

MTH 225

Homework #10

Due Date: April 16, 2025

1. Suppose that $A = U\Sigma V^*$ is a singular value decomposition of a matrix $A \in M_{n \times n}(\mathbb{C})$. Find the singular decomposition of A^* .
2. Suppose $P \in M_{n \times n}(\mathbb{C})$ is a unitary matrix and $A \in M_{n \times n}(\mathbb{C})$. Show that PA has the same singular values as A .
3. Consider the following matrices

$$A = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \quad C = \begin{bmatrix} 0 & 0 & 2 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad D = \begin{bmatrix} 3i & -3i \\ i & -i \end{bmatrix}.$$
 (a) By inspection, find the kernel of each of these matrices.
 (b) By inspection, find the image of each of these matrices.
 (c) By *not* using the Gram matrix, find the SVD for each of these matrices.
4. Prove that if $A \in M_{n \times n}(\mathbb{C})$ is a Hermitian matrix with eigenvalues $|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_n|$ then its singular values satisfy $\sigma_1 = |\lambda_1|, \sigma_2 = |\lambda_2|, \dots, \sigma_n = |\lambda_n|$.
5. If $A, B \in M_{n \times n}(\mathbb{C})$, prove or provide a counterexample that if A and B are similar then A and B have the same singular values.
6. Give examples of $A, B \in M_{2 \times 2}(\mathbb{C})$ that have:
 - (a) equal singular values but distinct eigenvalues,
 - (b) equal eigenvalues but distinct singular values.
7. If $A \in M_{n \times n}(\mathbb{C})$, prove or provide a counterexample that the singular values of A^2 are the squares of σ_i^2 of the singular values of A .
8. Use the Gram matrix to find the singular value decomposition of the matrix

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

Hint: There is not a quick way to do this problem by inspection. However, once you find \mathbf{v}_1 and σ_1 , you can directly compute \mathbf{u}_1 and then use orthogonality to obtain \mathbf{v}_2 and thus obtain \mathbf{u}_2 and σ_2 .

Homework #10

#1

Suppose that $A = U \Sigma V^*$ is a singular value decomposition of a matrix $A \in M_{n \times n}(\mathbb{C})$. Find the singular value decomposition of A^* .

Solution:

Since $A = U \Sigma V^*$ it follows that $A^* = V \Sigma U^*$ which is the SVD of A^* .

#2.

Suppose $P \in M_{n \times n}(\mathbb{C})$ is unitary and $A \in M_{n \times n}(\mathbb{C})$. Show that PA has the same singular values as A .

Solution:

Let $A = U \Sigma V^*$ be the SVD of A . If we let $\tilde{U} = PV$ it follows that \tilde{U} is unitary and

$$PA = \tilde{U} \Sigma V^*$$

is an SVD of PA . Consequently, PA and A have the same singular values.

#3

Consider the following matrices

$$A = \begin{bmatrix} 1 & 0 \\ -1 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, C = \begin{bmatrix} 0 & 0 & 2 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, D = \begin{bmatrix} 3i & -3i \\ i & -i \end{bmatrix}$$

(a) By inspection, find the kernel of these matrices.

(b) By inspection, find the image of these matrices.

(c) Find the SVD of these matrices.

Solution:

$$(a) \ker(A) = \text{span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$$

$$\ker(B) = \{0\}$$

$$\ker(C) = \text{span} \left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

$$\ker(D) = \text{span} \left\{ \sqrt{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}.$$

$$(b) \text{im}(A) = \text{span} \left\{ \sqrt{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

$$\text{im}(B) = \text{span} \left\{ \begin{bmatrix} 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right\}$$

$$\text{im}(C) = \text{span} \left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\}$$

$$\text{im}(D) = \text{span} \left\{ \sqrt{10} \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix} \right\}$$

(c) Case A:

$$\alpha_2 = 0, \vec{v}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Therefore,

$$\vec{v}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\Rightarrow A\vec{v}_1 = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \sqrt{2} \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}.$$

Therefore

$$\vec{v}_1 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}, \sigma_1 = \sqrt{2}.$$

From orthonormality

$$\vec{v}_2 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix}.$$

Consequently, $A = U \Sigma V^*$

$$\Rightarrow A = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Case B:

$$\sigma_1 = \sigma_2 = 1$$

$$\vec{v}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$B\vec{v}_1 = \begin{bmatrix} 0 \\ -1 \end{bmatrix} \Rightarrow \vec{u}_1 = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

$$B\vec{v}_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \Rightarrow \vec{u}_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

Therefore, $B = U \Sigma V^*$

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

Case C:

$$\sigma_1 = 2, \sigma_2 = 1, \sigma_3 = 0$$

$$\vec{v}_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix},$$

$$\vec{v}_3 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \vec{v}_4 = \begin{bmatrix} 0 \\ 1 \end{bmatrix},$$

$$\vec{v}_5 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \vec{v}_6 = \begin{bmatrix} 0 \\ 0 \end{bmatrix},$$

$$\Rightarrow C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix}$$

Case D:

$$\sigma_2 = 0, \vec{v}_2 = \begin{bmatrix} i/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

By orthogonality, we select

$$\vec{v}_1 = \begin{bmatrix} i/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

and thus

$$D\vec{v}_1 = \begin{bmatrix} 3i & -3i \\ i & -i \end{bmatrix} \begin{bmatrix} i/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 6i/\sqrt{2} \\ 2i/\sqrt{2} \end{bmatrix} = 2 \begin{bmatrix} 3i/\sqrt{2} \\ i/\sqrt{2} \end{bmatrix} = 2\sqrt{5} \begin{bmatrix} 3i/\sqrt{10} \\ i/\sqrt{10} \end{bmatrix}$$

