

MAT-042: Taller 4

Felipe Osorio

<http://fosorios.mat.utfsm.cl>

Departamento de Matemática, UTFSM



Ejercicio 1.a)

Para el conjunto de datos $x = \{x_1, x_2, \dots, x_{10}\}$. Tenemos,

$$\sum_{i=1}^{10} x_i = 140, \quad \sum_{i=1}^{10} x_i^2 = 5230.$$

De este modo, $\bar{x} = 140/10 = 14$. Mientras que,

$$\begin{aligned} s^2 &= \frac{1}{10-1} \sum_{i=1}^{10} (x_i - \bar{x})^2 = \frac{1}{9} \left(\sum_{i=1}^{10} x_i^2 - 10 \cdot \bar{x}^2 \right) = \frac{1}{9} (5230 - 10 \cdot 14^2) \\ &= \frac{1}{9} (5230 - 1960) = \frac{3270}{9} = \frac{1090}{3} = 363.3333. \end{aligned}$$

Además,

$$CV = \frac{s}{\bar{x}} = \frac{\sqrt{1090/3}}{14} = \frac{19.0613}{14} = 1.3615.$$



Ejercicio 1.b)

Tenemos que los valores ordenados, $x_{(1)} < x_{(2)} < \dots < x_{(10)}$ son dados por:

$$2, 3, 5, 7, 8, 10, 11, 12, 15, 67.$$

Como $n = 10$, sigue que

$$me = \frac{8 + 10}{2} = 9.$$

Para calcular Q_1 y Q_3 considere los nuevos conjuntos de datos ordenados

$$D_1 = \{2, 3, 5, 7, 8\}, \quad \text{y} \quad D_2 = \{10, 11, 12, 15, 67\}.$$

De ahí que $Q_1 = 5$ y $Q_3 = 12$, y $IQR = Q_3 - Q_1 = 12 - 5 = 7$. Esto permite obtener

$$b_G = \frac{(Q_3 - me) - (me - Q_1)}{IQR} = \frac{(12 - 9) - (9 - 5)}{7} = \frac{3 - 4}{7} = -\frac{1}{7} = -0.1429.$$



Ejercicio 1.c)

Sabemos que

$$\begin{aligned}\bar{y} &= -1.3\bar{x} + 7 = -1.3 \cdot 14 + 7 = -11.2 \\ \text{var}(\mathbf{y}) &= (-1.3)^2 \text{var}(\mathbf{x}) = 1.69 \cdot 363.3333 = 614.0333.\end{aligned}$$

De este modo,

$$\text{CV}_y = \frac{\sqrt{\text{var}(\mathbf{y})}}{|\bar{y}|} = \frac{\sqrt{614.0333}}{11.2} = \frac{24.7797}{11.2} = 2.2125.$$



Ejercicio 1.d)

Note que

$$y_i = ax_i + b, \quad i = 1, 2, \dots, n,$$

con $a = -1.3$ y $b = 7$. Esto nos permite escribir

$$\begin{aligned} \text{cov}(\mathbf{x}, \mathbf{y}) &= \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(ax_i + b - a\bar{x} - b) \\ &= a \cdot \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = a \text{var}(\mathbf{x}). \end{aligned}$$

Como $\text{var}(\mathbf{y}) = a^2 \text{var}(\mathbf{x})$, sigue que

$$\text{corr}(\mathbf{x}, \mathbf{y}) = \frac{\text{cov}(\mathbf{x}, \mathbf{y})}{\sqrt{\text{var}(\mathbf{x}) \text{var}(\mathbf{y})}} = \frac{a \text{var}(\mathbf{x})}{\sqrt{a^2 \text{var}^2(\mathbf{x})}},$$

y como en nuestro caso particular $a < 0$, sigue que

$$\text{corr}(\mathbf{x}, \mathbf{y}) = \frac{a}{\sqrt{a^2}} = \frac{a}{|a|} = -1.$$



Ejercicio 2)

