

2) We have the condition for least squares :

$$\| \vec{y} - \hat{y} \|^2 \min$$

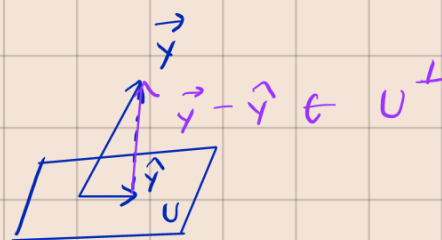
$$\Rightarrow SSR = \| \vec{y} - \hat{y} \|^2 \min \quad (1)$$

$$\text{where } \hat{y} = \begin{pmatrix} 1 & x_1 & x_1^2 \\ & \vdots & \\ 1 & x_n & x_n^2 \end{pmatrix} \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix} = A \cdot \hat{\beta}$$

$$\text{let } U = \text{Span} \left( \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}, \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \begin{pmatrix} x_1^2 \\ \vdots \\ x_n^2 \end{pmatrix} \right), \text{ we have } \hat{y} \in U$$

$$\text{From (1)} \Rightarrow \hat{y} = \text{proj}_U \vec{y}$$

$$\Rightarrow \vec{y} - \hat{y} = \text{proj}_{U^\perp} \vec{y}$$



From lecture, we have  $\hat{\beta} = (A^T A)^{-1} A^T \vec{y}$  ( $A^T A$  is invertible when  $x_i = i \ \forall i$ )

$$\Rightarrow A \hat{\beta} = \hat{y} = \underbrace{A}_{3 \times n} \underbrace{(A^T A)^{-1}}_{3 \times 3} \underbrace{A^T}_{n \times 3} \vec{y}$$

$$\Rightarrow \vec{y} - \hat{y} = \left( I_n - A (A^T A)^{-1} A^T \right) \vec{y}$$

Therefore, we have  $\text{proj}_{U^\perp} = I_n - A (A^T A)^{-1} A^T = B_1$

$$\text{and } \text{proj}_U = A (A^T A)^{-1} A^T = B_2$$

•) Consider  $(A(A^T A)^{-1} A^T)^T = A (A^T A)^{-T} A^T \quad (2)$

We have  $(A^T A)^T = A^T A \Rightarrow A^T A$  is symmetric

$\Rightarrow \exists$  an orthogonal matrix  $P$  and a diagonal matrix  $D$  s.t

$$P D P^T = A^T A$$

From (2)  $\Rightarrow (A (A^T A)^{-1} A^T)^{-T} = A \cdot (P D P^T)^{-T} A^T$

$$= A (P D^{-1} P^T)^T A^T = A (P D^{-1} P^T) A^T$$

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

Therefore  $B_1$  and  $B_2$  are symmetric

$$\|B_1 \vec{y}\|^2 = (B_1 \vec{y})^T (B_1 \vec{y}) = \vec{y}^T B_1^T B_1 \vec{y} \quad (3)$$

•) let  $\vec{y} \sim \mathcal{N}(\vec{\mu}_y, \Sigma_y)$

where  $\vec{\mu}_y = \begin{pmatrix} \beta_0 + \beta_1 x_1 + \beta_2 x_1^2 \\ \vdots \\ \beta_0 + \beta_1 x_n + \beta_2 x_n^2 \end{pmatrix}$

$$\Sigma_y = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix}$$

$$\Rightarrow \vec{z} = (\sqrt{\Sigma_y})^{-1} (y - \vec{\mu}_y) \sim \mathcal{N}(0, 1)$$

$$\Rightarrow \vec{Y} = I_n \cdot \vec{Z} + \vec{\mu}_Y = \vec{Z} + \vec{\mu}_Y$$

$$\text{From (3)} \Rightarrow SSR = (\beta_1 \vec{Y})^T \beta_1 \vec{Y}$$

$$= \vec{Y}^T \beta_1^T \beta_1 \vec{Y} = (\vec{Z} + \vec{\mu}_Y)^T \beta_1^T \beta_1 (\vec{Z} + \vec{\mu}_Y)$$

Because  $\beta_1$  is symmetric and is a projection matrix,

$$\beta_1^T \beta_1 = \beta_1^2 = \beta_1$$

$$\begin{aligned} \Rightarrow SSR &= (\vec{Z} + \vec{\mu}_Y)^T \beta_1 (\vec{Z} + \vec{\mu}_Y) = \vec{Z}^T \beta_1 \vec{Z} + \vec{Z}^T \beta_1 \vec{\mu}_Y \\ &+ \vec{\mu}_Y^T \beta_1 \vec{Z} + \vec{\mu}_Y^T \beta_1 \vec{\mu}_Y \end{aligned}$$

We have  $\beta_1 = \text{proj}_{U^\perp}$  and  $\vec{\mu}_Y \in U$

$$\Rightarrow \beta_1 \vec{\mu}_Y = 0 \quad \text{and} \quad \vec{\mu}_Y^T \beta_1 \vec{v} = \langle \vec{\mu}_Y, \beta_1 \vec{v} \rangle = 0$$

$\forall v \in \mathbb{R}^n$

$$\text{Then } SSR = \vec{Z}^T \beta_1 \vec{Z}$$

We have  $\text{rank}(\beta_1) + \text{rank}(\beta_2) = \dim U + \dim U^\perp = n$

$\beta_1, \beta_2$  are symmetric and  $\beta_1 + \beta_2 = I_n$

$$\vec{Z} \sim \mathcal{N}(0, I_n)$$

By Cochran's Theorem, we have  $SSR \sim \chi^2(n-3)$