

1. Residual Sum of Squares

Distribution

As

$$\text{Res.S.S.} = \mathbf{Y}^T (\mathbf{I} - \mathbf{X}(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T) \mathbf{Y}$$

By Theorem 3.4, it can be proved that

$$\frac{\text{Res. S. S.}}{\sigma^2} \sim \chi^2(n - p', \lambda).$$

where $\lambda = 0$.

2. Independence of $\hat{\beta}$ and $\hat{\sigma}^2$

For $p = 1$

$$\text{Cov}(\hat{e}_i, \hat{\beta}_0) = 0 \text{ and } \text{Cov}(\hat{e}_i, \hat{\beta}_1) = 0$$

$\Rightarrow \hat{\sigma}^2$ and $(\hat{\beta}_0, \hat{\beta}_1)$ are independent.

For any p

$$\text{Cov}(\hat{\mathbf{e}}, \hat{\beta}) = 0$$

$\Rightarrow \hat{\sigma}^2$ and $\hat{\beta}$ are independent.

3. Confidence Interval & Hypothesis Testing

$$\hat{\beta} \sim N(\beta, (\mathbf{X}^T \mathbf{X})^{-1} \sigma^2)$$

$$\frac{(n - p') \hat{\sigma}^2}{\sigma^2} = \frac{RSS}{\sigma^2} \sim \chi_{(n - p')}$$

Res S.S. and $\hat{\beta}$ are independent.

For $p=1$

$$H_0 : \beta_1 = \beta_{10}$$

$$t_{\text{obs}} = \frac{\hat{\beta}_1 - \beta_{10}}{\hat{\sigma} / \sqrt{S_{x_1 x_1}}} \sim t_{n-2}$$

Reject H_0 if $|t_{\text{obs}}| > t_{\alpha/2, n-2}$ for two-sided alternative. Reject H_0 if $|t_{\text{obs}}| > t_{\alpha, n-2}$ for one-sided alternative.

$(1 - \alpha)100\%$ C.I. for β_1 is

$$\left(\hat{\beta}_1 - t_{\alpha/2, (n-2)} \frac{\hat{\sigma}}{\sqrt{S_{x_1 x_1}}}, \hat{\beta}_1 + t_{\alpha/2, (n-2)} \frac{\hat{\sigma}}{\sqrt{S_{x_1 x_1}}} \right)$$