Ejercicio 1

Equipo

2024-03-26

1. Regresión a través del origen.

$$y_i = \beta x_i + \xi_i \quad i:1,\dots,n$$

donde $\xi_1, \ldots, \varepsilon_n$ son v.a.i talque $\xi_i \sim N\left(0, \frac{\sigma^2}{w_i}\right) \ \forall i=1,\ldots,n$ Suponiendo σ^2 conocida y $w_i = \frac{1}{x_i^2}$ $i=1,\ldots,n$ I)

$$\sum_{i=1}^{n} \varepsilon_{i}^{2} = \sum_{i=1}^{n} (y_{i} - \beta x_{i})^{2}$$

$$\frac{d}{d\beta} = 2 \sum_{i=1}^{n} (y_{i} - \beta x_{i}) (-x_{i}) = 0$$

$$\to \sum_{i=1}^{n} (y_{i} - \beta x_{i}) (-x_{i}) = 0 \to \sum_{i=1}^{n} y_{i} (-x_{i}) + \sum_{i=1}^{n} \beta x_{i}^{2} = 0$$

$$\therefore \hat{\beta} = \frac{\sum_{i=1}^{n} y_{i} x_{i}}{\sum_{i=1}^{n} x_{i}^{2}}$$

II) $Var(\hat{\beta})$

$$Var(\hat{\beta}) = Var\left(\frac{\sum Y_i X_i}{\sum X_i^2}\right)$$
$$= Var\left(\frac{\sum X_i (\beta X_i + \xi_i)}{\sum X_i^2}\right)$$

II)

$$\operatorname{Var}(\hat{\beta}) = \operatorname{Var}\left(\frac{\sum X_i^2 \beta + \sum X_i \xi_i}{\sum X_i^2}\right)$$

$$= \operatorname{Var}\left(\beta + \frac{\sum X_i \xi_i}{\sum X_i^2}\right) = \operatorname{Var}\left(\frac{\sum X_i \xi_i}{\sum X_i^2}\right)$$

$$\operatorname{sea} C_i = \frac{X_i}{\sum X_i^2} \longrightarrow \operatorname{Var}(\hat{\beta}) = \sum C_i \xi_i$$

$$\operatorname{Var}(\hat{\beta}) = \left(\sum C_i\right)^2 \operatorname{Var}(\xi_i) = \left(\sum C_i\right)^2 \frac{\sigma^2}{\omega_i^2}$$

$$\therefore \operatorname{Var}(\hat{\beta}) = \frac{\left(\sum X_i\right)^2 \sigma^2}{\left(\sum X_i^2\right)^3}$$

III) Veamos que $\hat{\beta}$ es estimator lineal

$$\hat{B} = \frac{\sum X_i Y_i}{\sum_{i=1}^n X_i^2} \quad \text{Sea} \quad c_i = \frac{X_i}{\sum X_i^2}$$

$$\rightarrow \hat{B} = \sum_{i=1}^n C_i Y_i \quad \therefore \text{ es estimator lineal}$$

Ademas

$$\mathbb{E}(\hat{\beta}) = \mathbb{E}\left(\frac{\sum X_i^2 \beta + \sum X_i \xi_i}{\sum X_i^2}\right)$$
$$= \mathbb{E}(\beta) + \mathbb{E}\left(\frac{\sum X_i \xi_i}{\sum X_i^2}\right) = \beta + \frac{\sum X_i \mathbb{E}\left(\xi_i\right)}{\sum X_i^2}$$

Como $\varepsilon_i \sim N\left(0, \frac{\sigma^2}{w_i}\right)$ entrances $\mathbb{E}(\hat{\beta}) = \beta$ De esta forma como $\hat{\beta}$ es estimador lineal y ademas es insesgado, por el teorema Gauss - Markov $\hat{\beta}$ es el UMVUE