

1. Suponha  $\hat{\theta}_1$  e  $\hat{\theta}_2$  dois estimadores não-viesados para  $\theta$ , independentes.  
sejam  $\text{Var}(\hat{\theta}_1) = V_1$  e  $\text{Var}(\hat{\theta}_2) = V_2$ .

$\hat{\theta} = c_1 \hat{\theta}_1 + c_2 \hat{\theta}_2$  Qual proporcione um estimador não-viesado com EQM mínimo.  
 $\uparrow$   $\uparrow$

$$\text{EQM}(\hat{\theta}) = \underbrace{\text{Var}(\hat{\theta})}_{(2)} + \underbrace{\text{Viés}^2(\hat{\theta})}_{(1)}$$

a. Calculando o viés do  $\hat{\theta}$

$$\mathbb{E}(\hat{\theta}) = \mathbb{E}[c_1 \hat{\theta}_1 + c_2 \hat{\theta}_2] = c_1 \mathbb{E}\hat{\theta}_1 + c_2 \mathbb{E}\hat{\theta}_2 = c_1 \theta + c_2 \theta = (c_1 + c_2) \theta$$

para  $\hat{\theta}$  ser não-viesado, então  $c_1 + c_2 = 1$ .

b. Minimizando o  $EQM(\hat{\theta})$ .

b1. viés:  $\hat{\theta}$  é não-viesado quando  $c_1 + c_2 = 1$  ou  $c_2 = 1 - c_1$

b2. variância.

$$\begin{aligned}\text{Var}(\hat{\theta}) &= \text{Var}(c_1 \hat{\theta}_1 + c_2 \hat{\theta}_2) = c_1^2 \text{Var}(\hat{\theta}_1) + c_2^2 \text{Var}(\hat{\theta}_2) = c_1^2 \sigma_1 + c_2^2 \sigma_2 \\ &= c_1^2 \sigma_1 + (1 - c_1)^2 \sigma_2 //\end{aligned}$$

$$\frac{\partial \text{Var}(\hat{\theta})}{\partial c_1} = 2c_1 \sigma_1 + 2(1 - c_1)(-1)\sigma_2 = 2c_1 \sigma_1 - 2(1 - c_1)\sigma_2 \leftarrow$$

$$2c_1^* \sigma_1 - 2(1 - c_1^*) \sigma_2 = 0$$

$$\rightarrow c_1^* = \frac{\sigma_2}{\sigma_1 + \sigma_2}$$

Calculando a segunda derivada

$$\frac{\partial^2 \text{var}(\hat{\theta})}{\partial c_1^2} = 2\sigma_1 + 2\sigma_2 > 0, \text{ daí o } c_1^* \text{ é um } \underline{\text{mínimo}}!$$

Dessa forma

$$\hat{\theta} = c_1 \hat{\theta}_1 + c_2 \hat{\theta}_2 = \frac{\sigma_2}{\sigma_1 + \sigma_2} \hat{\theta}_1 + \underbrace{\left(1 - \frac{\sigma_2}{\sigma_1 + \sigma_2}\right)}_{\frac{\sigma_1}{\sigma_1 + \sigma_2}} \hat{\theta}_2 = \frac{\sigma_2}{\sigma_1 + \sigma_2} \hat{\theta}_1 + \frac{\sigma_1}{\sigma_1 + \sigma_2} \hat{\theta}_2 \quad \square$$

$$\langle c, x \rangle = \langle x, c \rangle = \sum c_i x_i$$

2. Considere  $y_i = \beta x_i + \varepsilon_i, i=1, 2, \dots, n$

$$y = X\beta + \varepsilon, \quad \mathbb{E}(\varepsilon | X) = 0 \quad \text{e} \quad \text{cov}(\varepsilon | X) = \sigma^2 I$$

$n \times 1$     $n \times 1$     $1 \times 1$     $n \times 1$

a. Qual o estimador linear com EQM mínimo para  $\beta$ ?

considere  $\hat{\beta} = c'y = c'(X\beta + \varepsilon) = c'X\beta + c'\varepsilon$

a1. vies

$$\mathbb{E}(\hat{\beta}) = \mathbb{E}(c'X\beta + c'\varepsilon) = \mathbb{E}(\beta \underbrace{c'X}) + \mathbb{E}(\underbrace{c'\varepsilon}) = \beta \mathbb{E}(c'X) + \mathbb{E}(c'\varepsilon)$$

$$= \beta c'X + c' \cancel{\mathbb{E}(\varepsilon)}^0 = \beta c'X.$$

$$\Rightarrow \boxed{\mathbb{E}(\hat{\beta}) = \beta c'X} \leftarrow \text{número}$$

$$\mathbb{E}(\hat{\beta} - \beta) = \beta c'X - \beta = \beta \underline{c'X - 1} \leftarrow \text{vies de } \hat{\beta}.$$

a<sub>2</sub>. Variância

$$\underline{\text{var}}(\hat{\beta}) = \text{var}(\underline{\beta'x} + c'\varepsilon) = \text{var}(c'\varepsilon) = \underline{\sigma^2 c'c}$$

$$\begin{aligned}\underline{\text{cov}}(\hat{\beta}) &= \text{cov}(c'\varepsilon) = \mathbb{E}[c'\varepsilon][c'\varepsilon]' = \mathbb{E}[c'\varepsilon\varepsilon'c] = c'\mathbb{E}(\varepsilon\varepsilon')c = c'\sigma^2 I c \\ &= \sigma^2 c'c\end{aligned}$$

