



Instituto Tecnológico
de Buenos Aires

23/OCTUBRE

MODELO LINEAL

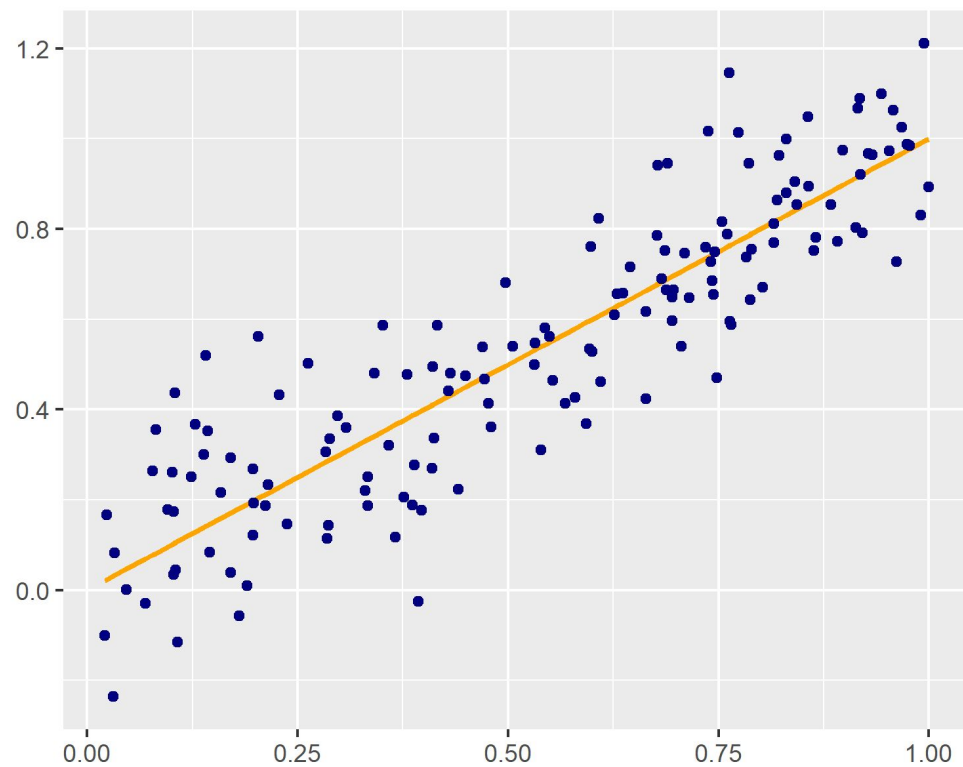
EXTENSIONES

—

El modelo lineal

$$y = f(x) + \varepsilon$$

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_p x_p + \varepsilon$$



$$\frac{dy}{dx_j} = \beta_j$$

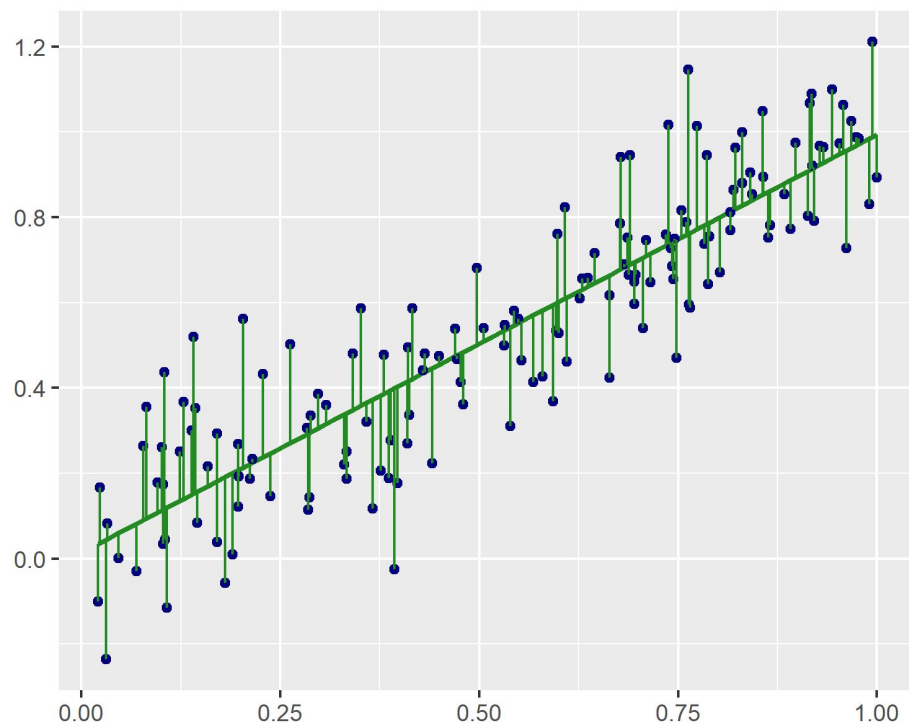
¿Cómo se
interpreta β_j ?



