

Q.3) $y_i = \beta_1 x_{i1} + \beta_2 x_{i2} + \varepsilon_i, \quad i = 1, 2, \dots, n$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \quad X = \begin{bmatrix} x_{11} & x_{12} \\ x_{21} & x_{22} \\ \vdots & \vdots \\ x_{n1} & x_{n2} \end{bmatrix} \quad \beta = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} \quad \rightarrow \text{intercept doesn't exist.}$$

$$\hat{\beta} = (X'X)^{-1} X'y, \quad X = [x_1, x_2]$$

$$1) \quad X'X = \begin{bmatrix} x_{11} & x_{12} & \dots & x_{1n} \\ x_{21} & x_{22} & \dots & x_{2n} \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^n x_{1i}^2 & \sum_{i=1}^n x_{1i} x_{2i} \\ \sum_{i=1}^n x_{2i} x_{1i} & \sum_{i=1}^n x_{2i}^2 \end{bmatrix}$$

$$= \begin{bmatrix} x_1'x_1 & x_1'x_2 \\ x_2'x_1 & x_2'x_2 \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$$

$$\therefore (X'X)^{-1} = \frac{1}{3} \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$$

$$2) \quad X'y = \begin{bmatrix} x_1'y \\ x_2'y \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\therefore \hat{\beta} = \frac{1}{3} \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix}$$

$$\therefore y_i = 1 \cdot x_{i1} + 0 \cdot x_{i2} + \varepsilon_i, \quad i = 1, 2, \dots, n.$$

$$\text{Var}(\hat{\beta}) = \text{Var}((X'X)^{-1} X'y)$$

$$= (X'X)^{-1} X' \text{Var}(y) (X'X)^{-1} X'$$

$$= (X'X)^{-1} X' \text{Var}(y) X (X'X)^{-1}$$

$$\rightarrow \text{Var}(y) = \text{Var}(X\beta - \varepsilon) = \text{Var}(\varepsilon)$$

$$\therefore \text{Var}(y) = \sigma^2 I$$

$$= \sigma^2 I (X'X)^{-1} X' X (X'X)^{-1}$$

$$\therefore \text{Var}(\hat{\beta}) = \sigma^2 (X'X)^{-1}$$

1-1) Der. P.I. for observation 12.

$$y_{12} = x_{12}'\beta + \varepsilon_{12}, \quad \hat{y}_{12} = x_{12}'\hat{\beta}$$

$$\begin{aligned} \text{Var}(y_{12} - \hat{y}_{12}) &= \text{Var}(y_{12}) + \text{Var}(\hat{y}_{12}) \\ &= \text{Var}(x_{12}'\beta + \varepsilon_{12}) + \text{Var}(x_{12}'\hat{\beta}) \\ &= \text{Var}(\varepsilon_{12}) + x_{12}' \text{Var}(\hat{\beta}) x_{12} \\ &= \sigma^2 + \sigma^2 x_{12}' (X'X)^{-1} x_{12} \\ &= \sigma^2 (1 + x_{12}' (X'X)^{-1} x_{12}) \end{aligned}$$

$$i) \quad x_{12} = \begin{bmatrix} x_{12} \\ x_{22} \end{bmatrix} = \begin{bmatrix} 5 \\ -2 \end{bmatrix} \quad \therefore x_{12}' = \begin{bmatrix} 5 & -2 \end{bmatrix}$$

$$x_{12}' (X'X)^{-1} x_{12}$$

$$= \begin{bmatrix} 5 & -2 \end{bmatrix} \begin{bmatrix} \frac{2}{3} & -\frac{1}{3} \\ -\frac{1}{3} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 5 \\ -2 \end{bmatrix} = 26$$

ii) σ^2 : We don't know it, instead,

we use s^2 , which is $\frac{\sum \varepsilon_i^2}{n-p} \Rightarrow \frac{\sum (y_i - \hat{y}_i)^2}{n-p}$ (n : sample size, p : no. of parameters)

$$\begin{aligned} \therefore s^2 &= \frac{\sum (y_i - \hat{y}_i)^2}{n-2} = \frac{\sum (y_i - x_i'\hat{\beta})^2}{9} = \frac{\sum (y_i^2 - 2y_i x_i'\hat{\beta} + x_i'\hat{\beta}^2)}{9} \\ &= \frac{9 - 2 \cdot 2 + 2}{9} = \frac{1}{27} \end{aligned}$$

$$\therefore \text{Var}(y_{12} - \hat{y}_{12}) = \frac{1}{27} \times (1 + 26) = 1.$$

