

# DEMOSTRACIONES

$$\bar{x} = \sum_{j=1}^p f_{.j} \bar{x}_j$$

$$\bar{x} = \sum_{i=1}^N f_{i.} x_i = \sum_{i=1}^N \sum_{j=1}^p f_{ij} x_i = \sum_{j=1}^p \sum_{i=1}^N f_{ij} x_i = \sum_{j=1}^p f_{.j} \bar{x}_j$$

$$\bar{y} = \sum_{i=1}^N f_{i.} \bar{y}_i$$

$$\bar{y} = \sum_{j=1}^p \sum_{i=1}^N f_{ij} y_j = \sum_{j=1}^p \sum_{i=1}^N f_{ij} y_j = \sum_{j=1}^p y_j \sum_{i=1}^N f_{ij} = \sum_{j=1}^p f_{.j} \bar{y}_j$$

$$\sigma_x^2 = \sum_{j=1}^p f_{.j} \sigma_{x_j}^2 + \sum_{j=1}^p f_{.j} (\bar{x}_j - \bar{x})^2$$

$$\sigma_x^2 = \sum_{j=1}^p f_{.j} (x_i - \bar{x})^2 = \sum_{j=1}^p \sum_{i=1}^N f_{ij} (x_i - \bar{x})^2 = \sum_{j=1}^p \sum_{i=1}^N f_{ij} f_{i.}^j (x_i - \bar{x})^2 =$$

$$= \sum_{j=1}^p f_{.j} \left[ \sum_{i=1}^N f_{i.}^j (x_i - \bar{x})^2 \right] = \sum_{j=1}^p f_{.j} \left[ \sum_{i=1}^N f_{i.}^j (x_i - \bar{x}_j + \bar{x}_j - \bar{x})^2 \right] =$$

$$= \sum_{j=1}^p f_{.j} \left[ \sum_{i=1}^N f_{i.}^j (x_i - \bar{x}_j)^2 + \sum_{i=1}^N f_{i.}^j (\bar{x}_j - \bar{x})^2 + 2 \sum_{i=1}^N f_{i.}^j (x_i - \bar{x}_j) (\bar{x}_j - \bar{x}) \right] =$$

$$= \sum_{j=1}^p f_{.j} \left[ \sigma_{x_j}^2 + (\bar{x}_j - \bar{x})^2 \right]$$

■  $X \text{ indep. } Y \iff Y \text{ indep. } X$

$\Rightarrow$

$\exists S: f_i^j = f_{i \cdot} \Rightarrow f_j^i = f_{\cdot j} \quad ?$

$$f_{ij} = f_{i \cdot} f_j^i = f_{\cdot j} f_i^j$$

$$\cancel{f_{i \cdot}} f_j^i = f_{\cdot j} \cancel{f_i^j}$$

$$f_j^i = f_{\cdot j} \Rightarrow Y \text{ indep. } X$$

$\Leftarrow$

$\exists S: f_j^i = f_{\cdot j} \Rightarrow f_i^j = f_{i \cdot} \quad ?$

$$f_{ij} = f_{\cdot j} f_i^j = f_{i \cdot} f_j^i$$

$$\cancel{f_{\cdot j}} f_i^j = f_{i \cdot} \cancel{f_j^i}$$

$$f_i^j = f_{i \cdot} \Rightarrow X \text{ indep. } Y$$

■  $f_{i \cdot j} = f_{i \cdot}$

$$f_{i \cdot} = \sum_{j=1}^p f_{ij} = \sum_{j=1}^p f_{i \cdot j} f_{\cdot j} = f_{i \cdot j} \sum_{j=1}^p f_{\cdot j} = f_{i \cdot j}$$

■  $X \text{ e } Y \text{ indep.} \iff f_{ij} = f_{i \cdot} f_{\cdot j} \iff n_{ij} = \frac{n_{i \cdot} n_{\cdot j}}{n}$

$$f_{i \cdot j} = f_{i \cdot} \iff \frac{f_{i \cdot j} f_{\cdot j}}{f_{\cdot j}} = f_{i \cdot} f_{\cdot j} \iff n f_{ij} = n f_{i \cdot} f_{\cdot j} \iff n_{ij} = n_{i \cdot} \cdot \frac{n_{\cdot j}}{n}$$

■  $m_{ro} = \sum_{i=1}^k f_{i \cdot} x_i^r$

$$m_{ro} = \sum_{i=1}^k \left( \sum_{j=1}^p f_{ij} x_i^r \right) = \sum_{i=1}^k x_i^r \left( \sum_{j=1}^p f_{ij} \right) = \sum_{i=1}^k f_{i \cdot} x_i^r$$