Estimación por MCO

$$\min SCR = \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_{1i} + \hat{\beta}_2 x_{2i} + \dots + \hat{\beta}_p x_{pi}$$

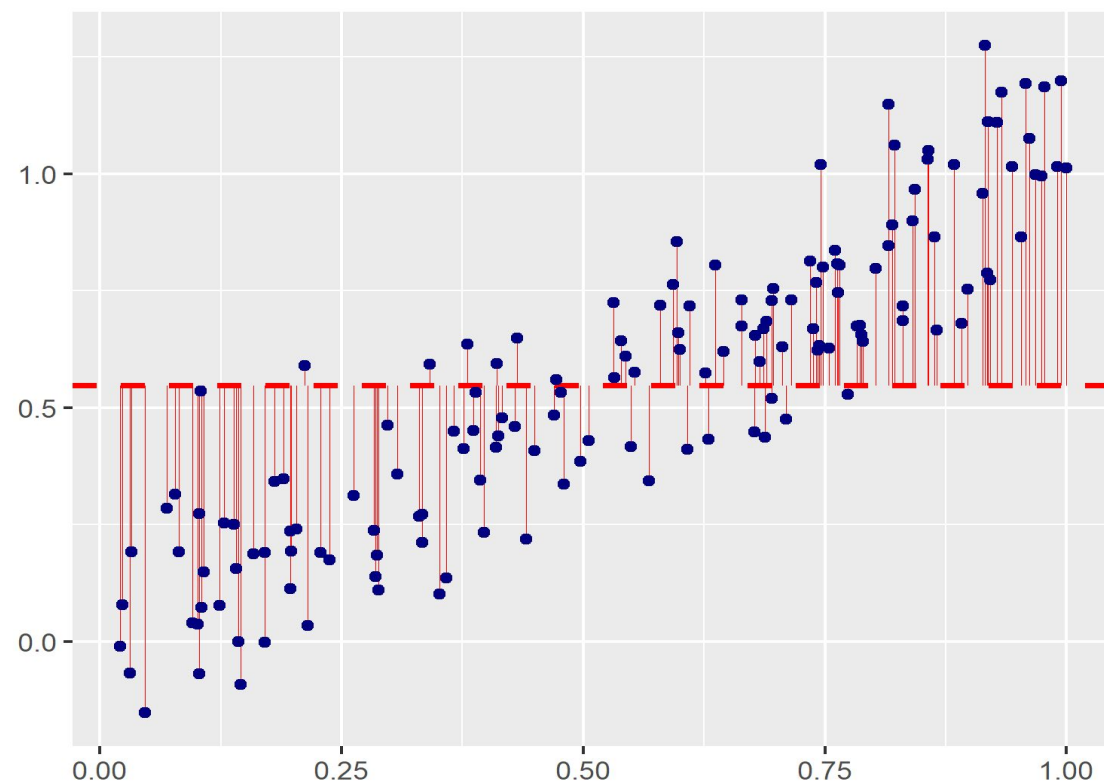
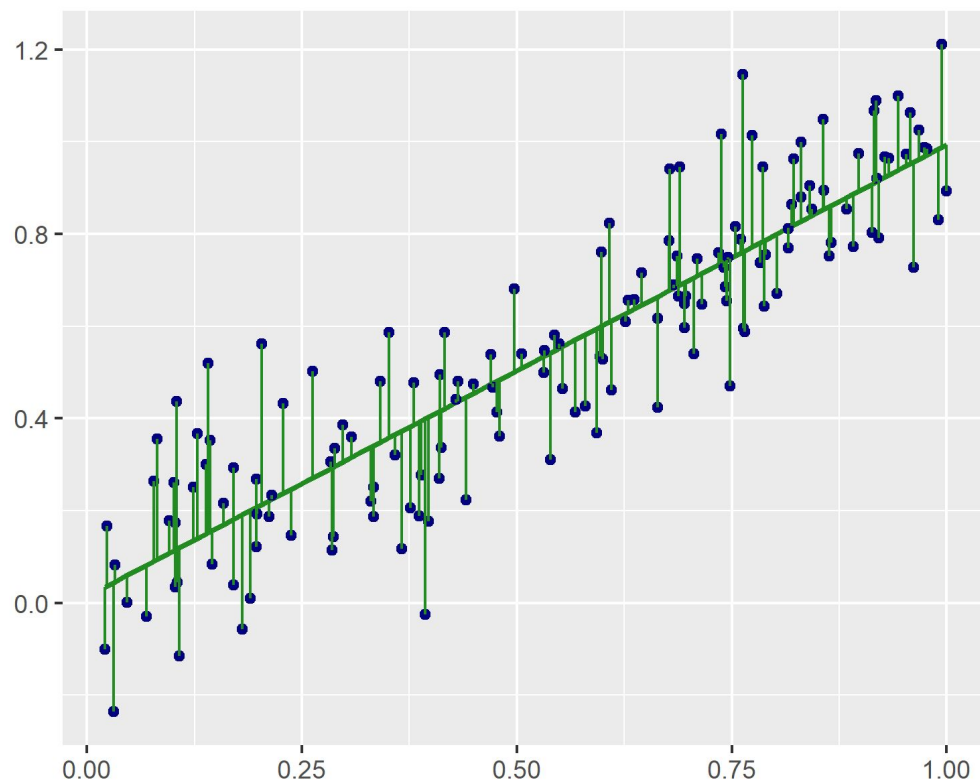


