

Hypothesis testing

1. Wald test

An approximate χ^2_1 statistic for testing

$$H_o : \beta_j = 0$$

$$H_i : \beta_j \neq 0$$

can be used by computing

$$\chi^2 = \frac{\hat{\beta}_j^2}{c^{jj}} \quad (j = 0, 1, \dots, p)$$

where the c^{jj} is the $(j + 1)^{th}$ diagonal element of \mathbf{C}^{-1} , where \mathbf{C}^{-1} is the estimated variance-covariance matrix of estimates of regression coefficients and the $(i, j)^{th}$ element in \mathbf{C} is defined as

$$c_{ij} = \frac{-\partial^2 \log L(\hat{\beta})}{\partial \hat{\beta}_i \partial \hat{\beta}_j} \quad (i = 0, 1, \dots, p; j = 0, 1, \dots, p)$$

Wald statistic for testing $\mathbf{L}^T \boldsymbol{\beta} = \mathbf{d}$ is defined by

$$(\mathbf{L}^T \hat{\boldsymbol{\beta}} - \mathbf{d})^T (\mathbf{L}^T \mathbf{C}^{-1} \mathbf{L})^{-1} (\mathbf{L}^T \hat{\boldsymbol{\beta}} - \mathbf{d})$$

where $\hat{\boldsymbol{\beta}}$ is the maximum likelihood estimates and \mathbf{C}^{-1} is its estimated covariance matrix. It is χ^2_r , where r is the rank of \mathbf{L} .

2. Likelihood ratio test

For testing

$$H_o : \beta_1 = 0$$

$$H_1 : \beta_1 \neq 0$$

$$\begin{aligned} \text{test statistic} &= \lambda(\boldsymbol{\beta}_1 | \boldsymbol{\beta}_2) \\ &= \lambda(\boldsymbol{\beta}_2) - \lambda(\boldsymbol{\beta}) \\ &\sim \chi^2_{(r)} \end{aligned}$$

That is, reject H_o if $\lambda(\boldsymbol{\beta}_1 | \boldsymbol{\beta}_2) > \chi^2_{\alpha, r}$

3. Score test

Score statistic is defined to be

$$\mathbf{U}^T(\boldsymbol{\beta}_0) \mathbf{I}^{-1}(\boldsymbol{\beta}_0) \mathbf{U}(\boldsymbol{\beta}_0).$$

Under $\boldsymbol{\beta} = \boldsymbol{\beta}_0$, score statistic $\sim \chi^2_r$ with r is the dimension of $\boldsymbol{\beta}_0$. $\mathbf{U}(\boldsymbol{\beta})$ is the vector of partial derivatives of the log-likelihood with respect to the parameter vector $\boldsymbol{\beta}$ and $\mathbf{I}(\boldsymbol{\beta})$ is the matrix of the negative second partial derivatives of the log-likelihood with respect to $\boldsymbol{\beta}$.