Daí,

$$EQM(\hat{\beta}) = \text{var}(\hat{\beta}) + [\text{viés}(\hat{\beta})]^2 = \sigma^2 c'c + \beta^2 (c'x - 1)^2$$

$$EQM(\hat{\beta}) = \sigma^2 c'c + \beta^2 (c'x - 1)^2$$

→ número

Minimizando  $EQM(\hat{\beta})$

$X'X$  número.  
 $XX'$  matriz

$$\frac{\partial EQM(\hat{\beta})}{\partial \beta} = 2\sigma^2 \underline{\beta} + 2\beta^2 (\underline{C}'X - 1) \underline{X} \rightarrow \text{vetor}$$

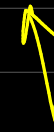
igualando a zero

$$2\sigma^2 \underline{\beta} + 2\beta^2 (\underline{C}'X - 1) \underline{X} = 0 \quad \text{ou}$$

$$\underline{\beta} = -\frac{\beta^2}{\sigma^2} (\underline{C}'X - 1) \underline{X}$$

$$\underline{\beta^2 (C'X - 1) X} = -\sigma^2 \underline{\beta}$$

$$\beta^2 (C'X - 1) X'X = -\sigma^2 X' \underline{\beta}$$



vetor

$$X' C^* = C^{*'} X$$

$$\langle X, C^* \rangle = \langle C^*, X \rangle$$

$$\beta^2 (\underbrace{C^{*'} X - 1}) X' X = -\sigma^2 \underbrace{X' C^*} = -\sigma^2 \underbrace{C^{*'} X}$$

$$\beta^2 C^{*'} X \cdot X' X - \underbrace{\beta^2 X' X} = -\sigma^2 \underbrace{C^{*'} X}$$

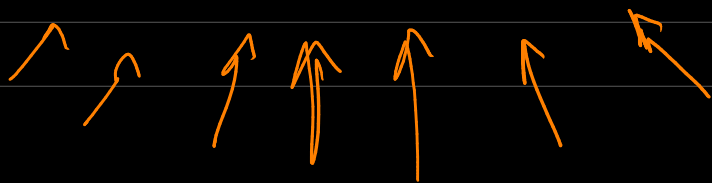
$$\beta^2 \underbrace{C^{*'} X} \cdot X' X + \sigma^2 \underbrace{C^{*'} X} = \beta^2 X' X$$

$$(\beta^2 X' X + \sigma^2) \underbrace{C^{*'} X} = \beta^2 \underbrace{X' X}$$

$$\underline{\underline{C^{*'} X}} = \frac{\beta^2 X' X}{\beta^2 X' X + \sigma^2}$$

Rescrevendo

$$C^* = -\frac{\beta^2}{\sigma^2} (C^{*'} X - 1) X$$



$$C^* = -\frac{\beta^2}{\sigma^2} (C^{*'}X - 1)X = -\frac{\beta^2}{\sigma^2} \left[ \frac{\beta^2 X'X}{\beta^2 X'X + \sigma^2} - 1 \right] X = -\frac{\beta^2}{\sigma^2} \left[ \frac{\cancel{\beta^2 X'X} - \cancel{\beta^2 X'X} - \sigma^2}{\sigma^2 + \beta^2 X'X} \right] X$$

$$= -\frac{\beta^2}{\sigma^2} \left[ \frac{-\sigma^2}{\sigma^2 + \beta^2 X'X} \right] X = \frac{\beta^2}{\sigma^2} \left[ \frac{\sigma^2}{\sigma^2 + \beta^2 X'X} \right] X$$

$$C^* = \frac{\beta^2}{\sigma^2} \left[ \frac{1}{1 + \frac{\beta^2}{\sigma^2} X'X} \right] X$$

$$C^* = \left[ \frac{1}{\frac{\sigma^2}{\beta^2} + X'X} \right] X$$

ponto crítico.

calculando a 2ª derivada.

$$\frac{\partial^2 EQM(\hat{\beta})}{\partial C' \partial C} = \underset{\uparrow}{2\sigma^2} + 2\beta^2 \underset{\downarrow}{X'X} > 0, \text{ daí } C^* \text{ é } \underline{\underline{\text{mínimo}}}$$



Finalmente,

$$\hat{\beta} = C^* Y = \underbrace{\begin{bmatrix} 1 \\ \frac{\sigma^2}{\beta^2 + X'X} \end{bmatrix}}_{\text{wavy line}} X'Y$$

↪ Eis 0  $\hat{\beta}$  ✓

b)  $\frac{EQM(\hat{\beta})}{EQM(\hat{\beta}_{MQ})} = \frac{\tau^2}{1 + \tau^2}$  ,  $\tau^2 = \frac{\beta^2}{\sigma^2 / X'X} = \frac{X'X \beta^2}{\sigma^2} = \frac{X'X}{\frac{\sigma^2}{\beta^2}}$