Bondad de ajuste

SCR

$$R^2(y, \hat{y}) = 1 - \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y})^2}$$

SCT



Atributos categóricos

$$educ = \begin{cases} \text{primario} \\ \text{secundario} \\ \text{universitario} \end{cases}$$

$$educ_U = \begin{cases} 1 & \text{univ.} \\ 0 & \text{NO univ.} \end{cases}$$

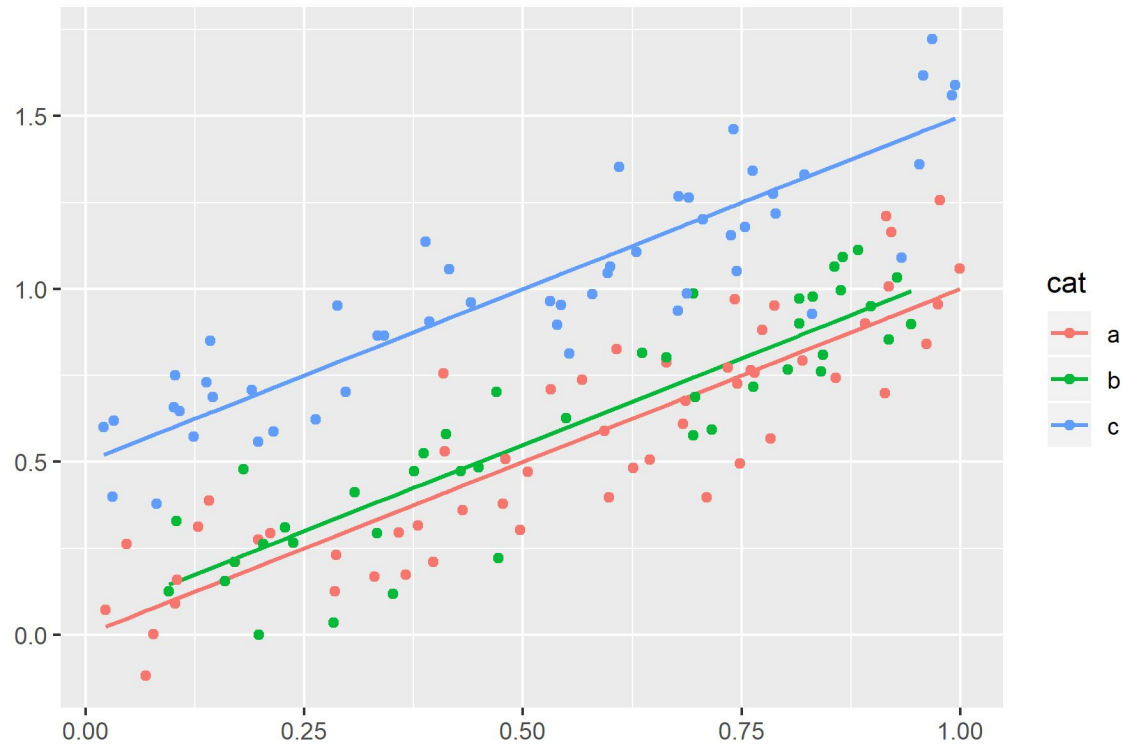
$$educ_S = \begin{cases} 1 & \text{sec.} \\ 0 & \text{NO sec.} \end{cases}$$

