STK333 Pengantar Model Linear

Model Linier Penuh

- Model Penuh
- Pendugaan Parameter Model

Model Linier:

$$y_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \dots + \beta_k x_k + \varepsilon_i, i = 1, 2, \dots, n$$

Model ini disebut model linier karena parameter-parameternya linier

Dengan notasi matriks:

$$\mathbf{y} = \mathbf{X}\mathbf{\beta} + \mathbf{\epsilon}$$

di mana: $\mathbf{y}_{n \times 1}$ = vektor respon $X_{n \times (k+1)} = \text{matriks peubah bebas (independent)}$ $\boldsymbol{\beta}_{(k+1) \times 1} = \text{vektor parameter}$ $\boldsymbol{\varepsilon}_{n \times 1} = \text{vektor peubah acak}$

Contoh Model non Linier:

$$y = e^{\beta_0 + \beta_1 x} + \epsilon$$

$$y = \frac{1}{(1 + e^{-\beta_0 - \beta_1 x + \epsilon})}$$

Contoh:

$$y = \beta_0 + \beta_1 x + \epsilon$$

$$y_{1} = \beta_{0} + \beta_{1}x_{1} + \epsilon_{1}$$

$$y_{2} = \beta_{0} + \beta_{1}x_{2} + \epsilon_{2}$$

$$\vdots$$

$$y_{n} = \beta_{0} + \beta_{1}x_{n} + \epsilon_{n}$$

$$y = \begin{bmatrix} y_{1} \\ y_{2} \\ \vdots \\ y_{n} \end{bmatrix} \quad \boldsymbol{\epsilon} = \begin{bmatrix} \beta_{0} \\ \beta_{1} \end{bmatrix} \quad \boldsymbol{\epsilon} = \begin{bmatrix} \varepsilon_{1} \\ \epsilon_{2} \\ \vdots \\ \epsilon_{n} \end{bmatrix} \quad \boldsymbol{X} = \begin{bmatrix} 1 & x_{1} \\ 1 & x_{2} \\ \vdots & \vdots \\ 1 & x_{n} \end{bmatrix}$$

Pendugaan koefisien regresi $\beta_0 \mathrm{dan} \ \beta_1 \mathrm{berdasarkan}$ sistem persamaan linier dengan dua bilangan yang tidak diketahui.

Model Linier disebut model penuh bila $r(X) = k+1 \rightarrow |X'X| \neq 0$

Model
$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

Asumsi: $\mathsf{E}[\boldsymbol{\varepsilon}] = \mathbf{0}$, $\mathsf{V}[\boldsymbol{\varepsilon}] = \sigma^2 I$, $\sigma^2 > 0$
maka $\mathsf{E}[\mathbf{y}] = \mathsf{E}[\mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}] = \mathbf{X}\boldsymbol{\beta} + \mathsf{E}[\boldsymbol{\varepsilon}] = \mathbf{X}\boldsymbol{\beta}$
 $\boldsymbol{\varepsilon} = \mathbf{y} - \mathbf{X}\boldsymbol{\beta} = \mathbf{y} - \mathsf{E}[\mathbf{y}]$

$$X'X = \begin{pmatrix} n & \sum x_i \\ \sum x_i & \sum x_i^2 \end{pmatrix}$$

Bila ada penduga parameter $\widehat{\boldsymbol{\beta}}$ = **b** dan $\widehat{\boldsymbol{\epsilon}}$ = **e** maka **e**=**y** - X**b**

Penduga parameter $\widehat{\beta}$ diperoleh dengan metode kuadrat terkecil (*Least Square*) dengan cara meminimumkan $\epsilon' \epsilon \rightarrow LS$ estimator

Penentuan *least square estimator* bagi β_0 , β_1 , ..., β_k

$$y_{1} = b_{0} + b_{1}x_{11} + b_{2}x_{12} + \dots + b_{k}x_{1k} + e_{1}$$

$$y_{2} = b_{0} + b_{1}x_{21} + b_{2}x_{22} + \dots + b_{k}x_{2k} + e_{2}$$

$$\vdots$$

$$y_{n} = b_{0} + b_{1}x_{n1} + b_{2}x_{n2} + \dots + b_{k}x_{nk} + e_{n}$$

$$X = \begin{bmatrix} 1 & x_{11} & x_{12} & \cdots & x_{1k} \\ 1 & x_{21} & x_{22} & \cdots & x_{2k} \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & x_{n1} & x_{n2} & & x_{nk} \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} b_0 \\ b_1 \\ \vdots \\ b_k \end{bmatrix}, \quad \mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \quad \text{and} \quad \mathbf{e} = \begin{bmatrix} e_1 \\ e_2 \\ \vdots \\ e_n \end{bmatrix}$$

