

13.01.25

Part 1: SLR

Simple Linear Regression: $y = \alpha + \beta x + \epsilon$

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

$$\hat{\alpha} = \bar{y} - \hat{\beta} \bar{x}$$

Step 1: Formulating the OLS problem

$$S(\alpha, \beta) = \sum_{i=1}^n (y_i - \alpha - \beta x_i)^2$$

Step 2: Deriving the Estimator for α

$$0: \frac{\partial S}{\partial \alpha} = \sum_{i=1}^n 2(y_i - \alpha - \beta x_i)(-1) = -2 \sum_{i=1}^n (y_i - \alpha - \beta x_i)$$

$$0: \sum_{i=1}^n y_i - n\alpha - \beta \sum_{i=1}^n x_i = 0$$

$$0: n\alpha = \sum_{i=1}^n y_i - \beta \sum_{i=1}^n x_i$$

$$0: \alpha = \bar{y} - \beta \bar{x}$$

where $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$ and $\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$ are the

simple means.

Step 3: Deriving the Estimator for β

$$\frac{\partial S}{\partial \beta} = \sum_{i=1}^n 2(y_i - \alpha - \beta x_i)(-x_i) = -2 \sum_{i=1}^n x_i (y_i - \alpha - \beta x_i)$$

Setting this to zero:

$$\sum_{i=1}^n x_i y_i - a \sum_{i=1}^n x_i - \beta \sum_{i=1}^n x_i^2 = 0$$

Substituting $a = \bar{y} - \beta \bar{x}$:

$$\sum_{i=1}^n x_i y_i - (\bar{y} - \beta \bar{x}) \sum_{i=1}^n x_i - \beta \sum_{i=1}^n x_i^2 = 0$$

$$\sum_{i=1}^n x_i y_i - \bar{y} \sum_{i=1}^n x_i + \beta \bar{x} \sum_{i=1}^n x_i - \beta \sum_{i=1}^n x_i^2 = 0$$

Rearrange for β :

$$\beta \left(\sum_{i=1}^n x_i^2 - \bar{x} \sum_{i=1}^n x_i \right) = \sum_{i=1}^n x_i y_i - \bar{y} \sum_{i=1}^n x_i$$

Using the identities:

$$\sum_{i=1}^n x_i^2 - n \bar{x}^2 = \sum_{i=1}^n (x_i - \bar{x})^2$$

$$\sum_{i=1}^n x_i y_i - n \bar{x} \bar{y} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

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

and

$$\hat{a} = \bar{y} - \hat{\beta} \bar{x}$$

Part 2: MLR

Multiple Linear Regression: $y = \beta x + \epsilon$

1. The $p \times 1$ vector of regression coefficients
2. The $n \times p$ design matrix containing independent variables

3. x is $n \times p$

β is $p \times 1$

4. Proof of $\hat{\beta} = (x^T x)^{-1} x^T y$:

Solve OLS: $x^T y = x^T x \beta$

If $x^T x$ is invertible: $\hat{\beta} = (x^T x)^{-1} x^T y$

5. Dimensions of each element in $\hat{\beta}$:

- $\hat{\beta}$ is $p \times 1$

- $x^T x$ is $p \times p$

- $x^T y$ is $p \times 1$

6. Conditions for $x^T x$ to be invertible:

- x must be have full column rank n
- $n \geq p$ (more observations than predictors)
- No linear dependence among columns of x

Part 3: Calculate simple Linear Regression coefficients

x	y
1	3
2	5
3	4
4	7
5	6

1. Calculate the Intercept ($\hat{\alpha}$) and Slope ($\hat{\beta}$)

$$\hat{\beta} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$

1) Compute means

$$\bar{x} = \frac{1 + 2 + 3 + 4 + 5}{5} = 3$$

$$\bar{y} = \frac{3 + 5 + 4 + 7 + 6}{5} = 5$$

2) Compute $\hat{\beta}$

$$\sum (x_i - \bar{x})(y_i - \bar{y}) = (1-3)(3-5) + (2-3)(5-5) +$$

$$(3-3)(4-5) + (4-3)(7-5) + (5-3)(6-5) =$$

$$= (-2) \cdot (-2) + (-1) \cdot 0 + (0) \cdot (-1) + (1) \cdot 2 + 2 \cdot 1 = 8$$

$$\sum (x_i - \bar{x})^2 = (-2)^2 + (-1)^2 + (0)^2 + (1)^2 + 2^2 =$$

$$= 4 + 1 + 0 + 1 + 4 = 10$$

$$\hat{\beta} = \frac{8}{10} = 0,8$$

3) Compute \hat{e}^1

$$\hat{e}^1 = 5 - (0,8 \cdot 3) = 5 - 2,4 = 2,6$$

$$\hat{y}^1 = 2,6 + 0,8x$$

3 Calculate Residuals:

$$\text{Residual} = y_i - \hat{y}_i$$

x	y	\hat{y}	Residual ($y - \hat{y}$)
1	3	$2,6 + 0,8 \cdot 1 = 3,4$	$3 - 3,4 = -0,4$
2	5	$2,6 + 0,8 \cdot 2 = 4,2$	$5 - 4,2 = 0,8$
3	4	$2,6 + 0,8 \cdot 3 = 5$	$4 - 5 = -1$
4	4	$2,6 + 0,8 \cdot 4 = 5,8$	$4 - 5,8 = -1,8$
5	6	$2,6 + 0,8 \cdot 5 = 6,6$	$6 - 6,6 = -0,6$

4. Calculate SSR

$$\begin{aligned} SSR &= \sum (y_i - \hat{y}_i)^2 = (-0,4)^2 + (0,8)^2 + (-1,0)^2 + (-1,2)^2 + (-0,6)^2 \\ &= 0,16 + 0,64 + 1 + 1,44 + 0,36 = 3,6 \end{aligned}$$

5. Calculate R^2 .

$$R^2 = 1 - \frac{SSR}{SST}$$

$$\begin{aligned} SST &= \sum (y_i - \bar{y})^2 = (3-5)^2 + (5-5)^2 + (4-5)^2 + (4-5)^2 + (6-5)^2 \\ &= (-2)^2 + 0 + (-1)^2 + (-1)^2 + 1 = 10 \end{aligned}$$

$$R^2 = 1 - \frac{3,6}{10} = 1 - 0,36 = 0,64$$