↓  $Cov(\hat{\beta}_{MQ}) = \sigma^2 (X'X)^{-1}$

↑  $\frac{\beta^2}{\sigma^2} (X'X)$

var de  $\hat{\beta}$

$$E\hat{\beta} = E \left[ \frac{1}{\frac{\sigma^2}{\beta^2} + X'X} X'Y \right] = \frac{1}{\frac{\sigma^2}{\beta^2} + X'X} X' EY = \frac{1}{\frac{\sigma^2}{\beta^2} + X'X} X'X\beta$$

$$E\hat{\beta} = \frac{\beta X'X}{\frac{\sigma^2}{\beta^2} + X'X}$$

var:  $E\hat{\beta} - \beta = \frac{\beta X'X}{\frac{\sigma^2}{\beta^2} + X'X} - \beta = \frac{\cancel{\beta X'X} - \beta \frac{\sigma^2}{\beta^2} - \cancel{\beta X'X}}{\frac{\sigma^2}{\beta^2} + X'X} = -\frac{\sigma^2}{\beta} \cdot \left[ \frac{1}{\frac{\sigma^2}{\beta^2} + X'X} \right]$

Variança  $\hat{\beta}$

$$\text{Var}(\hat{\beta}) = \text{Var} \left[ \frac{X'Y}{\frac{\sigma^2}{n} + X'X} \right] = \text{Var} \left[ \frac{X'(X\beta + \varepsilon)}{\frac{\sigma^2}{n} + X'X} \right] = \text{Var} \left[ \frac{\boxed{X'X\beta} + X'\varepsilon}{\frac{\sigma^2}{n} + X'X} \right]$$

$$= \text{Var} \left[ \frac{X'\varepsilon}{\frac{\sigma^2}{n} + X'X} \right] = \frac{1}{\left(\frac{\sigma^2}{n} + X'X\right)^2} \text{Var}(X'\varepsilon)$$

$$E(X'\varepsilon)(X'\varepsilon)' = E[X'\varepsilon\varepsilon'X] = X'\sigma^2 I X = \sigma^2 X'X$$

$$\rightarrow \text{Var}(\hat{\beta}) = \frac{1}{\left(\frac{\sigma^2}{n} + X'X\right)^2} \cdot \sigma^2 X'X //$$

$$EQM(\hat{\beta}) = \text{Var}(\hat{\beta}) + \text{bias}^2(\hat{\beta})$$

$$= \frac{\sigma^2(X'X)}{\left(\frac{\sigma^2}{\beta^2} + X'X\right)^2} + \frac{\sigma^4}{\beta^2} \cdot \frac{1}{\left(\frac{\sigma^2}{\beta^2} + X'X\right)^2} = \frac{1}{\left(\frac{\sigma^2}{\beta^2} + X'X\right)^2} \left(\sigma^2(X'X) + \frac{\sigma^4}{\beta^2}\right)$$

$$\frac{EQM(\hat{\beta})}{EQM(\hat{\beta}_{MLE})} = \frac{\left(\sigma^2(X'X) + \frac{\sigma^4}{\beta^2}\right) \cdot \frac{1}{\left(\frac{\sigma^2}{\beta^2} + X'X\right)^2}}{\sigma^2(X'X)^{-1}} = \frac{(X'X)^2 + \frac{\sigma^2}{\beta^2}(X'X)}{\left(\frac{\sigma^2}{\beta^2} + X'X\right)^2} = \frac{(X'X) \left[ X'X + \frac{\sigma^2}{\beta^2} \right]}{\left(\frac{\sigma^2}{\beta^2} + X'X\right)^2}$$

$\downarrow$   
 $\sigma^2 / X'X$

$$= \frac{X'X}{\frac{\sigma^2}{\beta^2} + X'X} = \frac{EQM(\hat{\beta})}{EQM(\hat{\beta}_{MLE})}$$

$$\frac{EQM(\hat{\beta})}{EQM(\hat{\beta}_{MQ})} = \frac{X'X}{\frac{\sigma^2}{\rho^2} + X'X} = \frac{\frac{\beta^2}{\sigma^2} X'X}{1 + \frac{\beta^2}{\sigma^2} X'X} = \frac{\tilde{\tau}^2}{1 + \tilde{\tau}^2} \quad \square$$

$$\tilde{\tau} \rightarrow \infty, \quad \frac{EQM(\hat{\beta})}{EQM(\hat{\beta}_{MQ})} \rightarrow 1.$$

## EXERCÍCIO PARA SEXTA

Considere o seguinte modelo de regressão.

$$EY_i = \beta_0 + \beta_1 x_i + \beta_2 (3x_i^2 - 2), \quad i=1,2,3$$

onde  $x_1 = -1$ ,  $x_2 = 0$  e  $x_3 = 1$ .

Encontre os estimadores MQO para  $\beta_0$ ,  $\beta_1$  e  $\beta_2$ .