Penentuan *least square estimator* bagi β_0 , β_1 , ..., β_k

Model
$$y = Xb + e$$

Metode kuadrat terkecil meminimumkan jumlah kuadrat sisaan (residual) untuk mendapatkan penduga koefisien regresi b_0, b_1, \dots, b_k yaitu meminimumkan

$$\mathbf{e}'\mathbf{e} = \sum_{i=1}^{n} e_i^2 = (\mathbf{y} - X\mathbf{b})'(\mathbf{y} - X\mathbf{b}) = y'y - y'Xb - (Xb)'y + (Xb)'Xb$$
$$= y'y - y'Xb - b'X'y + b'X'Xb$$

Penentuan least square estimator (LSE) bagi $\beta_0, \beta_1, ..., \beta_k$

$$\frac{\partial \mathbf{e}' \mathbf{e}}{\partial \mathbf{b}} = -2X'\mathbf{y} + (X'X)\mathbf{b} + (X'X)'\mathbf{b}$$

$$= -2X'\mathbf{y} + 2(X'X)\mathbf{b}$$

$$-2X'\mathbf{y} + 2(X'X)\mathbf{b} = \mathbf{0}$$

$$(X'X)\mathbf{b} = X'\mathbf{y}$$

$$\mathbf{b} = (X'X)^{-1}X'\mathbf{y}$$

Contoh 1:

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \varepsilon$$

$$\mathbf{y} = \begin{bmatrix} 50 \\ 40 \\ 52 \\ 47 \\ 65 \end{bmatrix} \qquad X = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 5 & 1 \\ 1 & 5 & 2 \\ 1 & 10 & 2 \\ 1 & 20 & 3 \end{bmatrix} \qquad X'X = \begin{bmatrix} 5 & 41 & 9 \\ 41 & 551 & 96 \\ 9 & 96 & 19 \end{bmatrix} \qquad X'\mathbf{y} = \begin{bmatrix} 254 \\ 2280 \\ 483 \end{bmatrix}$$

Contoh 1:

$$(X'X)^{-1} = \begin{bmatrix} 2.307551 & 0.1565378 & -1.88398 \\ 0.1565378 & 0.02578269 & -0.20442 \\ -1.88398 & -0.20442 & 1.977901 \end{bmatrix}$$

$$(X'X)^{-1}X'\mathbf{y} = \begin{bmatrix} 2.307551 & 0.1565378 & -1.88398 \\ 0.1565378 & 0.02578269 & -0.20442 \\ -1.88398 & -0.20442 & 1.977901 \end{bmatrix} \begin{bmatrix} 254 \\ 2280 \\ 483 \end{bmatrix}$$

$$\vdots \begin{bmatrix} 33.06 \\ -0.189 \end{bmatrix}$$

Contoh 1:

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \varepsilon$$

$$y = b_0 + b_1 x_1 + b_2 x_2 + e$$

$$y = 33.06 - 0.189x_1 + 10.718x_2 + e$$

$$y = b_0 + b_1 x + e$$

$$\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \qquad \mathbf{b} = \begin{bmatrix} b_0 \\ b_1 \end{bmatrix} \qquad x = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix}$$

$$X'X = \begin{bmatrix} n & \sum_{i=1}^{n} x_i \\ \sum_{i=1}^{n} x_i & \sum_{i=1}^{n} x_i^2 \end{bmatrix} \qquad X'y = \begin{bmatrix} \sum_{i=1}^{n} y_i \\ \sum_{i=1}^{n} x_i y_i \end{bmatrix}$$

$$(X'X)^{-1} = \frac{1}{n \sum_{i=1}^{n} x_i^2 - \left(\sum_{i=1}^{n} x_i\right)^2} \begin{bmatrix} \sum_{i=1}^{n} x_i^2 & -\sum_{i=1}^{n} x_i \\ -\sum_{i=1}^{n} x_i & n \end{bmatrix}$$

$$\mathbf{b} = (X'X)^{-1}X'\mathbf{y}$$

$$= \frac{1}{n \sum_{i=1}^{n} x_{i}^{2} - \left(\sum_{i=1}^{n} x_{i}\right)^{2}} \begin{bmatrix} \sum_{i=1}^{n} x_{i}^{2} & -\sum_{i=1}^{n} x_{i} \\ -\sum_{i=1}^{n} x_{i} & n \end{bmatrix} \begin{bmatrix} \sum_{i=1}^{n} y_{i} \\ \sum_{i=1}^{n} x_{i} y_{i} \end{bmatrix}$$

$$= \frac{1}{n \sum_{i=1}^{n} x_{i}^{2} - \left(\sum_{i=1}^{n} x_{i}\right)^{2}} \begin{bmatrix} \sum_{i=1}^{n} x_{i}^{2} \sum_{i=1}^{n} y_{i} - \sum_{i=1}^{n} x_{i} \sum_{i=1}^{n} x_{i} y_{i} \\ n \sum_{i=1}^{n} x_{i} y_{i} - \sum_{i=1}^{n} x_{i} \sum_{i=1}^{n} y_{i} \end{bmatrix}$$