Consequently, $\sigma_1 = 2\sqrt{5}$, $\vec{v}_1 = \begin{bmatrix} 3i/\sqrt{10} \\ i/\sqrt{10} \end{bmatrix}$ and by orthogonality

$$\vec{v}_2 = \begin{bmatrix} i/\sqrt{10} \\ -3i/\sqrt{10} \end{bmatrix}$$

Therefore, $D = U \Sigma V^*$

$$\Rightarrow D = \begin{bmatrix} 3\sqrt{10} & i\sqrt{10} \\ i\sqrt{10} & -3\sqrt{10} \end{bmatrix} \begin{bmatrix} 2\sqrt{5} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}$$

#4

Prove that if $A \in M_{n \times n}(\mathbb{C})$ is Hermitian with eigenvalues $|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_n|$ then its singular values satisfy $\sigma_1 = |\lambda_1|$, $\sigma_2 = |\lambda_2|, \dots, \sigma_n = |\lambda_n|$.

Solution:

Since A is Hermitian, by the spectral theorem for Hermitian matrices there exists unitary V such that

$$A = V \Delta V^*$$

where $\Delta = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$. Now,

$$\begin{aligned} A^* A &= V \Delta^* V^* V \Delta V^* \\ &= V \Delta^* \Delta V \\ &= V \begin{bmatrix} |\lambda_1|^2 & & \\ & \ddots & \\ & & |\lambda_n|^2 \end{bmatrix} V^* \end{aligned}$$

Thus the singular values, which are square roots of the eigenvalues of $A^* A$, satisfy

$$\sigma_1 = |\lambda_1|, \sigma_2 = |\lambda_2|, \dots, \sigma_n = |\lambda_n|.$$

#5

If $A, B \in M_{n \times n}(\mathbb{C})$, prove or provide a counterexample that if A and B are similar then A and B have the same singular values.

Solution:

Consider

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

The eigenvalues of A are given by $\lambda = 1, 2$ and the eigenvalues of B are given by $\lambda = 1, 2$. The eigenvectors of A are given by

$$\vec{v}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} \sqrt{2} \\ \sqrt{2} \end{bmatrix}.$$

Therefore

$$A = \begin{bmatrix} 1 & \sqrt{2} \\ 0 & \sqrt{2} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & \sqrt{2} \\ 0 & \sqrt{2} \end{bmatrix}^{-1}$$

which proves A and B are similar. Now,

$$A^*A = \begin{bmatrix} 1 & 1 \\ 1 & 4 \end{bmatrix}$$

Which has eigenvalues

$$\lambda = \frac{5 \pm \sqrt{25 - 4 \cdot 3}}{2}$$

$$= \frac{5 \pm \sqrt{13}}{2}.$$

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Consequently, the singular values of A are given by

$$\sigma_1 = \sqrt{\frac{5 + \sqrt{13}}{2}}, \quad \sigma_2 = \sqrt{\frac{5 - \sqrt{13}}{2}}.$$

However, the singular values of B are clearly $\sigma_1 = 2, \sigma_2 = 1$.

#6

Give examples of $A, B \in M_{2 \times 2}(\mathbb{C})$ that have

- (a) equal singular values but distinct eigenvalues
- (b) equal eigenvalues but distinct singular values.

Solution:

$$(a) A = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$(b) \text{ From } \#5, A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

#7

If $A \in M_{n \times n}(\mathbb{C})$, prove or provide a counterexample that the singular values of A^2 are the squares σ_i^2 of the singular values of A .

Solution:

Let $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$. The singular values of A are $\sigma_1 = 1, \sigma_2 = 0$. However, $A^2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ which has singular values $\sigma_1 = \sigma_2 = 0$.

#8

Use the Gram matrix to find the singular value decomposition of the matrix

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

Solution:

Now, $A^*A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$ which has eigenvalues

$$\sigma_1^2 = 3 \pm \sqrt{5}$$

For $\sigma_1^2 = \frac{1}{2}(3 + \sqrt{5})$ we have:

$$A^*A - \sigma_1^2 I = \begin{bmatrix} -\frac{1}{2} - \frac{\sqrt{5}}{2} & 1 \\ 1 & \frac{1}{2} - \frac{\sqrt{5}}{2} \end{bmatrix}.$$

An eigenvector is

$$\vec{x} = \begin{bmatrix} 1 \\ \frac{1}{2} + \frac{\sqrt{5}}{2} \end{bmatrix}.$$

$$\begin{aligned}\|\vec{x}\|^2 &= 1 + \frac{1}{4} + \sqrt{5} + \frac{5}{4} \\ &= \frac{5}{2} + \sqrt{5} \\ &= 5 + 2\sqrt{5}\end{aligned}$$

$$\Rightarrow \|\vec{x}\| = \sqrt{\frac{5 + 2\sqrt{5}}{2}}$$

Consequently,

$$\vec{v}_1 = \frac{\sqrt{2}}{\sqrt{5 + 2\sqrt{5}}} \begin{bmatrix} 1 \\ \frac{1}{2} + \frac{\sqrt{5}}{2} \end{bmatrix}$$

$$\begin{aligned}A\vec{v}_1 &= \frac{\sqrt{2}}{\sqrt{5 + 2\sqrt{5}}} \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ \frac{1}{2} + \frac{\sqrt{5}}{2} \end{bmatrix} \\ &= \frac{\sqrt{2}}{\sqrt{5 + 2\sqrt{5}}} \begin{bmatrix} \frac{3}{2} + \frac{\sqrt{5}}{2} \\ 2 + \sqrt{5} \end{bmatrix}\end{aligned}$$

:

and so on.