

# Assignment 1

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1. a. The linear regression model can be written as,

$$P = \beta_0 + \beta_1 C + \epsilon$$

where,

- $P$  is the prime interest rate
- $C$  is the core inflation rate
- $\beta_0$  is the intercept of the regression line
- $\beta_1$  is the slope of the regression line
- $\epsilon$  is the error term

Assumptions of the linear regression model are,

- i. **Linearity:** The relationship between  $P$  and  $C$  is linear
  - ii. **Independence:** The error term  $\epsilon$  is independent of  $C$
  - iii. **Normality:** The error term  $\epsilon$  is normally distributed
  - iv. **Homoscedasticity:** The error term  $\epsilon$  has a constant variance across all levels of  $C$
2. If the simple linear regression model can be written as  $y_i = \beta_0 + \beta_1 x_i + \epsilon_i$ , where  $E[\epsilon_i] = 0$  and  $E[\epsilon_i^2] = \sigma^2$  and where the  $\epsilon$ 's are independent, the  $i$ th fitted value is denoted by  $\hat{y}_i$  and the  $i$ th residual is denoted by  $e_i$  ( $i = 1, 2, \dots, n$ ), we can show that:
- (a)  $\sum_{i=1}^n e_i = 0$

*Proof.* Let  $e_i$  be the  $i$ th residual term. Then for each observation, we can see that the residual for each observation  $i$  is defined as:

$$e_i = y_i - \hat{y}_i$$

where,

- $y_i$  is the actual observed value of the dependent variable.
- $\hat{y}_i$  is the predicted value obtained from the regression model.

then we can say that the predicted value  $\hat{y}_i$  is equivalent to

$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i$$

Where,

- $\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$
- $\hat{\beta}_1 = \frac{S_{xy}}{S_{xx}}$
- $S_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$
- $S_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2$

We want to show that:

$$\sum_{i=1}^n e_i = 0$$

LHS:

$$\begin{aligned} \sum_{i=1}^n e_i &= \sum_{i=1}^n (y_i - \hat{y}_i) \\ &= \sum_{i=1}^n (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i) \\ &= \sum_{i=1}^n (y_i - \bar{y} + \hat{\beta}_1 \bar{x} - \hat{\beta}_1 x_i) \\ &= \sum_{i=1}^n y_i - n\bar{y} + \hat{\beta}_1 n\bar{x} - \hat{\beta}_1 \sum_{i=1}^n x_i \\ &= n\bar{y} - n\bar{y} + \hat{\beta}_1 n\bar{x} - \hat{\beta}_1 n\bar{x} \\ &= 0 \end{aligned}$$

$$LHS = 0 = RHS$$

$\therefore$  The sum of residuals  $e_i$  is zero.

□