■  $m_{rs} = m_{ro} m_{os} \quad (X \text{ e } Y \text{ indep.})$

$$m_{rs} = \sum_{i=1}^k \sum_{j=1}^p f_{ij} x_i^r y_j^s = \sum_{i=1}^k \sum_{j=1}^p f_{i \cdot} f_{\cdot j} x_i^r y_j^s = \sum_{i=1}^k f_{i \cdot} x_i^r \left( \sum_{j=1}^p f_{\cdot j} y_j^s \right) = m_{ro} m_{os}$$

$$\sigma_{xy} = \mu_{11} = m_{11} - m_{10}m_{01}$$

$$\begin{aligned}\sigma_{xy} &= \sum_{i=1}^k \sum_{j=1}^p f_{ij} (x_i - \bar{x})(y_j - \bar{y}) = \sum_{i=1}^k \sum_{j=1}^p f_{ij} (x_i y_j - \bar{x} y_j - \bar{y} x_i + \bar{x} \bar{y}) = \\ &= \sum_{i=1}^k \sum_{j=1}^p f_{ij} x_i y_j - \sum_{i=1}^k \sum_{j=1}^p f_{ij} \bar{x} y_j - \sum_{i=1}^k \sum_{j=1}^p f_{ij} \bar{y} x_i + \sum_{i=1}^k \sum_{j=1}^p f_{ij} \bar{x} \bar{y} = \\ &= m_{11} - \bar{x} \bar{y} - \bar{x} \bar{y} + \bar{x} \bar{y} = m_{11} - \bar{x} \bar{y} = m_{11} - m_{10} m_{01}\end{aligned}$$



$$\begin{aligned}\frac{\partial \psi(a,b)}{\partial a} &= 0 \\ \frac{\partial \psi(a,b)}{\partial b} &= 0\end{aligned} \Rightarrow \begin{cases} m_{11} = a m_{20} + b m_{10} \\ m_{01} = a m_{10} + b \end{cases} \Rightarrow a = \frac{\sigma_{xy}}{\sigma_x^2}, \quad b = \bar{y} - \frac{\sigma_{xy}}{\sigma_x^2} \bar{x}$$

$$\begin{aligned}\frac{\partial \psi(a,b)}{\partial a} &= 0 \Rightarrow \sum_i \sum_j f_{ij} (y_j - a x_i - b) x_i = 0 = \sum_i \sum_j f_{ij} x_i y_j - a \sum_i \sum_j f_{ij} x_i^2 - b \sum_i \sum_j f_{ij} x_i \\ \frac{\partial \psi(a,b)}{\partial b} &= 0 \Rightarrow \sum_i \sum_j f_{ij} (y_j - a x_i - b) = 0 = \sum_i \sum_j f_{ij} y_j - a \sum_i \sum_j f_{ij} x_i - b \sum_i \sum_j f_{ij}\end{aligned}$$

$$\Rightarrow \begin{cases} \sum_i \sum_j f_{ij} x_i y_j = a \sum_i \sum_j f_{ij} x_i^2 + b \sum_i \sum_j f_{ij} x_i \\ \sum_i \sum_j f_{ij} y_j = a \sum_i \sum_j f_{ij} x_i + b \sum_i \sum_j f_{ij} \end{cases} \Rightarrow \begin{cases} m_{11} = a m_{20} + b m_{10} \\ m_{01} = a m_{10} + b \end{cases}$$

$$\sigma_x^2 = m_{20} - m_{10}^2 \neq 0 \Rightarrow \text{tiene sol y es única}$$

$$a = \frac{m_{11} - m_{10} m_{01}}{m_{20} - m_{10}^2} = \frac{\sigma_{xy}}{\sigma_x^2}$$

$$b = m_{01} - \frac{\sigma_{xy}}{\sigma_x^2} m_{10} = \bar{y} - \frac{\sigma_{xy}}{\sigma_x^2} \bar{x}$$

■ Buscar  $f(x_1), \dots, f(x_k) : \sum_{i=1}^k \sum_{j=1}^p f_{ij} (y_j - f(x_i))^2$  sea mínima

$$\sum_{i=1}^k \sum_{j=1}^p f_{ij} (y_j - f(x_i))^2 = \sum_{i=1}^k \sum_{j=1}^p f_{ij} f_i (y_j - f(x_i))^2 = \sum_{i=1}^k f_i \left[ \sum_{j=1}^p f_{ji} (y_j - f(x_i))^2 \right]$$

