal projector **P** onto $\text{range}(\mathbf{B})$, and what is the image under **P** of the vector $(1, 2, 3)^*$?

Given,

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

The orthogonal projector **P** onto $\text{range}(\mathbf{A})$ is,

$$P = B(B^T B)^{-1} B$$

$$\begin{aligned}
&= \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \left(\begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \right)^{-1} \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix} \\
&= \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 2 & 2 \\ 2 & 5 \end{pmatrix}^{-1} \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix} \\
&= \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \frac{1}{6} \begin{pmatrix} 5 & -2 \\ -2 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix} \\
&= \begin{pmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \frac{5}{6} & -\frac{1}{3} \\ -\frac{1}{3} & \frac{1}{3} \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix} \\
&= \begin{pmatrix} \frac{1}{6} & \frac{1}{3} \\ -\frac{1}{3} & \frac{1}{3} \\ \frac{5}{6} & -\frac{1}{3} \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix} \\
&= \begin{pmatrix} \frac{5}{6} & \frac{1}{3} & \frac{1}{6} \\ \frac{1}{3} & \frac{1}{3} & -\frac{1}{3} \\ \frac{1}{6} & -\frac{1}{3} & \frac{5}{6} \end{pmatrix}
\end{aligned}$$

$$\text{So, } \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} \frac{5}{6} & \frac{1}{3} & \frac{1}{6} \\ \frac{1}{3} & \frac{1}{3} & -\frac{1}{3} \\ \frac{1}{6} & -\frac{1}{3} & \frac{5}{6} \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \\ 2 \end{pmatrix}$$

12. (Question)(6 points) Let A be a matrix with the property that columns 1,3,5,7,... are orthogonal to columns 2,4,6,8,... In a reduced QR factorization $A = \hat{Q}\hat{R}$, what special structure does \hat{R} possess? You may assume that A has full rank.

Answer:

For a given matrix with vectors a_1, a_2, \dots we can construct orthogonal vectors q_1, q_2, \dots using the Gram-Schmidt Orthogonalization. Here we know,

$$v_j = a_j - (q_1^* a_j)q_1 - (q_2^* a_j)q_2 - \dots - (q_{j-1}^* a_j)q_{j-1}$$

If we normalize, our new orthogonal vectors takes the following form,

$$\begin{aligned} q_1 &= \frac{a_1}{r_{11}} \\ q_2 &= \frac{a_2 - r_{12}q_1}{r_{22}} \\ q_3 &= \frac{a_3 - r_{13}q_1 - r_{23}q_2}{r_{33}} \\ q_4 &= \frac{a_4 - r_{14}q_1 - r_{24}q_2 - r_{34}q_3}{r_{44}} \\ &\dots \\ q_n &= \frac{a_n - \sum_{i=1}^{n-1} r_{in}q_i}{r_{nn}} \end{aligned}$$

Without normalization, we can write,

$$\begin{aligned} q_1 &= a_1 \\ \text{so, } q_2 &= a_2 - r_{12}q_1 = a_2 - \frac{\text{span}(a_2, q_1)}{\|q_1\|}q_1 = a_2 - \frac{a_2 \cdot a_1}{\|a_1\|}a_1 \\ \text{Here, as } a_2, a_1 &\text{ orthogonal, their dot product is zero. So, } r_{12} = 0 \text{ and,} \\ q_2 &= a_2 \\ \text{similarly, } q_3 &= a_3 - r_{13}q_1 - r_{23}q_2 = a_3 - \frac{\text{span}(a_3, q_1)}{\|q_1\|}q_1 - \frac{\text{span}(a_3, q_2)}{\|q_2\|}q_2 \\ &= a_3 - \frac{\text{span}(a_3, a_1)}{\|a_1\|}a_1 - \frac{\text{span}(a_3, a_2)}{\|a_2\|}a_2 \\ \text{Here, as } a_3, a_2 &\text{ orthogonal, their dot product is zero. So, } r_{23} = 0 \text{ and,} \\ q_3 &= a_3 - r_{13}q_1 \\ \text{or, } r_{13} &= \frac{a_3 \cdot a_1}{\|a_1\|} \end{aligned}$$