dummies / flags

$$y = \beta_0 + \beta_1 exp + \beta_2 educ_U + \beta_3 educ_S + \epsilon$$

$$y = \begin{cases} (\beta_0 + \beta_2) + \beta_1 exp + \epsilon & \text{univ.} \\ (\beta_0 + \beta_3) + \beta_1 exp + \epsilon & \text{sec.} \\ \beta_0 + \beta_1 exp + \epsilon & \text{prim.} \end{cases}$$

Atributos categóricos



$$y = \beta_0 + \beta_1 exp + \beta_2 educ_U + \beta_3 educ_S + \epsilon$$

$$educ_U = \begin{cases} 1 & \text{univ.} \\ 0 & \text{NO univ.} \end{cases}$$

$$educ_S = \begin{cases} 1 & \text{sec.} \\ 0 & \text{NO sec.} \end{cases}$$

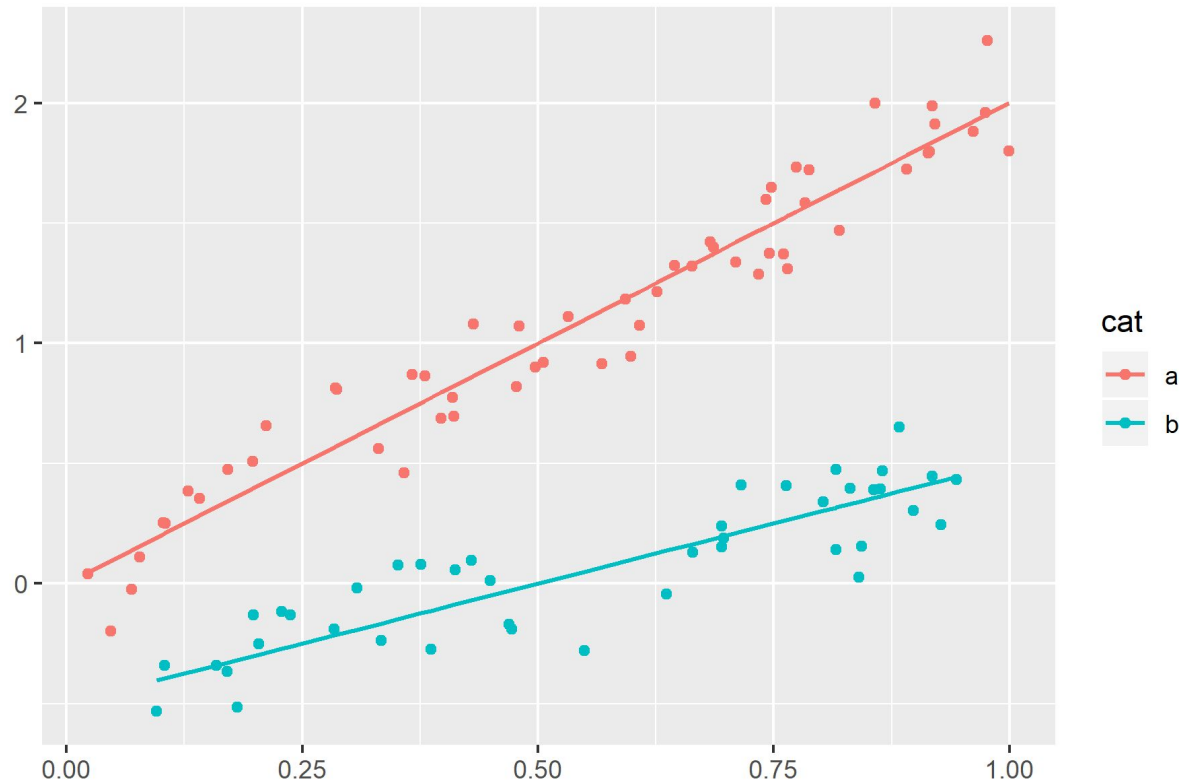
Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.007509	0.030774	0.244	0.808
x	0.993354	0.042606	23.315	<2e-16 ***
catb	0.045692	0.030489	1.499	0.136
catc	0.483520	0.028832	16.770	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Interacciones

