

## 5. Prediction

new  $x_0$  to predict the  $y_0$

$$E(y_0) = \beta_0 + \beta_1 x_0$$

$$\text{var}(y_0) = \sigma^2$$

$$y_0 \perp \{y_i : i=1, 2, \dots, n\}$$

### 5.1 Point prediction 点(单值)预测

$$\hat{y}_0 = \hat{\beta}_0 + \hat{\beta}_1 x_0$$

$\hat{y}_0$  是对  $y_0$  的预测], 却是对  $E(y_0)$  的无偏估计

Proof

$$E(\hat{y}_0) = E(\hat{\beta}_0) + E(\hat{\beta}_1) \cdot x_0 = \beta_0 + \beta_1 x_0 = E(y_0)$$

故  $\hat{y}_0$  是对  $E(y_0)$  的无偏估计

Remark: 估计  $E(y_0)$ , 而没有估计  $y_0$

Remark:  $y_0$  是一个随机变量

$E(y_0)$  是一个未知的参数

### 5.2 Prediction Interval 区间预测

对于给定的显著性水平  $\alpha$  (例)  $\alpha = 0.05$ ), 找一个区间  $(T_1, T_2)$

使  $P(T_1 < y_0 < T_2) = 1 - \alpha$

## 5.2.1 $y_0$ Prediction Interval : PI

As we know

$$y_0 \sim N(\beta_0 + \beta_1 x_0, \sigma^2)$$

Remember:

$$\hat{y}_0 = \sum_{i=1}^n \left( \frac{1}{n} + \frac{(x_i - \bar{x})(x_0 - \bar{x})}{L_{xx}} \right) y_i$$

So:

$$\text{var}(\hat{y}_0) = \left( \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{L_{xx}} \right) \sigma^2$$

Therefore

$$\hat{y}_0 \sim N(\beta_0 + \beta_1 x_0, \left( \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{L_{xx}} \right) \sigma^2)$$

$$\text{ie } h_{00} = \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{L_{xx}}$$

独立性:  $y_0 \perp y_1, y_2, \dots, y_n$

$\hat{y}_0$  = linear function ( $y_1, y_2, \dots, y_n$ )

$\Rightarrow y_0 \perp \hat{y}_0$

二者是独立的, 正态分布的样本, 且

$$\begin{cases} E(y_0 - \hat{y}_0) = 0 \\ \text{var}(y_0 - \hat{y}_0) = \sigma^2 + h_{00} \sigma^2 \end{cases}$$

从而:

★  $y_0 - \hat{y}_0 \sim N(0, (1+h_{00}) \sigma^2)$

① t 分布

$$t = \frac{y_0 - \hat{y}_0}{\sqrt{1+h_{00}} \cdot \hat{\sigma}} \sim t(n-2)$$

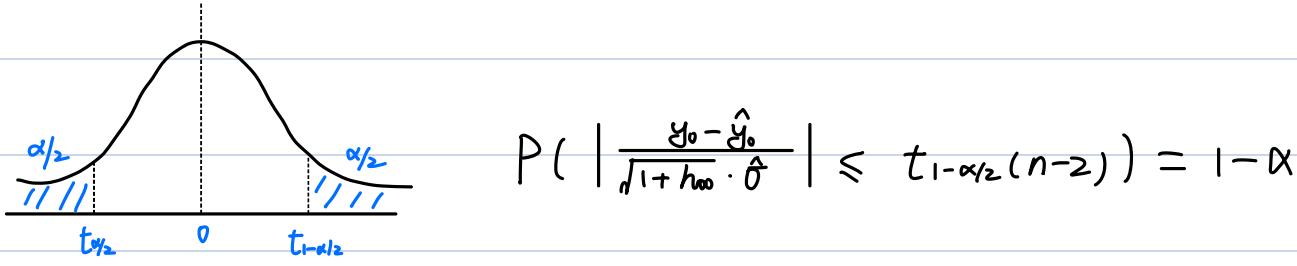
Proof:  $\frac{y_0 - \hat{y}_0}{\sqrt{1+h_{00}} \cdot \hat{\sigma}} \sim N(0,1)$

$$\frac{(n-2) \hat{\sigma}^2}{\hat{\sigma}^2} \sim \chi^2(n-2)$$

$$\Rightarrow \frac{y_0 - \hat{y}_0}{\sqrt{1+h_{00}} \cdot \hat{\sigma}} \sqrt{\frac{(n-2)\hat{\sigma}^2}{\hat{\sigma}^2/(n-2)}} \sim t(n-2)$$

If  $t = \frac{y_0 - \hat{y}_0}{\sqrt{1+h_{00}} \cdot \hat{\sigma}} \sim t(n-2)$

## ② PI ( $y_0$ 的预测区间)



PI  $[\hat{y}_0 - t_{1-\alpha/2}(n-1) \cdot \sqrt{1+h_{00}} \cdot \hat{\sigma}, \hat{y}_0 + t_{1-\alpha/2}(n-1) \cdot \sqrt{1+h_{00}} \cdot \hat{\sigma}]$

$\uparrow$   $y_0$  的预测区间

## 5.2.2 $E(y_0)$ Confidence Interval : CI

### ① $E(y_0)$ 参数

$$E(y_0) = \beta_0 + \beta_1 x_0$$

是一个参数 (未知的常数)

## ② t 分布

$$\hat{y}_0 \sim N(\beta_0 + \beta_1 x_0, h_{00} \sigma^2)$$

$$\hat{y}_0 - E(y_0) \sim N(0, h_{00} \sigma^2)$$

$$\frac{(n-2) \hat{\sigma}^2}{\sigma^2} \sim \chi^2(n-2)$$

\*  $t = \frac{\hat{y}_0 - E(y_0)}{\sqrt{h_{00}} \hat{\sigma}} \sim t(n-2)$

## ③ CI

$$P\left( \left| \frac{\hat{y}_0 - E(y_0)}{\sqrt{h_{00}} \hat{\sigma}} \right| \leq t_{1-\alpha/2}(n-2) \right) = 1-\alpha$$

**CI**  $[\hat{y}_0 - t_{1-\alpha/2}(n-2) \cdot \sqrt{h_{00}} \hat{\sigma}, \hat{y}_0 + t_{1-\alpha/2}(n-2) \cdot \sqrt{h_{00}} \hat{\sigma}]$

↑  $E(y_0)$  的置信区间

## 5.3 Relationship of PI, CI

$y_0$  PI  $[\hat{y}_0 - t_{1-\alpha/2}(n-1) \cdot \sqrt{1+h_{00}} \cdot \hat{\sigma}, \hat{y}_0 + t_{1-\alpha/2}(n-1) \cdot \sqrt{1+h_{00}} \cdot \hat{\sigma}]$

$E(y_0)$  CI  $[\hat{y}_0 - t_{1-\alpha/2}(n-2) \cdot \sqrt{h_{00}} \hat{\sigma}, \hat{y}_0 + t_{1-\alpha/2}(n-2) \cdot \sqrt{h_{00}} \hat{\sigma}]$

$\text{len(CI)} < \text{len(PI)}$  : CI 更精确

And  $n \rightarrow \infty$   $h_{00} \rightarrow 0$

CI  $\rightarrow \hat{y}_0$  PI  $\rightarrow [\hat{y}_0 \pm t_{1-\alpha/2}(n-2) \hat{\sigma}]$