Similarly, for $q_4 = a_4 - r_{24}q_2$ and the terms $r_{14} = r_{34} = 0$ and $r_{24} = \frac{a_4 \cdot a_2}{\|a_2\|}$ and $r_{44} = \left\| a_4 - \frac{a_4 \cdot a_2}{\|a_2\|}a_2 \right\|$

Here we can easily see that, $q_1 \in (a_1), q_3 \in \text{span}(a_1, a_3), \dots, q_{2k+1} \in \text{span}(a_1, \dots, a_{2k+1})$, and similarly $q_2 \in (a_2), q_4 \in \text{span}(a_2, a_4), \dots, q_{2k} \in \text{span}(a_2, \dots, a_{2k})$. So, here we can write, on the other hand, if we consider the normalization,

$$\begin{aligned} \text{for } q_1, r_{11} &= \|a_1\|, q_2, r_{22} = \|a_2\|, q_3, r_{33} = \left\| a_3 - \frac{a_3 \cdot a_1}{\|a_1\|}a_1 \right\| \dots \text{ and} \\ r_{nn} &= \left\| a_n - \sum_{i=1}^{n-1} \frac{a_n \cdot a_i}{\|a_i\|}a_i \right\|, \text{ where } n \text{ and } i \text{ have to be odd or even at the} \end{aligned}$$

same time.

As a result our matrix R takes an upper triangular check-board metrics form like this,

$$R = \begin{pmatrix} \|a_1\| & 0 & \frac{a_3 \cdot a_1}{\|a_1\|} & 0 & \dots \\ 0 & \|a_2\| & 0 & \frac{a_4 \cdot a_2}{\|a_2\|} & \dots \\ 0 & 0 & \left\| a_3 - \frac{a_3 \cdot a_1}{\|a_1\|} \right\| & 0 & \dots \\ 0 & 0 & 0 & \left\| a_4 - \frac{a_4 \cdot a_2}{\|a_2\|} \right\| & \dots \\ \dots & \dots & \dots & \dots & \dots \end{pmatrix}$$

where,

if $i = j : r_{ij} = \|a_i\|$

if $i < j$ and both i or j are odd or even: $r_{ij} = \frac{a_j \cdot a_i}{\|a_i\|}$

if $i > j : r_{ij} = 0$ and

else, $r_{ij} = 0$

13. **(Answer:)(8 points)** Write a MATLAB function $[Q, R] = \text{mgs}(A)$ that computes a reduced QR factorization $A = \hat{Q}\hat{R}$ of an $m \times n$ matrix A with $m \geq n$ using modified Gram-Schmidt orthogonalization. The output variables are a matrix $Q \in \mathbb{R}^{m \times n}$ with orthonormal columns and a triangular matrix $R \in \mathbb{R}^{n \times n}$. Use your MGS QR factorization to solve the linear system $Ax = b$ where

$$A = \begin{pmatrix} 1 & 3 & 2 \\ 4 & 5 & 6 \\ 9 & 8 & 7 \end{pmatrix}, b = \begin{pmatrix} 13 \\ 32 \\ 46 \end{pmatrix}$$

```
Matrix A =
[[1 3 2]
 [4 5 6]
 [9 8 7]]

Vector B =
[[13]
 [32]
 [46]]

Q Matrix=
[[ 0.10101525  0.83541209 -0.54026156]
 [ 0.40406102  0.46178558  0.78961305]
 [ 0.90913729 -0.2980616  -0.29091007]]

R Matrix=
[[9.89949494 9.59644917 8.99035765]
 [0.          2.43067136 2.35510645]
 [0.          0.          1.62078469]]

Dot of QR=
[[1. 3. 2.]
 [4. 5. 6.]
 [9. 8. 7.]]

Solution of X=
[[1.]
 [2.]
 [3.]]

Process finished with exit code 0
```