$$y = \beta_0 + \beta_1 exp + \beta_2 inf + \beta_3 exp \cdot inf + \epsilon$$



$$informal = \begin{cases} 1 & \text{informal} \\ 0 & \text{caso contrario} \end{cases}$$

$$\frac{dy}{dexp} = \begin{cases} \beta_1 + \beta_3 & \text{informal} \\ \beta_1 & \text{caso contrario} \end{cases}$$

Coefficients:

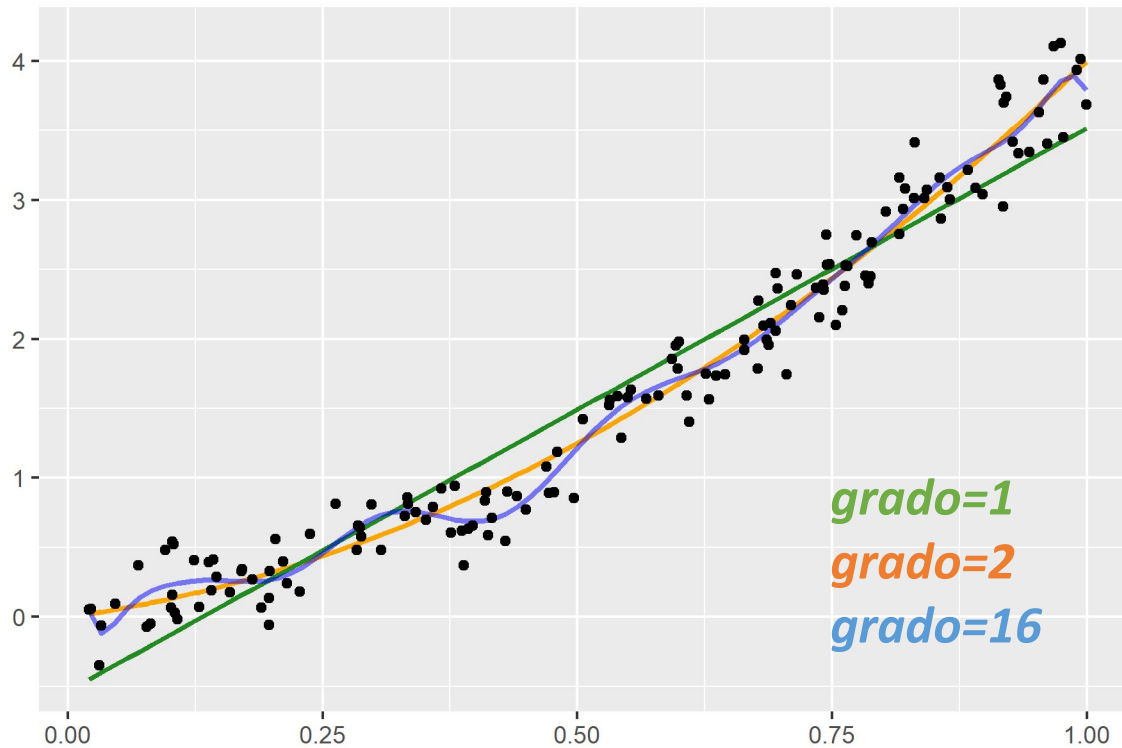
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.004983	0.043494	0.115	0.909
x	1.991350	0.070244	28.349	< 2e-16 ***
catb	-0.466229	0.067880	-6.868	6.95e-10 ***
x:catb	-1.052446	0.109473	-9.614	1.22e-15 ***