Desarrollando el cuadrado de binomio y sumando, obtenemos

$$\begin{aligned}\sum_{i=1}^k n_i (x_i - \bar{x})^2 &= \sum_{i=1}^k n_i (x_i^2 - 2x_i\bar{x} + \bar{x}^2) = \sum_{i=1}^k n_i x_i^2 - 2\bar{x} \sum_{i=1}^k n_i x_i + \bar{x}^2 \sum_{i=1}^k n_i \\ &= \sum_{i=1}^k n_i x_i^2 - 2n\bar{x}^2 + n\bar{x}^2 = \sum_{i=1}^k n_i x_i^2 - n\bar{x}^2,\end{aligned}$$

lo que verifica el resultado.



Ejercicio 3)

Tenemos $n = 6$, y

$$\begin{aligned}\sum_{i=1}^n x_i &= 21, & \sum_{i=1}^n x_i^2 &= 91, & \sum_{i=1}^n x_i y_i &= 280 \\ \sum_{i=1}^n y_i &= 65, & \sum_{i=1}^n y_i^2 &= 879.\end{aligned}$$

Es decir,

$$\sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - n\bar{x}^2 = 91 - 21^2/6 = 17.5000$$

$$\sum_{i=1}^n (y_i - \bar{y})^2 = \sum_{i=1}^n y_i^2 - n\bar{y}^2 = 879 - 65^2/6 = 174.8333$$

$$\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \sum_{i=1}^n x_i y_i - n\bar{x}\bar{y} = 280 - 21 \cdot 65/6 = 52.5000.$$



Ejercicio 3)

De este modo,

$$\hat{\beta} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{52.5}{17.5} = 3.0,$$

y portanto

$$\hat{\alpha} = \bar{y} - \hat{\beta}\bar{x} = \frac{1}{6}(65 - 3 \cdot 21) = \frac{1}{3}.$$

Además, tenemos que los residuos, $e_i = y_i - \hat{\alpha} - \hat{\beta}x_i$, para $i = 1, \dots, n$, son dados por:

$$\begin{aligned} e &= \left\{ 5 - \frac{1}{3} - 3, 7 - \frac{1}{3} - 6, 7 - \frac{1}{3} - 9, 10 - \frac{1}{3} - 12, 16 - \frac{1}{3} - 15, 20 - \frac{1}{3} - 18 \right\} \\ &= \left\{ 2 - \frac{1}{3}, 1 - \frac{1}{3}, -2 - \frac{1}{3}, -2 - \frac{1}{3}, 1 - \frac{1}{3}, 2 - \frac{1}{3} \right\} \\ &= \left\{ (6 - 1)/3, (3 - 1)/3, (-6 - 1)/3, (-6 - 1)/3, (3 - 1)/3, (6 - 1)/3 \right\} \\ &= \left\{ 5/3, 2/3, -7/3, -7/3, 2/3, 5/3 \right\}, \end{aligned}$$



Ejercicio 3)

Tenemos que,

$$RSS = \sum_{i=1}^n e_i^2 = \frac{1}{3^2} (5^2 + 2^2 + (-7)^2 + (-7)^2 + 2^2 + 5^2) = \frac{156}{9} = 17.3333.$$

Es decir, $s^2 = RSS/(n - 2) = 17.3333/4 = 4.3333$. Finalmente

$$R^2 = 1 - \frac{RSS}{\sum_{i=1}^n (y_i - \bar{y})^2} = 1 - \frac{17.3333}{174.3333} = 0.9009.$$



Ejercicio 4.a)

Tenemos que

$$\begin{aligned}\frac{d}{d\theta} S_1(\theta) &= \sum_{i=1}^n \frac{d}{d\theta} (y_i - \theta x_i)^2 = 2 \sum_{i=1}^n (y_i - \theta x_i) \frac{d}{d\theta} (y_i - \theta x_i) \\ &= -2 \sum_{i=1}^n (y_i - \theta x_i) x_i.\end{aligned}$$

Resolviendo la condición $d S_1(\theta) / d\theta = 0$, sigue que

$$\sum_{i=1}^n (y_i - \theta x_i) x_i = 0 \quad \implies \quad \hat{\theta} = \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n x_i^2}.$$

Notando que

$$\frac{d^2}{d\theta^2} S_1(\theta) = 2 \sum_{i=1}^n x_i^2 > 0,$$

para cualquier $\theta \in \mathbb{R}$, sigue que $\hat{\theta}$ es mínimo global.



