

Discussion 6A

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Q1.

$$A = \begin{bmatrix} 1 & -1 \\ -2 & 2 \\ 2 & -2 \end{bmatrix} = \begin{matrix} \begin{matrix} 3 \times 2 \\ \uparrow \\ 3 \times 3 \end{matrix} \\ U \Sigma V^T \\ \begin{matrix} \downarrow \\ 3 \times 3 \end{matrix} \end{matrix}, \text{rank}(A) = 1$$

columns are linearly dependent
↓

a) (i) $A^T A = \begin{bmatrix} 1 & -2 & 2 \\ -1 & 2 & -2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ -2 & 2 \\ 2 & -2 \end{bmatrix} = \begin{bmatrix} 9 & -9 \\ -9 & 9 \end{bmatrix}$

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

$$\det \begin{pmatrix} 9-\lambda & -9 \\ -9 & 9-\lambda \end{pmatrix} = (9-\lambda)^2 - 81 = \lambda^2 - 18\lambda = \lambda(\lambda - 18) = 0$$

$$\therefore \lambda_1 = 18, \lambda_2 = 0$$

Corresponding eigenvectors

$$\lambda_1 = 18 \rightarrow \begin{bmatrix} -9 & -9 \\ -9 & -9 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 0 \quad -9v_1 - 9v_2 = 0 \quad v_1 = -v_2 \rightarrow \vec{v}_1 = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

$$\lambda_2 = 0$$

$$\rightarrow \vec{v}_2 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

(ii) $\Lambda = \begin{bmatrix} 18 & 0 \\ 0 & 0 \end{bmatrix} \quad V = \begin{bmatrix} \vec{v}_1 & \vec{v}_2 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$

unpack Λ
i.e. Λ_{ii}
→ diagonal matrix of $\sqrt{\lambda_i}$ for $i=1$ to r

$$\Lambda_r = [18]$$

$r=1 \Rightarrow$ only have 1 entry (i.e. λ)

square root all the entries

(iii) $\Sigma_r = \Lambda_r^{1/2} = [\sqrt{18}]$

$$\Sigma = \begin{bmatrix} \sqrt{18} & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}_{3 \times 2}$$

fill in 0s such that dimensionality matches w/ A

(iv) So far, we found

$$\begin{cases} V_r = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} \\ \Sigma_r = [\sqrt{18}] \end{cases} \quad \begin{matrix} V = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \\ \Sigma = \begin{bmatrix} \sqrt{18} & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \end{matrix}$$

→ Find U_r

$$A = U_r \Sigma_r V_r^T \Rightarrow A V_r = U_r \Sigma_r V_r^T V_r = U_r \Sigma_r$$

$$\Rightarrow U_r = A V_r \Sigma_r^{-1} = \begin{bmatrix} -1/3 \\ 2/3 \\ -2/3 \end{bmatrix}$$

But for full SVD: $U \rightarrow$ Gram Schmidt!
 3×3

$$\begin{bmatrix} -1/3 & 1 & 0 & 0 \\ 2/3 & 0 & 1 & 0 \\ -2/3 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{\text{G.S.}} \begin{bmatrix} -1/3 & \sqrt{3}/3 & 0 \\ 2/3 & 1/\sqrt{3} & 1/\sqrt{3} \\ -2/3 & -1/\sqrt{3} & 1/\sqrt{3} \end{bmatrix} = U$$

(v) • Full SVD = $U \Sigma V^T = \begin{bmatrix} -1/3 & \sqrt{3}/3 & 0 \\ 2/3 & 1/\sqrt{3} & 1/\sqrt{3} \\ -2/3 & -1/\sqrt{3} & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} \sqrt{18} & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \left(\begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \right)^T$

• Compact SVD = $U_r \Sigma_r V_r^T = \begin{bmatrix} -1/3 \\ 2/3 \\ -2/3 \end{bmatrix} [\sqrt{18}] \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}^T$

b) SVD of A^T

$$3 \times 4 \quad A = U \Sigma V^T$$

$$2 \times 3 \quad A^T = (U \Sigma V^T)^T = V \Sigma^T U^T$$

$$= \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} \sqrt{18} & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} -1/3 & \sqrt{3}/3 & 0 \\ 2/3 & 1/3\sqrt{2} & 1/\sqrt{2} \\ -2/3 & -1/3\sqrt{2} & 1/\sqrt{2} \end{bmatrix}^T$$

2×2 2×3 3×3

→ SVD dimensionality ch. to match the dimensionality of A^T

c) WTS: $\text{Null}(A) = \text{Col}(V_{n-r})$

$$\text{We can write } A = \sum_{i=1}^r \sigma_i u_i v_i^T \quad (\text{outer product form})$$

$$\text{Null}(A) \text{ means } A\vec{x} = 0, \vec{x} \neq 0$$

$$A\vec{x} = \sum_{i=1}^r \sigma_i u_i \underbrace{v_i^T \vec{x}}_0$$

What vector \vec{x} makes $v_i^T \vec{x} = 0$?

→ orthogonal vector to \vec{v}_i

⇒ \vec{x} must be orthogonal to \vec{v}_i for $i=1 \dots r$ (sum from 1 to r)

$$\vec{x} \perp \text{Col}(V_r)$$

$$\vec{x} \in \text{Col}(V_{n-r})$$

↳ V is orthogonal

$$V = \begin{bmatrix} V_r & V_{n-r} \end{bmatrix}$$

orthogonal to V_r !

$$\Rightarrow \text{Null}(A) = \text{Col}(V_{n-r})$$

d) $\text{Col}(A) = \text{Col}(U_r)$

$$A = \sum_{i=1}^r b_i \vec{u}_i \vec{v}_i^T$$

$\text{Col}(A)$ means that for a given ^{nonzero} vector \vec{b} , $A\vec{x} = \vec{b}$

$$\therefore A\vec{x} = \sum_{i=1}^r b_i \underbrace{\vec{u}_i \vec{v}_i^T \vec{x}}_{\neq 0} \quad \left. \begin{array}{l} \text{scalar} \\ \downarrow \\ \neq 0 \end{array} \right\}$$

$$= \sum_{i=1}^r (\vec{v}_i^T \vec{x}) b_i \vec{u}_i$$

$$= \sum_{i=1}^r \alpha_i \vec{u}_i \rightarrow \text{linear combination of } \vec{u}_1, \dots, \vec{u}_r$$

$$\therefore A\vec{x} = \vec{b} \in \text{col}(U_r)$$

$$\therefore \text{Col}(A) = \text{col}(U_r)$$

f) AA^T

$$r = \text{rank}(A)$$

$$(U, \Sigma) = \text{diagonalize}(AA^T)$$

$$\text{Unpack } U = [U_r \ U_{n-r}]$$

$$\text{Unpack } \Sigma = \begin{bmatrix} \Sigma_r & 0 \\ 0 & 0 \end{bmatrix}$$

$$\Sigma_r = \Sigma_r^{1/2}$$

Part 2

$$V_r = A^T U_r \Sigma_r^{-1}$$

$V = \text{extend basis } (V_r, \mathbb{R}^n)$

return (U, Σ, V)