■ Minimizar  $\sum_{j=1}^p f_{ji} (y_j - f(x_i))^2$ ,  $\forall i = 1, \dots, k$   
 ↓  
 Media de desviaciones cuadr. de Y respecto de  $f(x_i)$  en las desv. cuadr.  $Y/X=x_i$

$$f(x_i) = \bar{y}_i = \sum_{j=1}^p f_{ji} y_j$$

■  $\sum_{i=1}^k \bar{y}_i f_i = \bar{y}$

$$0 = \sum_{i=1}^k \sum_{j=1}^p f_{ij} (y_j - \bar{y}_i) = \sum_{i=1}^k \sum_{j=1}^p f_{ij} y_j - \sum_{i=1}^k \sum_{j=1}^p f_{ij} \bar{y}_i = m_{01} - \sum_{i=1}^k \bar{y}_i \left( \sum_{j=1}^p f_{ij} \right) = \bar{y} - \sum_{i=1}^k \bar{y}_i f_i$$

■  $\sigma_{ry}^2 = \sum_{i=1}^k \sum_{j=1}^p f_{ij} (r_{ij}^{ry})^2$   $\left[ \sum_{i=1}^k f_i \left( \bar{y} + \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) \right) = \sum_{i=1}^k f_i \bar{y} + \sum_{i=1}^k f_i \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) = \bar{y} \right]$

$$(r_{ij}^{ry} = y_j - \bar{y}_i, \quad k=1, \dots, k, \quad \forall j=1, \dots, p)$$

$$\begin{aligned} \sigma_{ry}^2 &= \sum_{i=1}^k \sum_{j=1}^p f_{ij} (r_{ij}^{ry})^2 \stackrel{\text{error cuadr. medio}}{=} \sum_{i=1}^k \sum_{j=1}^p f_{ij} (y_j - \bar{y}_i)^2 = \sum_{i=1}^k \sum_{j=1}^p f_{ij} f_i (y_j - \bar{y}_i)^2 = \\ &= \sum_{i=1}^k f_i \underbrace{\sum_{j=1}^p f_{ji} (y_j - \bar{y}_i)^2}_{\text{Varianza condic}} = \sum_{i=1}^k f_i \underbrace{\sigma_{y_i}^2}_{\text{Media de las varianzas condic}} \end{aligned}$$

■ Los resid. lin. mín. cuad. tienen media 0 y su varianza es el error cuadr. medio

$$\begin{aligned} \sum_{i=1}^k \sum_{j=1}^p f_{ij} r_{ij}^{ry} &= \sum_{i=1}^k \sum_{j=1}^p f_{ij} \left( y_j - \bar{y} + \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) \right) = \sum_{i=1}^k \sum_{j=1}^p f_{ij} y_j - \sum_{i=1}^k \sum_{j=1}^p f_{ij} \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) - \bar{y} \sum_{i=1}^k \sum_{j=1}^p f_{ij} = \\ &= \bar{y} - \bar{y} - \frac{\sigma_{xy}}{\sigma_x^2} \sum_{i=1}^k \sum_{j=1}^p f_{ij} \overset{0 \text{ (media)}}{\underset{0}{(x_i - \bar{x})}} = 0 \end{aligned}$$

$$\Rightarrow \sigma_{ry}^2 = \sum_{i=1}^k \sum_{j=1}^p f_{ij} (r_{ij}^{ry} - 0)^2 = \sum_{i=1}^k \sum_{j=1}^p f_{ij} (r_{ij}^{ry})^2 = ECM$$

↑  
varianza de resid.

↑  
xg la media de los resid. es 0

Varianza valores observ. = varianza val. ajustados + varianza resid.

$$\sigma_y^2 = \sigma_{ey}^2 + \sigma_{ry}^2$$

$$\begin{aligned} \bullet \sigma_{ey}^2 &= \sum_{i=1}^n f_i \left( \bar{y} + \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) - \bar{y} \right)^2 = \sum_{i=1}^n f_i \frac{\sigma_{xy}^2}{(\sigma_x^2)^2} (x_i - \bar{x})^2 = \\ &= \frac{\sigma_{xy}^2}{(\sigma_x^2)^2} \sum_{i=1}^n f_i (x_i - \bar{x})^2 = \frac{\sigma_{xy}^2}{(\sigma_x^2)^2} \sigma_x^2 = \frac{\sigma_{xy}^2}{\sigma_x^2} \Rightarrow \sigma_{ey}^2 = \frac{\sigma_{xy}^2}{\sigma_x^2} \end{aligned}$$

