

Yizhou Lin lin773@misc.edu

1. a) $\Sigma = X^T = (U \Sigma V^T)^T = V \Sigma U^T$

b) orthonormal basis: $U^{-1} \Sigma^{-1} U^{-T}$

2. a) for $y = Xw$, the soln. must be unique if
 $\text{rank}\{X\} = \text{rank}\{X|y\}$ and $\text{rank}\{X\} = \dim\{w\}$

b) $\hat{y} = (X^T X + \lambda I)^{-1} X^T y$

c) when we need a unique soln., $\text{rank}\{X\} = \text{rank}\{X|y\}$ &
 $\text{rank}\{X\} = \dim\{w\}$

3. a) Since $X^T X = (U \Sigma V^T)^T (U \Sigma V^T)$
 $= V \Sigma U^T U \Sigma V^T$
 $= V \Sigma^2 V^T$

$$\lambda I = V \lambda I V^T$$

then $(X^T X + \lambda I)^{-1} X^T = (V \Sigma^2 V^T + V \lambda I V^T)^{-1} X^T$

$$= \sum_{i=1}^p \frac{1}{\sigma_i^2 + \lambda} u_i u_i^T$$

b) if $X^T X$ is invertible, as $\lambda \rightarrow 0$

we get $X^+ = (X^T X + 0\lambda)^{-1} X^T$
 $= (X^T X)^{-1} X^T$

c) if X is square & invertible,

$$X^+ = (X^T X)^{-1} X^T = X^{-1} X^{-T} X^T = X^{-1}$$

d) if $r < p$, $\lambda > 0$.

$$(X^T X + \lambda I)^{-1} X^T = \sum_{i=1}^p \frac{\sigma_i^2}{\sigma_i^2 + \lambda} U_i V_i^T$$

e) if $r < p$; $\pi^+ = \lim_{\lambda \downarrow 0} (X^T X + \lambda I)^{-1} X^T$

$$= (X^T X)^{-1} X^T$$

$$= \begin{bmatrix} X^{-1} & X^{-T} X^T \end{bmatrix}$$

$$= (U \Sigma U^T)^{-1} (U \Sigma V^T)^{-T} (U \Sigma U^T)^T$$

$$= V^{-T} \Sigma^{-1} U^T (U \Sigma U^T)^{-T} (U \Sigma U^T)$$

$$= V \Sigma^{-1} U^T$$

$$= \sum_{i=1}^r \frac{1}{\sigma_i} V_i U_i^T$$

4. a) Yes; No.

b) No. Since one of the subspace's center is not close to 0.

c) Yes.

Removing the mean make the data more close the origin center, thereby increasing the PCA's derived effect.