$$y = b_0 + b_1 x + e$$

$$\frac{1}{n\sum_{i=1}^{n}x_{i}^{2}-\left(\sum_{i=1}^{n}x_{i}\right)^{2}}\begin{bmatrix}\sum_{i=1}^{n}x_{i}^{2}\sum_{i=1}^{n}y_{i}-\sum_{i=1}^{n}x_{i}\sum_{i=1}^{n}x_{i}y_{i}\\n\sum_{i=1}^{n}x_{i}y_{i}-\sum_{i=1}^{n}x_{i}\sum_{i=1}^{n}y_{i}\end{bmatrix}$$

$$b_{1}=\frac{n\sum_{i=1}^{n}x_{i}y_{i}-\sum_{i=1}^{n}x_{i}\sum_{i=1}^{n}y_{i}}{n\sum_{i=1}^{n}x_{i}^{2}-\left(\sum_{i=1}^{n}x_{i}\right)^{2}}$$

$$b_1 = \frac{n \sum_{i=1}^{n} x_i y_i - \sum_{i=1}^{n} x_i \sum_{i=1}^{n} y_i}{n \sum_{i=1}^{n} x_i^2 - \left(\sum_{i=1}^{n} x_i\right)^2}$$

$$b_0 = \frac{\sum_{i=1}^n x_i^2 \sum_{i=1}^n y_i - \sum_{i=1}^n x_i \sum_{i=1}^n x_i y_i}{n \sum_{i=1}^n x_i^2 - \left(\sum_{i=1}^n x_i\right)^2}$$

Teorema:
$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$
, $r(\mathbf{X}) = k+1$, $E[\boldsymbol{\varepsilon}] = \mathbf{0}$, $V[\boldsymbol{\varepsilon}] = \sigma^2 I$ maka penduga $\boldsymbol{\beta}$ adalah $\mathbf{b} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y}$

$$E(\mathbf{b}) = E[(X'X)^{-1}X'\mathbf{y}] = (X'X)^{-1}X'E(\mathbf{y}) = (X'X)^{-1}X'X\beta = \beta$$

$$var(\mathbf{b}) = var[(X'X)^{-1}X'\mathbf{y} = (X'X)^{-1}X'[var(\mathbf{y})][(X'X)^{-1}X']'$$

$$= (X'X)^{-1}X'(\sigma^{2}I)X(X'X)^{-1} \qquad \Rightarrow var(A\mathbf{y}) = A[var(\mathbf{y})]A'$$

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

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

b tidak berbias dan
$$var(\mathbf{b}) = \sigma^2(X'X)^{-1}$$

Teorema Gauss-Markoff:

$$\mathbf{y} = X\mathbf{\beta} + \mathbf{\varepsilon}$$
, $r(X) = k+1$, $E[\mathbf{\varepsilon}] = \mathbf{0}$, $V[\mathbf{\varepsilon}] = \sigma^2 I$ maka penduga $\mathbf{\beta}$ adalah $\mathbf{b} = (X'X)^{-1}X'\mathbf{y}$ b adalah penduga tak bias linier terbaik (BLUE) bagi $\mathbf{\beta}$

Teorema:

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$
, $\mathbf{r}(\mathbf{X}) = \mathbf{k} + \mathbf{1}$, $\mathbf{E}[\boldsymbol{\varepsilon}] = \mathbf{0}$, $\mathbf{V}[\boldsymbol{\varepsilon}] = \sigma^2 \mathbf{I}$ bila $\mathbf{t}_{(\mathbf{k}+1)\times 1}$ adalah vektor bilangan nyata tidak nol maka BLUE $\mathbf{t}'\boldsymbol{\beta}$ adalah $\mathbf{t}'\mathbf{b}$ dimana \mathbf{b} adalah penduga $\boldsymbol{\beta}$

Contoh 3:

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \varepsilon$$

$$y = 33.06 - 0.189x_1 + 10.718x_2 + \varepsilon$$

$$t' = \begin{bmatrix} 1 & 15 & 2.5 \end{bmatrix}$$

$$\widehat{E[y]} = t'b = \begin{bmatrix} 1 & 15 & 2.5 \end{bmatrix}b$$

$$= b_0 + 15b_1 + 2.5b_2$$

$$\widehat{E[y]} = 33.06 - 0.189(15) + 10.718(2.5) = 57.02$$

Latihan:

i	Х	У
1	8	8
2	12	15
3	14	16
4	16	20
5	16	25
6	20	40

Berdasarkan data tersebut:

- 1. Tentukan vektor **y** dan matriks X.
- 2. Tentukan X'X, X'y, dan $(X'X)^{-1}$.
- 3. Tenukan penduga koefisien regresi, b_0 dan b_1 , dengan $(X'X)^{-1} X'y$.
- 4. Ada suatu vektor $\mathbf{t}' = [1 \ 15]$, tentukan nilai harapan $\widehat{E[y]} = \mathbf{t}'\mathbf{b}$.