

* FUNCIÓN DE REGRESIÓN POBLACIONAL

$$y_i = \beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + \dots + \beta_k x_{ki} + u_i$$

↑

v.a

$$\begin{aligned} \cdot E(y_i) &= E(\beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + \dots + \beta_k x_{ki} + u_i) = \\ &= E(\beta_1) + E(\beta_2 x_{2i}) + E(\beta_3 x_{3i}) + \dots + E(\beta_k x_{ki}) + E(u_i) = \\ &= \beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + \dots + \beta_k x_{ki} \end{aligned}$$

$$E(y_i) = \beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + \dots + \beta_k x_{ki}$$

$$\begin{aligned} \cdot \text{Var}(y_i) &= \text{Var}(\beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + \dots + \beta_k x_{ki} + u_i) = \\ &= \text{Var}(\beta_1 + \beta_2 x_{2i} + \beta_3 x_{3i} + \dots + \beta_k x_{ki}) + \text{Var}(u_i) = \\ &= \text{Var}(u_i) = \sigma_u^2 \end{aligned}$$

$$\text{Var}(y_i) = \text{Var}(u_i) = \sigma_u^2$$

↑

varianza del modelo

varianza de la regresión

→ Varianza del término de perturbación

- $y_i - E(y_i) = u_i$

- $E(\underbrace{y_i - E(y_i)}_{u_i}) = E(u_i) = 0$

* FUNCION DE REGRESION MUESTRAL (Estimaciones ó Predicciones)

$$\hat{y}_i = \hat{\beta}_1 + \hat{\beta}_2 x_{2i} + \hat{\beta}_3 x_{3i} + \dots + \hat{\beta}_k x_{ki}$$

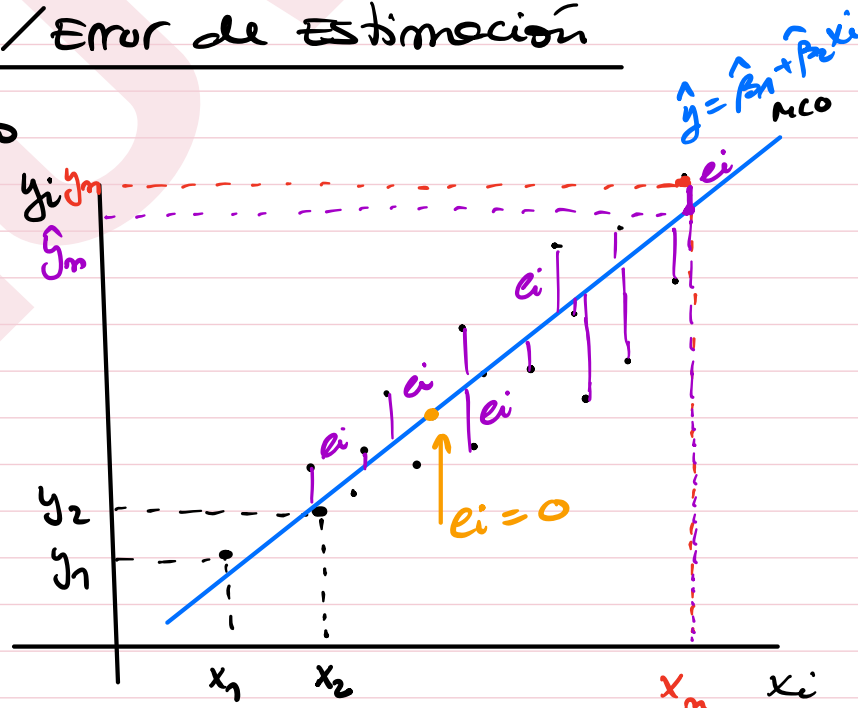
* Error de Predicción / Error de Estimación

Errores del Modelo

RESIDUOS

$$e_i = y_i - \hat{y}_i$$

↓
v.a



* Propiedades de los RESIDUOS

$$E(e_i) = 0 \quad \leftarrow \quad \overset{\text{MCO}}{\sum e_i = 0} \quad \rightarrow \quad \bar{e} = \frac{\sum e_i}{n} = 0$$

$$\text{Var}(e_i) = \sigma_{e_i}^2 \quad \left\{ \begin{array}{l} \text{Var}(e_1) = \sigma_{e_1}^2 \\ \text{Var}(e_2) = \sigma_{e_2}^2 \end{array} \right\} \quad \text{son distintas}$$

$e_i \sim \text{Normal}$

$$e_i \sim N(0, \sigma_{e_i}^2)$$

$$e \sim N(0, \sigma_u^2 M)$$

u_i

$$u_i \sim N(0, \sigma_u^2)$$

$$U \sim N(0, \sigma_u^2 I_n)$$

$\text{Var}(u)$ es
constante

$$e_i = y_i - \hat{y}_i$$


$$e_i \sim N(0, \sigma_{e_i}^2)$$

$$e \sim N(0, \sigma_u^2 M)$$

$\text{Var}(e)$ No
es constante

- $y_i = \beta_1 + \beta_2 x_{2i} + \dots + \beta_k x_{ki} + u_i$

- $\hat{y}_i = \hat{\beta}_1 + \hat{\beta}_2 x_{2i} + \dots + \hat{\beta}_k x_{ki}$

- $y_i = \hat{\beta}_1 + \hat{\beta}_2 x_{2i} + \dots + \hat{\beta}_k x_{ki} + e_i$


- $y_i = \hat{y}_i + e_i$

- $e_i = y_i - \hat{y}_i$