

## Part3 Lab3

Saturday, December 4, 2021

11:27 PM

2-1

$$y = x + e \in \mathbb{R}^p$$

$$E_x = \mu \quad E(x - \mu)(x - \mu)^T = \Sigma$$

$$E_e = 0 \quad E_e e^T = \sigma^2 I_{p \times p}$$

From noisy  $y \rightarrow \hat{x}$

$$\min_{\hat{x}} E[\|\hat{x} - x\|^2 | y]$$

$$\hat{x} = E[x|y]$$

Alternatively, we added

$$\hat{x} = \mu + H(y - \mu)$$

$H$ :  $p \times p$  matrix

$$\min_{H \in \mathbb{R}^{p \times p}} E[\|x - (\mu + H(y - \mu))\|^2]$$

$$H = \Sigma(\Sigma + \sigma^2 I)^{-1}$$

Step 1

$$J = E[\|x - \mu - H(x - \mu + e)\|^2]$$

$$= E \text{Tr}[(I - H)(x - \mu)(x - \mu)^T(I - H)^T + H e e^T H^T]$$

$$= \text{Tr}[(I - H)\Sigma(I - H)^T + \sigma^2 H H^T]$$

$$\|H\|^2 = x^T x = \text{Tr}[x^T x] = \text{Tr}(x x^T)$$

Step 2

$$\left(\frac{d}{dH} \text{Tr}(H B^T) = B\right) - 2(I - H)\Sigma + 2\sigma^2 H = 0$$

$$\Rightarrow \hat{x} = \mu + \Sigma(\Sigma + \sigma^2 I)^{-1}(y - \mu)$$