Ejercicio 4.b)

Para los datos del Ejercicio 3, tenemos

$$\hat{\theta} = \frac{280}{91} = 3.0769.$$

Calculando $e_i = y_i - \hat{\theta}x_i$, para $i = 1, \dots, 6$, resulta

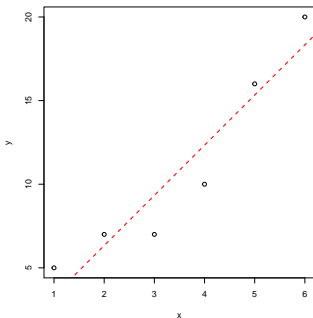
$$e = \{1.9231, 0.8462, -2.2308, -2.3077, 0.6154, 1.5385\}.$$

De este modo,

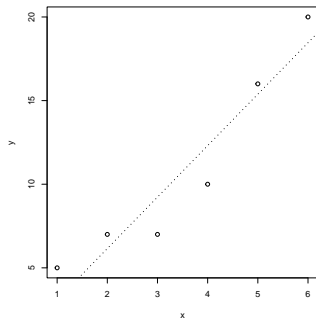
$$s_*^2 = \frac{1}{6-1} \sum_{i=1}^6 e_i^2 = \frac{17.4620}{5} = 3.4924.$$



Ejercicio 3 y 4)¹



(a) con intercepto



(b) sin intercepto

¹Además, $R_0^2 = \{\text{corr}(\mathbf{y}, \hat{\mathbf{y}})\}^2 = 0.9009$.

Ejercicio 3 y 4)

```
# ajuste del modelo 'con' intercepto
> fm <- lm(y ~ x, data = p3)
> fm
```

```
Call:
lm(formula = y ~ x, data = p3)
```

```
Coefficients:
(Intercept)          x
  0.3333         3.0000
```

```
# ajuste del modelo 'sin' intercepto
> f0 <- lm(y ~ -1 + x, data = p3)
> f0
```

```
Call:
lm(formula = y ~ -1 + x, data = p3)
```

```
Coefficients:
          x
  3.077
```



Ejercicio 3 y 4)

```
# resumen de estimación del modelo 'con' intercepto
```

```
> summary(fm)
```

```
Call:
```

```
lm(formula = y ~ x, data = p3)
```

```
Residuals:
```

1	2	3	4	5	6
1.6667	0.6667	-2.3333	-2.3333	0.6667	1.6667

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.3333	1.9379	0.172	0.87179
x	3.0000	0.4976	6.029	0.00382

```
Residual standard error: 2.082 on 4 degrees of freedom
```

```
Multiple R-squared: 0.9009, Adjusted R-squared: 0.8761
```

```
F-statistic: 36.35 on 1 and 4 DF, p-value: 0.003815
```



Ejercicio 3 y 4)

```
# resumen de estimación del modelo 'sin' intercepto  
> summary(f0)
```

```
Call:
```

```
lm(formula = y ~ -1 + x, data = p3)
```

```
Residuals:
```

1	2	3	4	5	6
1.9231	0.8462	-2.2308	-2.3077	0.6154	1.5385

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
x	3.0769	0.1959	15.71	1.9e-05

```
Residual standard error: 1.869 on 5 degrees of freedom
```

```
Multiple R-squared: 0.9801, Adjusted R-squared: 0.9762
```

```
F-statistic: 246.7 on 1 and 5 DF, p-value: 1.902e-05
```

