

矩阵理论与方法

12月

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- 课程介绍
- 矩阵理论与方法

第4章 矩阵分解

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1、LU分解

2、QR分解

3、满秩分解

4、SVD分解

定义 4.11 设 $A \in \mathbf{C}_r^{m \times n} (r > 0)$, $A^H A$ 的特征值为

$$\lambda_1 \geq \lambda_2 \geq \cdots \geq \lambda_r > \lambda_{r+1} = \cdots = \lambda_n = 0$$

则称 $\sigma_i = \sqrt{\lambda_i} (i = 1, 2, \cdots, n)$ 为 A 的奇异值; 当 A 为零矩阵时, 它的奇异值都是 0.

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定理 4.16 设 $A \in \mathbf{C}_r^{m \times n} (r > 0)$, 则存在 m 阶酉矩阵 U 和 n 阶酉矩阵 V , 使得

$$U^H A V = \begin{bmatrix} \Sigma & \mathbf{O} \\ \mathbf{O} & \mathbf{O} \end{bmatrix} = D \quad (4.4.4)$$

其中 $\Sigma = \text{diag}(\sigma_1, \sigma_2, \cdots, \sigma_r)$, 而 $\sigma_i (i = 1, 2, \cdots, r)$ 为矩阵 A 的全部非零奇异值.

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得到 $\lambda_k, x_k : Bx_k = \lambda_k x_k, k = 1, 2, \dots, N$

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$V = (v_1, \dots, v_N)$ 是酉矩阵: $V^H V = I$

$$BV = V\Lambda \Rightarrow A^H AV = V\Lambda \Rightarrow A^H AV_1 = V_1 \Sigma^2$$

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$$1、 A^H A V = V \Lambda \Rightarrow A^H A V_1 = V_1 \Sigma^2$$

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令 $A \in \mathbb{C}^{M \times N}$, 求酉矩阵 U, V , 使得

$$A = U D V^H$$

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$$\text{有 } A^H U_1 = V_1 \Sigma$$

$$\Rightarrow U_1^H A = \Sigma V_1^H \Rightarrow A = U_1 \Sigma V_1^H$$

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$$2、 \text{令 } U_1 = AV_1\Sigma^{-1} \Rightarrow A = U_1\Sigma V_1^H$$

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$$3、 \text{将 } U_1 \text{ 扩充成酉矩阵 } U = [U_1 : U_2]$$

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例10: 称 $A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$, 求 $A = UDV^T$

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解: $AA^T = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 2 \end{bmatrix} = B, |\lambda I - B| = \lambda(\lambda - 1)(\lambda - 3)$

$$\lambda_1 = 3: 3I - B = \begin{bmatrix} 2 & 0 & -1 \\ 0 & 2 & -1 \\ -1 & -1 & 1 \end{bmatrix}, \xi_1 = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$$

$$\lambda_2 = 1: 1I - B = \begin{bmatrix} 0 & 0 & -1 \\ 0 & 0 & -1 \\ -1 & -1 & -1 \end{bmatrix}, \xi_2 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$\lambda_3 = 0 : \quad 0I - B = \begin{bmatrix} -1 & 0 & -1 \\ 0 & -1 & -1 \\ -1 & -1 & -2 \end{bmatrix}, \xi_3 = \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$$

$$r_A = 2 : \quad \Sigma = \begin{bmatrix} \sqrt{3} & 0 \\ 0 & 1 \end{bmatrix}$$

$$\begin{aligned} 1、 A^H AV &= V\Lambda \Rightarrow A^H AV_1 = V_1\Sigma^2 \\ \Rightarrow A^H AV_1\Sigma^{-1} &= V_1\Sigma \end{aligned}$$

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$$V = \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} \\ \frac{2}{\sqrt{6}} & 0 & -\frac{1}{\sqrt{3}} \end{bmatrix}, \quad V_1 = \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} \\ \frac{2}{\sqrt{6}} & 0 \end{bmatrix}$$

$$U_1 = AV_1\Sigma^{-1} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ 0 & 0 \end{bmatrix}$$

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3、将 U_1 扩充成酉矩阵 $U = [U_1 : U_2]$

$$A = UDV^H$$

$$U_1 = AV_1\Sigma^{-1} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ 0 & 0 \end{bmatrix}, \text{ 取 } U_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \text{ 则 } U = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$U^T AV = \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} = D, \quad A = UDV^T$$

3、将 U_1 扩充成酉矩阵 $U = [U_1 : U_2]$

$$A = UDV^H$$

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

Diagram illustrating the full SVD decomposition of matrix A (size $m \times n$) into three matrices: U (size $m \times m$), Σ (size $m \times n$), and V^T (size $n \times n$).

$$A_{m \times n} \approx U_{m \times r} \Sigma_{r \times r} V^T_{r \times n}$$

Diagram illustrating the truncated SVD decomposition of matrix A (size $m \times n$) into three matrices: U (size $m \times r$), Σ (size $r \times r$), and V^T (size $r \times n$).

求 $A = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 0 & 0 \end{bmatrix}$ 的奇异值分解

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$$U = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 & -2 \\ 2 & 1 \end{bmatrix}, V = I, A = U \begin{bmatrix} \sqrt{5} & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} V^H$$

作业（第五版）

- 1、定义：4.11
- 2、定理：4.15、4.16
- 3、例题：4.14、4.15
- 4、习题4.4：2、4

作业（第三版）

- 1、定义：4.11
- 2、定理：4.15、4.16
- 3、例题：4.14、4.15
- 4、习题4.4：2、4

下课，谢谢大家！