

Bartlett Correction for likelihood ratio test

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1 Likelihood

The likelihood ratio statistic is given as:

$$\Lambda(X, Y) = -2 \sup \{l(\theta|X, Y; \theta \in \Theta_0)\} - \sup \{l(\theta|X, Y; \theta \in \Theta)\},$$

We then maximize this function under two conditions,

$$l_{H_0} = \sup_{\alpha, \delta, \mu_j^x = \mu_j^y} \{l(\alpha, \delta, \mu_1, \dots, \mu_m)\} \quad (1)$$

$$l_{H_1} = \sup_{\alpha, \delta, \mu_1, \dots, \mu_m} \{l(\alpha, \delta, \mu_1, \dots, \mu_m)\} \quad (2)$$

The full log-likelihood can be written as

$$l(\alpha, \delta, \mu_j^x, \mu_j^y) = m \left(\ln \Gamma(\alpha + \frac{n}{2}) - \ln \Gamma(\alpha) - \frac{n}{2} \ln(2\pi\delta) \right) - (\alpha + \frac{n}{2}) \sum_{j=1}^m \ln \left(\frac{\zeta_j}{2} + \delta \right), \quad (3)$$

where for notational simplicity, we introduce the quantities $\zeta_j = \sum_{i=1}^{n_x} (x_{ij} - \mu_j^x)^2 + \sum_{i=1}^{n_y} (y_{ij} - \mu_j^y)^2$. and $n = n_x + n_y$. Applying an asymptotic expansion for the gamma function to Equation (3), the log-likelihood at the maximum is given by,

$$\lim_{\delta \rightarrow \infty} l(\delta) = \frac{nm}{2} \left(\ln \left(\frac{nm}{2\pi \sum_{j=1}^m \zeta_j} \right) - 1 \right). \quad (4)$$

2 Bartlett correction

We define the Bartlett correction as

$$\begin{aligned} BC &= E_{H_0} \{ \Lambda(x, y) \} \\ &= -2E_{H_0} (l_{H_0} - l_{H_1}) \end{aligned} \quad (5)$$

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with l_{H_0} and l_{H_1} defined in (1) and (2). By defining the small sample likelihood ratio statistic as $\Lambda_{BC}(x, y) = m \frac{\Lambda(x, y)}{BC}$, we achieve that precisely the Bartlett corrected likelihood ratio statistic has $E_{H_0} \{LR_{BC}\} = m$ as would be expected for a χ_m^2 distribution.

Calculation of these two expectations in general is very involved. If we use the characterization for $l(\delta)$ defined in (4), we can get an explicit approximate expression for the Bartlett correction,

$$BC \approx -2E_{H_0}(\lim_{\delta \rightarrow \infty} l(\delta|H_0) - \lim_{\delta \rightarrow \infty} l(\delta|H_1)) \quad (6)$$

$$= mnE \ln \left(\frac{\sum_{j=1}^m \zeta_j^{H_0}}{\sum_{j=1}^m \zeta_j^{H_1}} \right) \quad (7)$$

$$= nmE \ln \left(1 + \frac{\sum_{j=1}^m S_{\Delta,j}}{\sum_{j=1}^m (S_{x,j} + S_{y,j})} \right) \quad (8)$$

whereby

$$\begin{aligned} \zeta_j^{H_0} &= \sum_{i=1}^{n_x} (x_{ij} - \hat{\mu}_j)^2 + \sum_{i=1}^{n_y} (y_{ij} - \hat{\mu}_j)^2 \\ \zeta_j^{H_1} &= \underbrace{\sum_{i=1}^{n_x} (x_{ij} - \bar{x}_{\cdot j})^2}_{S_{x,j}} + \underbrace{\sum_{i=1}^{n_y} (y_{ij} - \bar{y}_{\cdot j})^2}_{S_{y,j}} \\ S_{\Delta,j} &= \frac{n_x n_y (\bar{x}_{\cdot j} - \bar{y}_{\cdot j})^2}{n}, \end{aligned}$$

with $\hat{\mu}_j = \frac{n_x \bar{x}_{\cdot j} + n_y \bar{y}_{\cdot j}}{n_x + n_y}$ and $\bar{x}_{\cdot j} = \frac{1}{n_x} \sum_{i=1}^{n_x} x_{ij}$. Since for small values of m , the supremum of the likelihood in (1) and (2) are actually found at such degenerate δ , the approximation in (6) can be exact in certain cases. The log expressions are normally distributed, so that under the assumption of no differential expression and in the degenerate case $\sigma_j^2 = \sigma^2$, we have that $\bar{x}_{\cdot j} - \bar{y}_{\cdot j} \sim N(0, \frac{\sigma^2(n_x + n_y)}{n_x n_y})$ and therefore, $\sum_{j=1}^m \frac{S_{\Delta,j}}{\sigma^2} \sim \chi_m^2$ and $\sum_{j=1}^m \frac{S_{x,j} + S_{y,j}}{\sigma^2} \sim \chi_{m(n-2)}^2$. The ratio of χ^2 distributions yields an F distribution, thus

$$BC \approx nmE \ln \left(1 + \frac{F_{m, m(n-2)}}{n-2} \right). \quad (9)$$

The density of $F_{m, m(n-2)}$ [Hogg et al., 2005, p.185] is

$$f(x) = \frac{(n-2)^{m/2} x^{m/2-1}}{B(m, m(n-2))(1+x/(n-2))^{m(n-1)/2}}, \quad x > 0 \quad (10)$$

and $B(x, y)$ is the Beta function. The expected value in (9) can be found by applying a transformation theorem [Hogg et al., 2005, p.55], [Gradshteyn and Ryzhik, 2000, p.555]

$$BC \approx \frac{mn}{B(m, m(n-2))} \times \int_0^\infty \frac{\ln(1+x/(n-2)) \nu^{m/2} x^{m/2-1}}{(1+x/(n-2))^{m(n-1)/2}} dx \quad (11)$$

$$\begin{aligned}
&= mn \left[\psi \left(\frac{m(n-1)}{2} \right) - \psi \left(\frac{m(n-2)}{2} \right) \right] \\
&= mn \left[\ln \left(1 + \frac{m/2}{m(n-2)/2} \right) + O(n^{-2}) \right] \tag{12}
\end{aligned}$$

$$\approx mn \ln \left(\frac{n-1}{n-2} \right) \tag{13}$$

$$\approx m \frac{n}{n-2} \tag{14}$$

From this it follows that the Bartlett corrected likelihood ratio statistic is found as

$$\Lambda_{BC}(x, y) \approx \frac{\Lambda(x, y)}{n \left[\psi \left(\frac{m(n-1)}{2} \right) - \psi \left(\frac{m(n-2)}{2} \right) \right]} \tag{15}$$

$$\approx \frac{\Lambda(x, y)}{n \ln \left(\frac{n-1}{n-2} \right)} \tag{16}$$

$$\approx \frac{n-2}{n} \Lambda(x, y) \tag{17}$$

References

- I. S. Gradshteyn and I. M. Ryzhik. *Table of Integrals, Series, and Products*. Academic Press, 2000.
- R. V. Hogg, J. W. McKean, and A. T. Craig. *Introduction to Mathematical Statistics*. Pearson Prentice Hall, 2005.