

$$X \in \mathbb{R}^{m \times n}$$

$$x_{ij} \in X$$

$$x_j^{(i)} \in X$$

$$i \in 1, \dots, m$$

$$j \in 1, \dots, n$$

$$p=1 \quad (p < n)$$

$$x^{(i)} \in \mathbb{R}^{1 \times n}$$

$$U_1 \in \mathbb{R}^{n \times 1}$$

PROJECTION OF  $x^{(i)}$

$$x^{(i)} \cdot U_1 \in \mathbb{R}$$

$$\text{MEAN } \bar{X}$$

PROJECTION OF MEAN  $\bar{X}$

$$\bar{X} \cdot U_1 \in \mathbb{R}$$

$$\text{VARIANCE OF PROJECTED DATA } \sigma_p^2 = \frac{1}{m-1} \sum_{i=1}^m \left( x^{(i)} \cdot U_1 - \bar{X} \cdot U_1 \right)^2 =$$

$S_0$  COVARIANCE MATRIX OF ORIGINAL DATA

$$= U_1^T \cdot S_0 \cdot U_1$$

MAXIMIZE THE VARIANCE

$$\underset{U_1}{\operatorname{argmax}} \sigma_p^2 = \underset{U_1}{\operatorname{argmax}} U_1^T \cdot S_0 \cdot U_1 \in \mathbb{R}$$

$$\cancel{\|U_1\| \rightarrow \infty} \quad \|U_1\| = 1 \quad U_1^T \cdot U_1 = 1 = \sum_{i=1}^n u_{1i}^2$$

$$\underset{U_1}{\operatorname{argmax}} \left( U_1^T S_0 U_1 + \lambda (1 - U_1^T U_1) \right)$$

↓ DERIVATIVE WRT  $U_1$

$$(2 U_1^T S_0 + 0 - 2 \lambda U_1^T) = 0$$

↓

$$U_1^T S_0 = \lambda U_1^T$$

$\lambda$  EIGENVALUE  
 $U_1$  EIGENVECTOR

RIGHT MULTIPLY BY  $U_1$

$$U_1^T S_0 U_1 = \lambda U_1^T U_1$$

$$\boxed{U_1^T S_0 U_1} = \lambda$$

$$\sigma_p^2 = \lambda$$

$$\max \sigma_p^2 = \max_i \lambda_{(i)}$$

↓

CORRESPONDING EIGENVECTOR

$$p=L$$

$$\vdots$$

$$p=n$$