

$$① \hookrightarrow S(\beta_0, \beta_1) = \sum [y_i - (\beta_1 x_i + \beta_0)]^2$$

$$\frac{\partial S}{\partial \beta_0} = \frac{\partial S}{\partial \beta_1} = 0 \Rightarrow \hat{\beta}_1 = \frac{S_{xy}}{S_{xx}} \quad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} \quad \text{过 } (\bar{x}, \bar{y})$$

$$\begin{aligned} 227 \quad \text{Cov}(\hat{\beta}_0, \hat{\beta}_1) &= \text{Cov}(\bar{y} - \hat{\beta}_1 \bar{x}, \hat{\beta}_1) \\ &= -\bar{x} \text{Var}(\hat{\beta}_1) \rightarrow = -\bar{x} \left(\frac{\sum (x_i - \bar{x})^2}{S_{xx}} \right) \text{Var}(y_i) \\ &= -\frac{\bar{x} \sigma^2}{S_{xx}} = -\frac{\sigma^2}{S_{xx}} \end{aligned}$$

$\bar{x} = 0$ 不相关.

$$237 \quad \text{Var}(\hat{\beta}_1) = \frac{\sigma^2}{S_{xx}} \quad \text{越小} \Rightarrow s^2 \text{ 越大}$$

可取 $\begin{bmatrix} 1 \\ \bar{x} \end{bmatrix} \perp \begin{bmatrix} 1 \\ \bar{x} \end{bmatrix} \Rightarrow \begin{bmatrix} 1 \\ \bar{x} \end{bmatrix} \perp \begin{bmatrix} 1 \\ \bar{x} \end{bmatrix}$

$$247 \quad \beta_0 = 0 \Rightarrow \beta_1 = \frac{\sum x_i y_i}{\sum x_i^2}$$

$$2. \hookrightarrow \varepsilon_i \sim N(0, \sigma^2) \Rightarrow y_i \sim N(\beta_0 + \beta_1 x_i, \sigma^2)$$

$$\Rightarrow L(\beta_0, \beta_1, \sigma^2) = \prod \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{[y_i - (\beta_0 + \beta_1 x_i)]^2}{2\sigma^2}\right)$$

$$\Rightarrow \ell(\beta_0, \beta_1, \sigma^2) = -\frac{n}{2} \ln 2\pi - \frac{n}{2} \ln \sigma^2 - \frac{1}{2\sigma^2} \sum (y_i - \beta_0 - \beta_1 x_i)^2$$

$$\frac{\partial \ell}{\partial \sigma^2} = 0 \Rightarrow \hat{\sigma}^2 = \frac{\sum (y_i - \beta_0 - \beta_1 x_i)^2}{n} = \frac{RSS}{n}$$

$$\begin{aligned}
 117 \quad RSS &= \sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2 = \sum (y_i - (\bar{y} - \hat{\beta}_1 \bar{x}) - \hat{\beta}_1 x_i)^2 \\
 &= \sum [(y_i - \bar{y}) - \hat{\beta}_1 (x_i - \bar{x})]^2 \\
 &= \cancel{S_{yy}} - 2\hat{\beta}_1 \cancel{S_{xy}} + \hat{\beta}_1^2 S_{xx} \\
 &= \cancel{S_{yy}} - \frac{S_{xy}^2}{S_{xx}}
 \end{aligned}$$

$$\begin{aligned}
 \Rightarrow E(RSS) &= E(S_{yy}) - \frac{E(S_{xy}^2)}{S_{xx}} = \beta_1^2 S_{xx} + (n-1)\sigma^2 - \frac{\beta_1^2 S_{xx}^2 + \sigma^2 S_{xx}}{S_{xx}} \\
 &= (n-2)\sigma^2
 \end{aligned}$$

$$\uparrow \quad \hat{\sigma}^2 = \frac{RSS}{n-2}$$

$$137 \quad \beta_0 = 0 \quad \text{则} \quad E(RSS) = (n-1)\sigma^2 \quad \uparrow \quad \hat{\sigma}^2 = \frac{RSS}{n-1}$$

$$\text{又} \quad X = X_0 \quad \text{则} \quad y_0 = \beta_1 x_0 + \varepsilon_0 \quad \hat{y}_0 = \hat{\beta}_1 x_0$$

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

$$\hat{y}_0 \sim N(\beta_1 x_0, \frac{\sigma^2}{S_{xx}} \cdot x_0^2)$$

$$\Rightarrow y_0 - \hat{y}_0 \sim N(0, (1 + \frac{x_0^2}{S_{xx}})\sigma^2)$$

$$\text{又} \quad \frac{(n-1)\hat{\sigma}^2}{\sigma^2} \sim \chi^2_{(n-1)} \quad \Rightarrow \quad \frac{y_0 - \hat{y}_0}{\hat{\sigma} \sqrt{1 + \frac{x_0^2}{S_{xx}}}} \sim t_{(n-1)}$$

$$\Rightarrow \hat{\beta}_1 x_0 \pm t_{\frac{\alpha}{2}, (n-1)} \cdot \hat{\sigma} \sqrt{1 + \frac{x_0^2}{S_{xx}}}$$