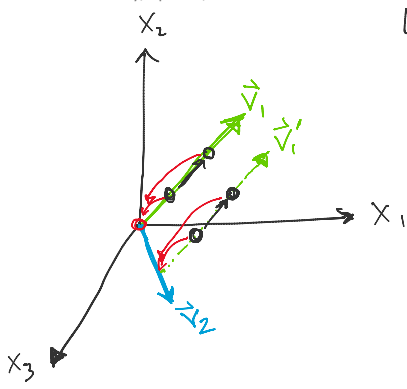


Section B

Wednesday, April 12, 2023 12:31 PM



M=3
K=2

$$\vec{U}^{(i)} = V \vec{X}^{(i)}$$

$$U_1^{(i)} = \vec{V}_1^T \vec{X}^{(i)}$$

$$U_2^{(i)} = \vec{V}_2^T \vec{X}^{(i)}$$

$$\vdots$$

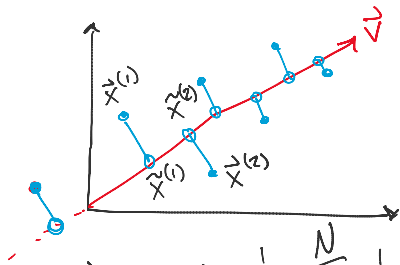
$$U_k^{(i)} = \vec{V}_k^T \vec{X}^{(i)}$$

$$V = \begin{bmatrix} \vec{V}_1^T \\ \vec{V}_2^T \\ \vdots \\ \vec{V}_k^T \end{bmatrix}$$

Objectives for PCA

What is the first principal component for PCA?
 \vec{V}_1

① Minimize the Reconstruction Error

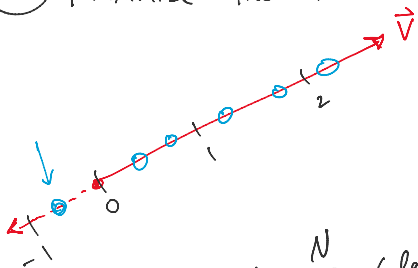


$$\vec{V}_1 = \underset{\vec{V}}{\operatorname{argmin}} \frac{1}{N} \sum_{i=1}^N \text{distance}(\vec{X}^{(i)}, \tilde{X}^{(i)})^2$$

$$= \underset{\vec{V}}{\operatorname{argmin}} \frac{1}{N} \sum_{i=1}^N \left\| \vec{X}^{(i)} - \underbrace{\left(\text{vector projection of } \vec{X}^{(i)} \text{ onto } \vec{V} \right)}_{\left(\frac{\vec{V}^T \vec{X}^{(i)}}{\|\vec{V}\|} \right) \frac{\vec{V}}{\|\vec{V}\|}} \right\|_2^2$$

$$= \underset{\vec{V} \text{ s.t. } \|\vec{V}\|_2 = 1}{\operatorname{argmin}} \frac{1}{N} \sum_{i=1}^N \left\| \vec{X}^{(i)} - \left(\vec{V}^T \vec{X}^{(i)} \right) \frac{\vec{V}}{\|\vec{V}\|} \right\|_2^2$$

② Maximize the Variance



$$\vec{V}_1 = \underset{\vec{V}}{\operatorname{argmax}} \frac{1}{N} \sum_{i=1}^N \left(\text{length of vector proj. of } \vec{X}^{(i)} \text{ onto } \vec{V} \right)^2$$

$$= \underset{\vec{V} \text{ s.t. } \|\vec{V}\| = 1}{\operatorname{argmax}} \frac{1}{N} \sum_{i=1}^N \left(\vec{V}^T \vec{X}^{(i)} \right)^2$$

$$= \operatorname{argmax}_{\vec{v}} \frac{1}{N} \sum_{i=1}^N (\vec{v}^T \vec{x}^{(i)})$$

s.t. $\|\vec{v}\|_2 = 1$

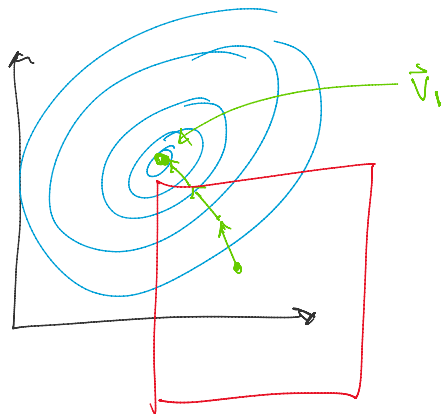
$$= \operatorname{argmax}_{\vec{v}} \frac{1}{N} (\vec{v}^T X^T) (X \vec{v})$$

s.t. $\|\vec{v}\|_2 = 1$

b/c $\Sigma = \frac{1}{N} X^T X$

$$= \operatorname{argmax}_{\vec{v}} \vec{v}^T \Sigma \vec{v}$$

s.t. $\|\vec{v}\|_2 = 1$



OH

BN

s^1	s^2
s^3	

$$(s^1, \rightarrow, 7, s^3)$$

$$(s^1, \rightarrow, 7, s^2)$$

$$P(s^3 | s^1, \rightarrow) = 0.5$$

$$P(s^2 | s^1, \rightarrow) = 0.5$$

$$Q(s^1, \rightarrow) = 7 + \lambda Q(s^3, \ominus)$$

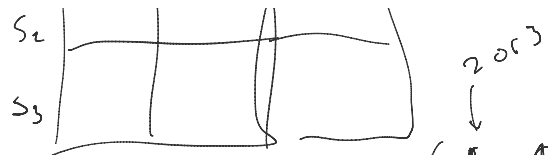
$$7 + \lambda Q(s^2, \ominus)$$

	\rightarrow	\downarrow	\ominus
s_1			
s_2			

2 or 3

Q1-1

$$Q(s', \rightarrow) = \gamma + \lambda Q(s^2, \circlearrowleft)$$



$$Q(s', \rightarrow) = (1 - \alpha_t) Q(s', \rightarrow) + (\alpha_t) (\gamma + Q(s^*, \circlearrowleft))$$

$$Q^*(s, a) = \max_a R(s, a) + \lambda \underbrace{\sum_{s'} p(s'|s, a)}_{\text{2 or 3}} \left[\max_{a'} Q^*(s', a') \right]$$

if $T=1$

$$ER = \vec{e} = [ER(1), ER(2), \dots, ER(k)]$$

$$\vec{p} = \text{softmax}(\vec{e}/T)$$

$$p(a) = p_a$$

$$p(2) = p_2$$



\vec{p} when $T=100$
 \vec{p} when $T=2$
 \vec{p} when $T=1$
 \vec{p} when $T=0.1$