

Question 1 Visualize in 2D

Find orthogonal matrix $U \in \mathbb{R}^{d \times 2}$ that maximize variance of data x_i (L_2 -norm squared of $AA^T x_i$) onto 2D space K

$$\hat{U} = \arg \max_{A \in \mathbb{R}^{d \times 2} \mid A^T A = I_2} \sum_{i=1}^n \|AA^T x_i\|_2^2$$

we can find \hat{U} by singular value decomposition of data $X \in \mathbb{R}^{d \times n}$

$$X = USV^T \quad U \in \mathbb{R}^{d \times d} \quad S \in \mathbb{R}^{d \times d} \quad V^T \in \mathbb{R}^{d \times n}$$

$$\Rightarrow \hat{U} = U_2 \quad \text{the first 2 column of } U$$

then calculate coordinate α

$$\tilde{\alpha} = U^T x_i = (\alpha_1, \alpha_2)$$

then we plot on a scatter plot

$$\begin{cases} x_1 \text{ axis} : \alpha_1 \\ x_2 \text{ axis} : \alpha_2 \end{cases}$$

Question 2 : low-rank approximation

we want to find a low-rank matrix approximation of X

$$\hat{X}_2 = \arg \min_{Y \mid \text{rank}(Y) \leq 2} \|Y - X\|_F^2$$

$$\Rightarrow \hat{X}_2 = U_2 S_2 V_2^T$$