

$$5.1 \quad A = \begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix} \quad A^* = \begin{bmatrix} 1 & 0 \\ 2 & 2 \end{bmatrix}$$

$$A^* A = \begin{bmatrix} 5 & 4 \\ 4 & 4 \end{bmatrix}$$

$$A^* A - \lambda I = 0 \Rightarrow \begin{bmatrix} 5-\lambda & 4 \\ 4 & 4-\lambda \end{bmatrix} = 0$$

$$\Rightarrow (\lambda-5)(\lambda-4) - 16 = 0$$

$$\Rightarrow \lambda^2 - 9\lambda + 4 = 0$$

$$\lambda = \frac{9 \pm \sqrt{81-16}}{2} = \frac{9 \pm \sqrt{65}}{2}$$

$$\therefore \sigma_{\max} = \left(\frac{9 + \sqrt{65}}{2} \right)^{1/2} \quad \sigma_{\min} = \left(\frac{9 - \sqrt{65}}{2} \right)^{1/2}$$

5.2 T.P: Set of full rank matrices is a dense subset of $\mathbb{C}^{m \times n}$.

Use $\|\cdot\|_2$ for proof

↓
we need to show that if we perturb
A by ε to get A_ε
 $\|A - A_\varepsilon\|_2 \rightarrow 0$ as $\varepsilon \rightarrow 0$

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

$$A_\varepsilon \triangleq U (\Sigma + \varepsilon I_{m \times n}) V^*$$

We know that if Q is unitary, $\|QA\|_2 = \|A\|_2$

so $\|A\|_2 = \|\Sigma\|_2$ for the SVD.

$$\|A - A_\varepsilon\|_2 = \|U (\varepsilon I_{m \times n}) V^*\|_2 = \varepsilon$$

$$\therefore \lim_{\varepsilon \rightarrow 0} \|A - A_\varepsilon\|_2 = 0$$

$$5.3 \quad A = \begin{bmatrix} -2 & 11 \\ -10 & 5 \end{bmatrix}$$

$$(a) \quad A^*A = \begin{bmatrix} -2 & -10 \\ 11 & 5 \end{bmatrix} \begin{bmatrix} -2 & 11 \\ -10 & 5 \end{bmatrix} = \begin{bmatrix} 104 & -72 \\ -72 & 146 \end{bmatrix}$$

$$\det(\lambda I - A^*A) = (\lambda - 104)(\lambda - 146) - 72 \cdot 72 = 0$$

$$\Rightarrow \lambda^2 - 250\lambda + (146 \cdot 104 - 72 \cdot 72) = 0$$

$$\Rightarrow \lambda^2 - 250\lambda + 10000 = 0$$

$$\lambda = 50, 200$$

$$\sigma_1 = \sqrt{200} = 10\sqrt{2}$$

$$\sigma_2 = \sqrt{50} = 5\sqrt{2}$$

$$\text{Also, } AA^* = \begin{bmatrix} 125 & 75 \\ 75 & 125 \end{bmatrix}$$

Columns of U and V are eigenvectors of A^*A and AA^* , respectively (recall Ex 4.1)

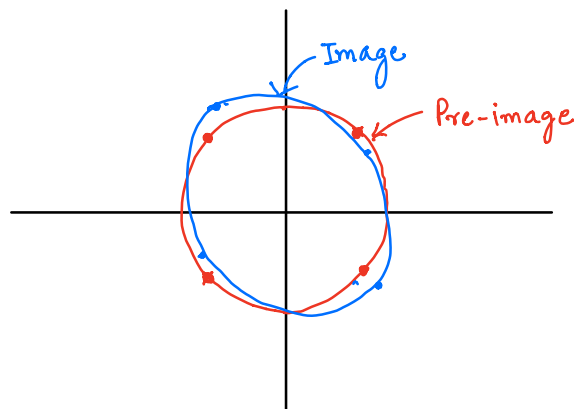
We can compute them to get:

$$U = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \quad \text{and} \quad V = \frac{1}{5} \begin{bmatrix} -3 & 4 \\ 4 & 3 \end{bmatrix}$$

$$(b) \quad \sigma_1 = 10\sqrt{2}, \quad \sigma_2 = 5\sqrt{2}$$

$$\text{Left singular vectors: } \pm \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}, \pm \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

$$\text{Right singular vectors: } \pm \begin{bmatrix} -3/5 \\ 4/5 \end{bmatrix}, \pm \begin{bmatrix} 4/5 \\ 3/5 \end{bmatrix}$$