No linealidades

$$y = \beta_0 + \beta_1 x + \beta_2 x^2 + \varepsilon$$

$$\frac{dy}{dx} = \beta_1 + 2\beta_2 x$$

Regresión
polinómica



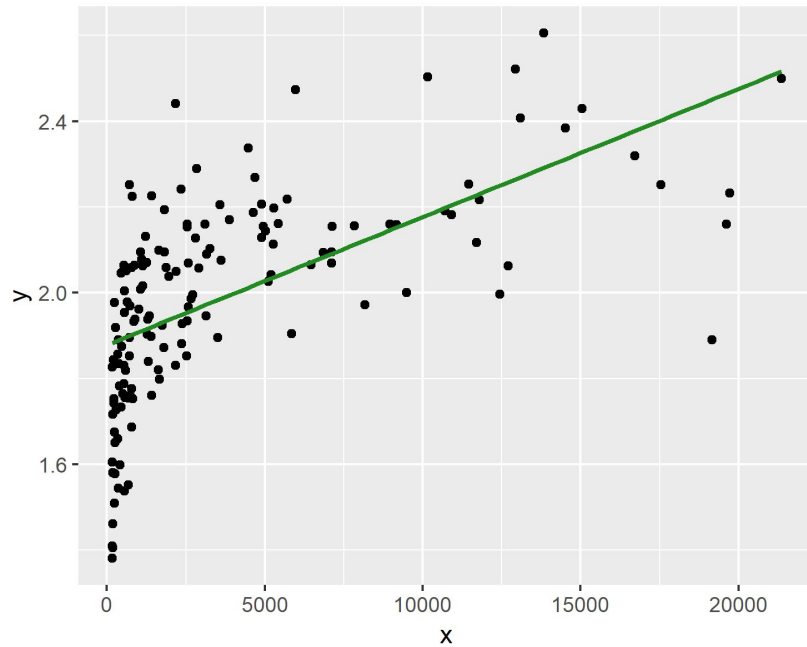
Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.04795	0.05441	0.881	0.37952
poly(x, 2, raw = T)1	0.76903	0.24570	3.130	0.00211 **
poly(x, 2, raw = T)2	3.18956	0.23365	13.651	< 2e-16 ***