$$\begin{aligned} \bullet \sigma_{ry}^2 &= \sum_{i=1}^n \sum_{j=1}^p f_{ij} \left( y_{ij} - \bar{y} - \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) \right)^2 = \\ &= \sum_{i=1}^n \sum_{j=1}^p f_{ij} \left[ (y_{ij} - \bar{y})^2 + \left( \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) \right)^2 - 2 (y_{ij} - \bar{y}) \left( \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) \right) \right] = \\ &= \sum_{i=1}^n \sum_{j=1}^p f_{ij} (y_{ij} - \bar{y})^2 + \sum_{i=1}^n \sum_{j=1}^p f_{ij} \left( \frac{\sigma_{xy}^2}{\sigma_x^2} \right) (x_i - \bar{x})^2 - 2 \sum_{i=1}^n \sum_{j=1}^p f_{ij} (y_{ij} - \bar{y}) \left( \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) \right) = \\ &= \sigma_y^2 + \frac{\sigma_{xy}^2}{(\sigma_x^2)^2} \sigma_x^2 - 2 \frac{\sigma_{xy}}{\sigma_x^2} \sigma_{xy} = \sigma_y^2 + \frac{\sigma_{xy}^2}{\sigma_x^2} - 2 \frac{\sigma_{xy}^2}{\sigma_x^2} = \sigma_y^2 - \frac{\sigma_{xy}^2}{\sigma_x^2} \Rightarrow \end{aligned}$$

$$\Rightarrow \sigma_{xy}^2 = \sigma_y^2 - \frac{\sigma_{xy}^2}{\sigma_x^2}$$

$X, Y$  indep  $\Rightarrow \sigma_{xy} = 0$

$$\sigma_{xy} = \sum \sum f_{ij} (x_i - \bar{x}) (y_{ij} - \bar{y}) = \sum f_i (x_i - \bar{x}) \left[ \sum f_{ij} (y_{ij} - \bar{y}) \right] = 0$$

$\sigma_y^2 = \sigma_{re}^2 + \sigma_{expl}^2$  (Caso lineal)

$$\begin{aligned} \sigma_y^2 &= \sum_{j=1}^p f_{.j} (y_{.j} - \bar{y})^2 = \sum_{j=1}^p f_{.j} \frac{(y_{.j} - a\bar{x}_j - b + a\bar{x}_j + b - \bar{y})^2}{\bar{y}_j - \bar{y}} = \\ &= \sum_{j=1}^p f_{.j} (y_{.j} - \bar{y}_j)^2 + \sum_{j=1}^p f_{.j} (\bar{y}_j - \bar{y})^2 + 2 \sum_{j=1}^p f_{.j} (y_{.j} - \bar{y}_j) (\bar{y}_j - \bar{y}) = \\ &= \sum_{j=1}^p f_{.j} (r_{.j} - 0)^2 + \sigma_{expl}^2 = \sigma_{re}^2 + \sigma_{expl}^2 \end{aligned}$$

$$r^2 = \frac{\sigma_{xy}^2}{\sigma_x^2 \sigma_y^2}$$

$$\begin{aligned} \sigma_{res}^2 &= \frac{1}{n} \sum \sum n_{ij} (y_j - ax_i - b)^2 = \sum \sum f_{ij} \left( y_j - \frac{\sigma_{xy}}{\sigma_x^2} x_i - \bar{y} + \frac{\sigma_{xy}}{\sigma_x^2} \bar{x} \right)^2 = \\ &= \sum \sum f_{ij} \left[ (y_j - \bar{y}) - \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) \right]^2 = \\ &= \sum \sum f_{ij} (y_j - \bar{y})^2 + \sum \sum f_{ij} \frac{\sigma_{xy}^2}{(\sigma_x^2)^2} (x_i - \bar{x})^2 - 2 \sum \sum f_{ij} \frac{\sigma_{xy}}{\sigma_x^2} (x_i - \bar{x}) (y_j - \bar{y}) = \\ &= \sigma_y^2 + \frac{\sigma_{xy}^2}{(\sigma_x^2)^2} \sigma_x^2 - 2 \frac{\sigma_{xy}^2}{\sigma_x^2} = \left( \sigma_y^2 - \frac{\sigma_{xy}^2}{\sigma_x^2} \right) \end{aligned}$$

$$\begin{aligned} \eta_{y/x}^2 &= 1 - \frac{\sigma_{res}^2}{\sigma_y^2} = 1 - \frac{\sigma_y^2 + \frac{\sigma_{xy}^2}{(\sigma_x^2)^2} \sigma_x^2 - 2 \frac{\sigma_{xy}^2}{\sigma_x^2}}{\sigma_y^2} = \frac{\sigma_{xy}^2}{\sigma_x^2 \sigma_y^2} = r^2 \\ \eta_{x/y}^2 & \end{aligned}$$