

Simple Linear Regression

① theory $Y = \alpha + \beta X + \epsilon$, $\mu_{Y|X} = \alpha + \beta X$ ("true" regression line)
 $\hat{Y} = A + Bx$, $\hat{y} = a + bx$ (fitted regression line; 拟合回归线)

process $Y_i = \alpha + \beta x_i + \epsilon_i$ (random variable of y_i for a given x_i)
 $\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$ (sample mean)
 $Y_i = A + Bx_i$, $\hat{y}_i = a + bx_i$

result $Y_0 = \alpha + \beta x_0 + \epsilon$
 $\hat{Y}_0 = A + Bx_0$, $\hat{y}_0 = a + bx_0$ (point predictor of y_0)

② Assume: $E\epsilon = 0$, $\text{Var}(\epsilon) = \sigma^2$; $E\epsilon_i = 0$, $\text{Var}(\epsilon_i) = \sigma^2$

$EY = \alpha + \beta X$, $\text{Var}(Y) = \sigma^2$

$EY_i = \alpha + \beta x_i$, $\text{Var}(Y_i) = \sigma^2$

$EB = \beta$, $\text{Var}(B) = \frac{\sigma^2}{S_{xx}}$

$EA = \alpha$, $\text{Var}(A) = \frac{\sum_{i=1}^n x_i^2}{n S_{xx}} \sigma^2$

$ES^2 = \sigma^2$

$E(Y_0 - \hat{Y}_0) = 0$, $\text{Var}(Y_0 - \hat{Y}_0) = \sigma^2 \left[1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right]$

③ $e_i \equiv y_i - \hat{y}_i$; $SSE \equiv \sum_{i=1}^n e_i^2$

$b = \frac{S_{xy}}{S_{xx}}$, $a = \bar{y} - b\bar{x}$

$S_{xx} \equiv \sum_{i=1}^n (x_i - \bar{x})^2$, $S_{yy} = \sum_{i=1}^n (y_i - \bar{y})^2$, $S_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$

$B = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$, $A = \bar{Y} - B\bar{x}$

$S^2 \equiv \frac{1}{n-2} \sum_{i=1}^n (y_i - \hat{y}_i)^2$, $s^2 = \frac{S_{yy} - bS_{xy}}{n-2}$

④ parameter = α , β , σ^2 .

point estimator = a , b , s^2 .

estimator = A , B , S^2

$$\textcircled{5} \quad \frac{B - \beta}{\frac{S}{\sqrt{S_{xx}}}} \sim t\text{-dist} \quad (V = n - 2)$$

$$\frac{A - \alpha}{S \sqrt{\frac{\sum x_i^2}{n S_{xx}}}} \sim t\text{-dist} \quad (V = n - 2)$$

$$\frac{\hat{y}_0 - y_0}{S \sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}}} \sim t\text{-dist} \quad (V = n - 2)$$

$$\textcircled{6} \quad (1 - \alpha) \times 100\% \text{ C.I. : } \beta = U(b, t_{\frac{\alpha}{2}} \frac{S}{\sqrt{S_{xx}}})$$

$$\alpha = U(a, t_{\frac{\alpha}{2}} \cdot S \cdot \sqrt{\frac{\sum x_i^2}{n S_{xx}}})$$

$$(1 - \alpha) \times 100\% \text{ prediction interval for } y_0: U(\hat{y}_0, t_{\frac{\alpha}{2}} \cdot S \cdot \sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}})$$

$$\text{testing for } \beta = a = \beta_0 + t_{\frac{\alpha}{2}} \frac{S}{\sqrt{S_{xx}}}; a, b = \beta \pm t_{\frac{\alpha}{2}} \frac{S}{\sqrt{S_{xx}}}$$

$$\text{for } \alpha = a = \alpha_0 + t_{\frac{\alpha}{2}} \cdot S \cdot \sqrt{\frac{\sum x_i^2}{n S_{xx}}}; a, b = \alpha \pm t_{\frac{\alpha}{2}} \cdot S \cdot \sqrt{\frac{\sum x_i^2}{n S_{xx}}}$$

$\textcircled{7}$ Sample Correlation:

$$r \equiv \frac{S_{xy}}{\sqrt{S_{xx} S_{yy}}} = b \sqrt{\frac{S_{xx}}{S_{yy}}} \quad (-1 \leq r \leq 1)$$

$$\chi^2 = \sum_{i=1}^k \frac{(O_i - e_i)^2}{e_i} \sim \chi^2 (V = k - 1)$$

$$\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim Z$$

$$\frac{\bar{X} - \mu}{\frac{S}{\sqrt{n}}} \sim t\text{-dist} \quad (V = n - 1)$$

$$\frac{(n-1) S^2}{\sigma^2} \sim \chi^2\text{-dist} \quad (V = n - 1)$$

You Say ENO!