Signif. codes:	0 '***'	0.001 '**'	0.01 '*'	0.05 '.' 0.1 ' ' 1

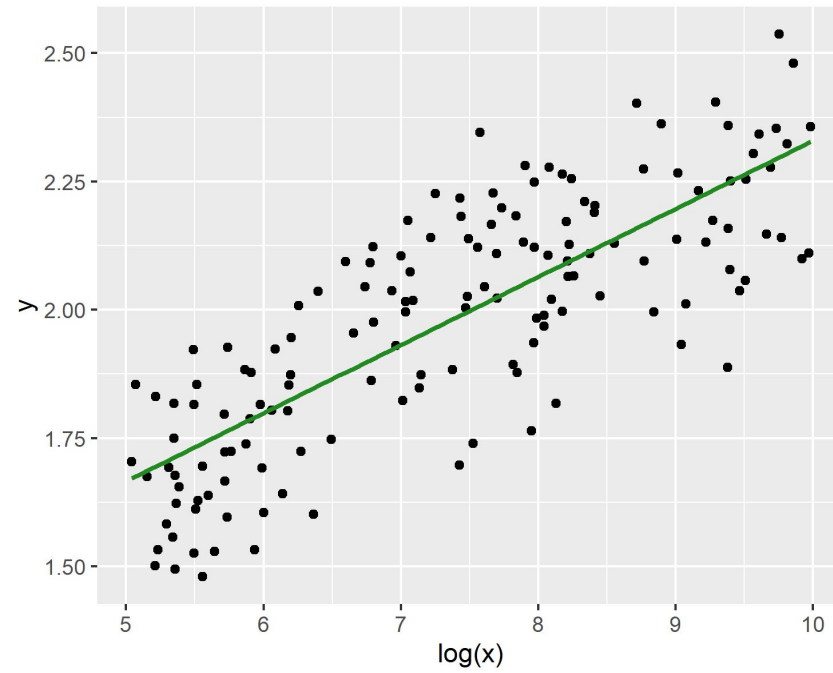
No linealidades

$$y = \beta_0 + \beta_1 x + \epsilon$$



lin-log

$$y = \beta_0 + \beta_1 \ln x + \epsilon$$

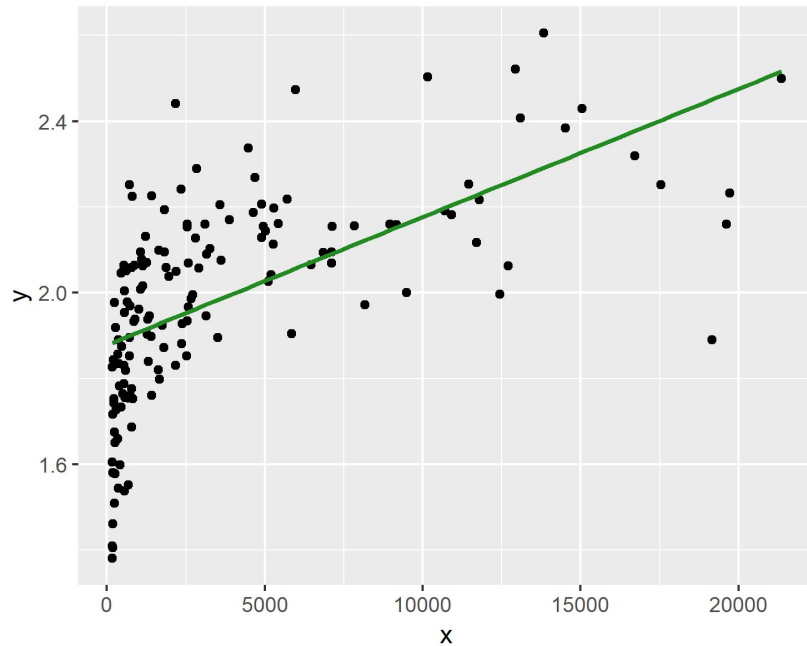


$$\beta_1 = \frac{dy}{dx/x}$$

$$100\beta_1 = \frac{\Delta y}{\Delta \%x}$$

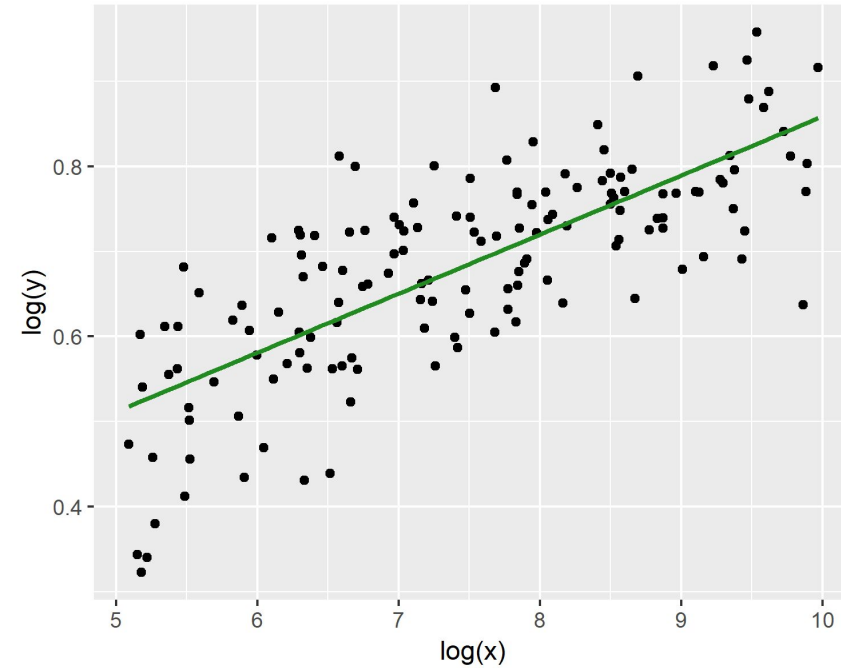
No linealidades

$$y = \beta_0 + \beta_1 x + \varepsilon$$



log-log

$$\ln y = \beta_0 + \beta_1 \ln x + \varepsilon$$



$$\beta_1 = \frac{dy/y}{dx/x}$$

$$\beta_1 = \frac{\Delta\%y}{\Delta\%x}$$

OJO! acá estamos prediciendo otra cosa

No linealidades