$$(c) \|A\|_2 = \sigma_{\max} = 10\sqrt{2}$$

$$\|A\|_F = \sqrt{\sigma_1^2 + \sigma_2^2} = 5\sqrt{10}$$

$$\|A\|_1 = \max^m \text{ column sum of } |A| = \max(12, 16) = 16$$

$$\|A\|_\infty = \max^m \text{ row sum of } |A| = \max(13, 15) = 15$$

$$(d) A^{-1} = (U \Sigma V^*)^{-1} = V \Sigma^{-1} U^*$$

$$= \frac{1}{5\sqrt{2}} \begin{bmatrix} -3 & 4 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} \frac{1}{10\sqrt{2}} & 0 \\ 0 & \frac{1}{5\sqrt{2}} \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$= \frac{1}{100} \begin{bmatrix} -3 & 4 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$= \frac{1}{100} \begin{bmatrix} 5 & -11 \\ 10 & -2 \end{bmatrix}$$

$$(e) \det(\lambda I - A) = 0$$

$$\Rightarrow (\lambda + 2)(\lambda - 5) + 110 = 0$$

$$\Rightarrow \lambda^2 - 3\lambda + 110 = 0$$

$$\therefore \lambda = \frac{3 \pm \sqrt{9 - 440}}{2} = \frac{3 \pm i\sqrt{431}}{2}$$

(f) can be easily verified

$$\boxed{\det A = \lambda_1 \lambda_2}$$

$$\boxed{|\det A| = \sigma_1 \sigma_2}$$

(g) The axes of the ellipsoid are σ_1 and σ_2 .

$$\text{Area} = \pi \cdot \sigma_1 \cdot \sigma_2 = 100\pi$$



5.4 $A \in \mathbb{C}^{m \times m}$ $A = U \Sigma V^*$

$$B = \begin{bmatrix} 0 & A^* \\ A & 0 \end{bmatrix} \quad B^* = B$$

$$B = \underbrace{X}_{\text{contain linearly independent eigenvectors of } B} \underbrace{\Lambda}_{\text{diagonal matrix whose entries are the eigenvalues of } B} X^{-1}$$

We cannot compute these matrices directly since we don't know A . So we must use properties.

$$B = \begin{bmatrix} 0 & A^* \\ A & 0 \end{bmatrix} = \begin{bmatrix} 0 & V \Sigma^* U^* \\ U \Sigma V^* & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & I_{m \times m} \\ I_{m \times m} & 0 \end{bmatrix} \begin{bmatrix} U \Sigma V^* & 0 \\ 0 & V \Sigma^* U^* \end{bmatrix}$$

(simple row transformation to get a diagonal looking matrix)

$$= \begin{bmatrix} 0 & I_m \\ I_m & 0 \end{bmatrix} \begin{bmatrix} U & 0 \\ 0 & V \end{bmatrix} \begin{bmatrix} \Sigma & 0 \\ 0 & \Sigma^* \end{bmatrix} \underbrace{\begin{bmatrix} V^* & 0 \\ 0 & U^* \end{bmatrix}}_{\text{suppose this is } X^{-1}, \text{ then if this is } X, \text{ we are done. Check!}}$$

suppose this is X , then if this is X^{-1} , we are done. Check!

$$X^{-1} = \begin{bmatrix} U^* & 0 \\ 0 & V^* \end{bmatrix} \begin{bmatrix} 0 & I_m \\ I_m & 0 \end{bmatrix} = \begin{bmatrix} 0 & U^* \\ V^* & 0 \end{bmatrix} \quad \text{not the same!}$$

BUT

$$\begin{bmatrix} V^* & 0 \\ 0 & U^* \end{bmatrix} = \begin{bmatrix} 0 & I_m \\ I_m & 0 \end{bmatrix} X^{-1}$$

Plug this in

$$B = X \begin{bmatrix} \Sigma & 0 \\ 0 & \Sigma^* \end{bmatrix} \begin{bmatrix} 0 & I_m \\ I_m & 0 \end{bmatrix} X^{-1}$$

$$B = X \begin{bmatrix} 0 & \Sigma \\ \Sigma & 0 \end{bmatrix} X^{-1} \quad \{ \Sigma = Z^* \}$$

this is still not eigenvalue decomposition because Λ needs to be a diagonal matrix

need to do some matrix multiplication magic and update X and X^{-1} .

$$\begin{bmatrix} 0 & \Sigma \\ \Sigma & 0 \end{bmatrix} = \frac{1}{2} \underbrace{\begin{bmatrix} I_m & I_m \\ I_m & -I_m \end{bmatrix}}_{\text{need to have these equal so}} \underbrace{\begin{bmatrix} \Sigma & 0 \\ 0 & -\Sigma \end{bmatrix}}_{\text{that } X \text{ and } X^{-1} \text{ are not}} \underbrace{\begin{bmatrix} I_m & I_m \\ I_m & -I_m \end{bmatrix}}_{\text{broken}}$$

$$\therefore B = X \begin{bmatrix} \Sigma & 0 \\ 0 & -\Sigma \end{bmatrix} X^{-1}$$

$$\text{for } X = \frac{1}{\sqrt{2}} \begin{bmatrix} 0 & I_m \\ I_m & 0 \end{bmatrix} \begin{bmatrix} U & 0 \\ 0 & V \end{bmatrix} \begin{bmatrix} I_m & I_m \\ I_m & -I_m \end{bmatrix}$$