

REGRESIÓN LINEAL

Modelo lineal: $\mathbf{E}[\mathbf{Y}|\mathbf{X} = \mathbf{x}] = \alpha + \beta\mathbf{x}, \quad \mathbf{Y}_i = \alpha + \beta\mathbf{x}_i + \mathbf{E}_i$

Distribución de los estimadores

$$\boxed{\hat{\beta}}$$

$$\hat{\beta} = \frac{\sum x_i y_i - n\bar{xy}}{\sum x_i^2 - n\bar{x}^2}, \quad E[\hat{\beta}] = \beta, \quad Var[\hat{\beta}] = \frac{\sigma^2}{nS_x^2}$$

Si n es grande (o las y_i son normales) y σ conocido

$$\frac{\frac{\hat{\beta} - \beta}{\sigma}}{\frac{S_x}{\sqrt{n}}} \equiv N(0, 1)$$

Si n es pequeño (σ desconocida)

$$\frac{\frac{\hat{\beta} - \beta}{S}}{\frac{S_x}{\sqrt{n}}} \equiv t_{n-2}$$

$$\boxed{\hat{\alpha}}$$

$$\hat{\alpha} = \bar{y} - \hat{\beta}\bar{x}, \quad E[\hat{\alpha}] = \alpha, \quad Var[\hat{\alpha}] = \frac{\sigma^2}{n} \left(1 + \frac{\bar{x}^2}{S_x^2}\right)$$

Si n es grande (o las y_i son normales) y σ conocido

$$\frac{\frac{\hat{\alpha} - \alpha}{\sigma}}{\sqrt{\frac{\sigma^2}{n} \left(1 + \frac{\bar{x}^2}{S_x^2}\right)}} \equiv N(0, 1)$$

Si n es pequeño (σ desconocida)

$$\frac{\frac{\hat{\alpha} - \alpha}{S}}{\sqrt{\frac{S^2}{n} \left(1 + \frac{\bar{x}^2}{S_x^2}\right)}} \equiv t_{n-2}$$

$$\boxed{\hat{\sigma}^2}$$

$$S^2 = \frac{1}{n-2} \sum \left(y_i - (\hat{\alpha} + \hat{\beta}x_i) \right)^2, \quad E[S^2] = \sigma^2$$

$$\frac{(n-2)S^2}{\sigma^2} \equiv \chi_{n-2}^2$$