

USING LSE METHOD

- From the model, make the error term the subject

$$y = \alpha + \beta x + \epsilon, \quad \text{make the error term the subject : } \epsilon = y - \alpha - \beta x$$

- Form sums of squares

$$\sum \epsilon^2 = \sum (y - \alpha - \beta x)^2$$

- Minimize the sums of squares (differentiate with respect to the parameters) and equate the results to zero.

$$\frac{d \sum \epsilon^2}{d\alpha} = -2 \sum (y - \alpha - \beta x) = 0, \text{ divide through by } -2 \text{ and equate the result to zero.}$$

$$\sum (y - \alpha - \beta x) = 0, \quad \text{open the brackets and make } y \text{ the subject.}$$

$$\sum y = n\alpha + \beta \sum x. \dots\dots\dots(a)$$

differentiate with respect to β to get the second normal equation

$$\frac{d \sum \epsilon^2}{d\beta} = -2x \sum (y - \alpha - \beta x) = 0. \text{ divide through by } -2 \text{ and equate to zero.}$$

$$\text{multiply by } x \text{ and make } y \text{ the subject. } \sum xy = \alpha \sum x + \beta \sum x^2 \dots\dots\dots(b)$$

- Expressions a and b are normal regression equations that can be solved to determine the parameters α and β .
- Solving expression a, divide through by n and make α the subject.

$$\alpha = \bar{y} - \beta \bar{x} \dots\dots\dots(c)$$

solving expressions a and b using cramer's rule. Write the two expressions in matrix format.

$$\begin{bmatrix} n & \sum x \\ \sum x & \sum x^2 \end{bmatrix} * \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} \sum y \\ \sum xy \end{bmatrix}$$

Solve for α and β

$\alpha = \frac{\sum x^2 \sum y - \sum x \sum xy}{n \sum x^2 - (\sum x)^2}$ and $\beta = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2}$ or $\beta = \frac{\sum(x-\bar{x})(y-\bar{y})}{\sum(x-\bar{x})^2}$ or $\beta = \frac{\sum xy}{\sum x^2}$ as the least squares estimators.

EXAMPLE I: show that LSE are unbiased.

SOLUTION

Required to show that the estimators $\hat{\beta}$ and $\hat{\alpha}$ are unbiased for the parameters β and α in the model $y = \alpha + \beta x_i + \varepsilon_i$. that is ; $E(\hat{\beta}) = \beta$ and $E(\hat{\alpha}) = \alpha$.

a) Show that $E(\hat{\alpha}) = \alpha$. From the expression $\hat{\alpha} = \bar{y} - \beta \bar{x}$, introduce expectations on either side $E(\hat{\alpha}) = E(\bar{y} - \beta \bar{x})$. substitute for $\bar{y} = \alpha + \beta \bar{x} + \bar{\varepsilon}$.

- $E(\hat{\alpha}) = E(\bar{y} - \beta \bar{x}) = E(\alpha + \beta \bar{x} + \bar{\varepsilon} - \beta \bar{x}) = E(\alpha + \bar{\varepsilon})$, but from the classical assumptions $E(\varepsilon) = 0$ and the expectation of a constant is a constant.
- therefore; $E(\hat{\alpha}) = \alpha$.

b) Show that $E(\hat{\beta}) = \beta$

- Using $\beta = \frac{\sum xy}{\sum x^2}$, substitute f for y using $y = \alpha + \beta x + e$, where lower case letters refer to a deviation from the mean. $x = X$ minus mean of X and e is the error minus mean of the errors.
- Introduce expectations on either side $E(\hat{\beta}) = E\left(\frac{\sum x(\alpha + \beta x + e)}{\sum x^2}\right) = E\left(\frac{\alpha \sum x + \beta \sum x^2 + \sum xe}{\sum x^2}\right)$, any deviations from the mean is always equal to zero $\sum x = 0$ and $E(xe) = 0$.
- $E(\hat{\beta}) = E\left(\frac{\beta \sum x^2}{\sum x^2}\right) = E(\beta)$, but the expectation or mean of a constant is a constant.
- therefore; $E(\hat{\beta}) = \beta$. this implies unbiasedness.

MEAN AND VARIANCE OF THE PARAMETERS α AND β

The mean for $\hat{\alpha} = E(\hat{\alpha}) = \alpha$ and the variance $v(\hat{\alpha}) = v(\bar{y} - \hat{\beta} \bar{x})$ this is from $\hat{\alpha} = \bar{y} - \hat{\beta} \bar{x}$. Therefore; $v(\hat{\alpha}) = v(\bar{y}) + \bar{x}^2 v(\hat{\beta})$. substituting for $v(Y)$ gives $v(\hat{\alpha}) = \delta^2 \left[\frac{1}{n} + \frac{\bar{x}^2}{\sum(x-\bar{x})^2} \right]$

The mean for $\hat{\beta} = E(\hat{\beta}) = \beta$ and the variance $v(\hat{\beta}) = \frac{\delta^2}{\sum(x_i - \bar{x})^2}$ where $\delta^2 = \frac{\sum e_i^2}{n-p} = \sum \frac{(y_i - \hat{y}_i)^2}{n-p}$. OR $\delta^2 = \frac{\sum y^2 - \alpha \sum y - \beta \sum xy}{n-p}$, where p are the parameters to be estimated. For simple regression model where only α and β are estimated, $p = 2$.

EXAMPLE: USING THE DATA BELOW, FIT A REGRESSION EQUATION TO THE DATA, DETERMINE THE MEAN AND VARIANCE OF THE ESTIMATED PARAMETERS.

X	77	50	71	72	81	94	96	99	67
Y	82	66	78	34	47	85	99	99	98