Prediction Interval for observation 12.

$$\hat{y}_{12} = 1 \cdot x_{12} + 0 \cdot x_{22}$$

$$= 1 \cdot 5 + 0 \cdot (-2) = 5$$

$$\therefore 5 \pm 1.383 \times \sqrt{1} \Rightarrow (3.617, 6.383) \rightarrow t_{0.1,9} = 1.383$$

1-2) Der. P.I. for observation 13.

$$i) \quad x_{13} = \begin{bmatrix} x_{13} \\ x_{23} \end{bmatrix} = \begin{bmatrix} 3 \\ -7 \end{bmatrix} \quad \therefore x_{13}' = \begin{bmatrix} 3 & -7 \end{bmatrix}$$

$$x_{13}' (X'X)^{-1} x_{13}$$

$$= \begin{bmatrix} 3 & -7 \end{bmatrix} \begin{bmatrix} \frac{2}{3} & -\frac{1}{3} \\ -\frac{1}{3} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 3 \\ -7 \end{bmatrix} = \frac{158}{3}$$

$$\text{Var}(y_{13} - \hat{y}_{13}) = \sigma^2 (1 + x_{13}' (X'X)^{-1} x_{13})$$

$$= \frac{1}{27} \times (1 + \frac{158}{3}) = \frac{1}{27} \times (\frac{161}{3}) = \frac{161}{81}$$

Prediction Interval for observation 13.

$$\hat{y}_{13} = 1 \cdot x_{13} + 0 \cdot x_{23}$$

$$= 1 \cdot 3 + 0 \cdot (-7) = 3$$

$$\therefore 3 \pm 1.383 \times \sqrt{\frac{161}{81}} = 3 \pm 1.383 \times 1.41 = 3 \pm 1.95003$$

$$\Rightarrow (1.04997, 4.95003)$$

2-1) Expected Value $y_{12} \rightarrow \hat{y}_{12}$

$$\hat{y}_{12} = x_{12}'\hat{\beta}$$

$$\text{Var}(\hat{y}_{12}) = \text{Var}(x_{12}'\hat{\beta})$$

$$= x_{12}' \text{Var}(\hat{\beta}) x_{12}$$

$$\therefore \text{Var}(\hat{\beta}) = \sigma^2 (X'X)^{-1}$$

$$= \sigma^2 x_{12}' (X'X)^{-1} x_{12}$$

From above, we know $\sigma^2 = \frac{1}{27}$, and $x_{12}' (X'X)^{-1} x_{12} = 26$.

$$\therefore \text{Var}(\hat{y}_{12}) = \frac{26}{27}$$

$$\hat{y}_{12} = x_{12}'\hat{\beta} = \begin{bmatrix} 5 & -2 \end{bmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = 5$$

Prediction Intervals for Expected Value y_{12} is

$$5 \pm 1.383 \times \sqrt{\frac{26}{27}} = 5 \pm 1.383 \times 0.9813$$

$$\Rightarrow (3.642, 6.358)$$

2-2) Expected Value of $y_{13} \rightarrow \hat{y}_{13}$

$$\hat{y}_{13} = x_{13}'\hat{\beta}$$

$$\text{Var}(\hat{y}_{13}) = \text{Var}(x_{13}'\hat{\beta})$$

$$= x_{13}' \text{Var}(\hat{\beta}) x_{13}$$

$$= \sigma^2 x_{13}' (X'X)^{-1} x_{13}$$

From above, we know $\sigma^2 = \frac{1}{27}$, and $x_{13}' (X'X)^{-1} x_{13} = \frac{158}{3}$

$$\therefore \text{Var}(\hat{y}_{13}) = \frac{158}{81}$$

$$\hat{y}_{13} = x_{13}'\hat{\beta} = \begin{bmatrix} 3 & -7 \end{bmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = 3$$

Prediction Intervals for Expected Value y_{13} is

$$3 \pm 1.383 \times \sqrt{\frac{158}{81}} = 3 \pm 1.383 \times 1.3966$$

$$\Rightarrow (1.0686, 4.9314)$$

3). The predicted intervals of observations of 12 & 13 is longer (wider) than the one for the expected value of y_{12} & y_{13} .

This can be interpreted that the differences are derived from that the predicted intervals of the observations 12, 13 take into consideration of ε , while the expected value of y_{12} & y_{13} don't.