



Generalized Cordeiro–Ferrari Bartlett-type adjustment

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ABSTRACT

The Bartlett-type adjustment is a higher-order asymptotic method for reducing the errors of the chi-squared approximations to the null distributions of various test statistics, which ensures that the resulting test has size $\alpha + o(N^{-1})$, where $0 < \alpha < 1$ is the significance level and N is the sample size. Recently, [Kakizawa \(2012\)](#) has revisited the Chandra–Mukerjee/Taniguchi adjustments in a unified way, since [Chandra and Mukerjee \(1991\)](#) and [Taniguchi \(1991b\)](#) originally considered the test of the simple null hypothesis, except for [Mukerjee \(1992\)](#). This paper considers a generalization of the adjustment due to [Cordeiro and Ferrari \(1991\)](#).

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1. Introduction

The null distributions of the likelihood ratio, Rao's and Wald's test statistics have asymptotic expansions in powers of N^{-1} , where N is the sample size. The Bartlett/Bartlett-type adjustment is then designed to make the chi-squared approximation accurate up to order N^{-1} . Historically, it was first exploited by [Bartlett \(1937\)](#) in his classical test for homogeneity of variances. In general, a simple mean adjustment for the likelihood ratio test statistic $LR^{(N)}$ through multiplication by a constant of the form $1 + b/N$ implies $P^{(N)}[(1 + b/N)LR^{(N)} \leq x] = \text{pr}[\chi_f^2 \leq x] + o(N^{-1})$ under the null hypothesis, where f is the number of restrictions under test; see [Lawley \(1956\)](#) and [Hayakawa \(1977\)](#). This fact became widely known as the Bartlett correctability of $LR^{(N)}$, where substitution of a suitable consistent estimator for b is allowed. Among the vast literature, we further mention [Taniguchi \(1988\)](#), [Bickel and Ghosh \(1990\)](#), [Jensen \(1993\)](#) and [Kakizawa \(2011\)](#) for the theoretical issues.

Strangely, the test statistics $T^{(N)}$ other than $LR^{(N)}$ are, in general, not Bartlett correctable. Thus, [Chandra and Mukerjee \(1991\)](#) and [Taniguchi \(1991b\)](#) first proposed the Bartlett-type adjustments on the basis of additional information of score or the maximum likelihood estimator (MLE), respectively. [Cordeiro and Ferrari \(1991\)](#) gave a k th order polynomial in $T^{(N)}$ without a constant term, where $k \in \mathbf{N}$ and the coefficients are determined according to an asymptotic expansion for the null distribution of $T^{(N)}$; see [Kakizawa \(1996\)](#) for the corresponding monotone version. Note that for the case $f = k = 1$, these proposals ([Chandra and Mukerjee, 1991](#); [Cordeiro and Ferrari, 1991](#); [Taniguchi, 1991b](#)) are essentially identical to the traditional multiplicative Bartlett adjustment. However, several attempts ([Chandra and Mukerjee, 1991](#); [Taniguchi, 1991b](#)) and a unified treatment for the Chandra–Mukerjee/Taniguchi adjustments ([Kakizawa, 2010](#)) were restricted to the test of the simple null hypothesis. Extending [Mukerjee \(1992\)](#), who considered Rao's test statistic about a scalar parameter in the presence of a scalar nuisance parameter, [Kakizawa \(2012\)](#) has recently revisited the Bartlett-type adjustments for a class of

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test statistics in the framework of a composite hypothesis about a subvector of the parameters, where both the parameter of interest and the nuisance parameter are multidimensional.

Notice that the literature on the Cordeiro–Ferrari adjustment (CF adjustment for short) with $k = 2, 3$ is very extensive; some related papers during the last two decades are found in Kakizawa (2012). A contribution of this paper is to generalize the CF adjustment in a sense, by noting that it is not a unique form, as pointed out by Kakizawa (2011) for the traditional Bartlett adjustment $(1 + b/N)LR^{(N)}$. Although, in this paper, we focus on the independent and identically distributed case for notational simplicity, we can construct the Bartlett-type adjustments even in a non-identical or dependent case where some regularity conditions are met for the log-likelihood derivatives according to the situations under consideration. So, the results are applicable in wide generality since they allow both the interest and nuisance parameters to be vector-valued, for which there is no assumption regarding the global parameter orthogonality.

2. Preliminaries

2.1. Notation

Let $P_{\theta}^{(N)}$ denote the θ -distribution of $\mathbf{X}_1, \dots, \mathbf{X}_N$, which are independent and identically distributed random vectors, taking values of \mathbf{R}^{d_x} , according to a density $f(\mathbf{x}, \theta)$, $\theta \in \Theta$, where Θ is an open convex subset of \mathbf{R}^p . Throughout this paper, we assume the same regularity conditions as in Kakizawa (2012). The parameter vector $\theta = (\theta_1, \dots, \theta_p)'$ is composed of two parts, a parameter of interest $\theta_{(1)} = (\theta_1, \dots, \theta_{p_1})'$ and a nuisance parameter $\theta_{(2)} = (\theta_{p_1+1}, \dots, \theta_{p_1+p_2})'$ with $p = p_1 + p_2$, where $\theta = (\theta'_{(1)}, \theta'_{(2)})'$ and $\Theta = \Theta_{(1)} \times \Theta_{(2)}$. The log-likelihood is denoted by $\mathcal{L}^{(N)}(\theta) = \sum_{i=1}^N \log f(\mathbf{X}_i, \theta)$. We want to test a composite hypothesis $H_0 : \theta_{(1)} = \theta_{(1)0}$ against $H_1 : \theta_{(1)} \neq \theta_{(1)0}$, where $\theta_{(1)0} \in \Theta_{(1)}$ is specified while $\theta_{(2)} \in \Theta_{(2)}$ remains unspecified. Let $\tilde{\theta}_{ML}^{(N)} \in \Theta$ be the MLE of θ , and let $\tilde{\theta}_{(2)ML}^{(N)} \in \Theta_{(2)}$ be the MLE of $\theta_{(2)}$ under the constraint $\theta_{(1)} = \theta_{(1)0}$, where we write

$$\tilde{\theta}_{ML}^{(N)} = \begin{pmatrix} \theta_{(1)0} \\ \tilde{\theta}_{(2)ML}^{(N)} \end{pmatrix}.$$

As usual, the R th partial derivative of the log density $\log f(\mathbf{x}, \theta)$ with respect to θ is denoted by

$$\ell_{j_1 \dots j_R}(\mathbf{x}, \theta) = \frac{\partial}{\partial \theta_{j_1}} \dots \frac{\partial}{\partial \theta_{j_R}} \log f(\mathbf{x}, \theta) \quad (R \in \mathbf{N}; j_1, \dots, j_R \in \{1, \dots, p\}).$$

We introduce $I_R = j_1 \dots j_R$ for notational simplicity and denote the cumulants of the $\ell_{I_R}(\mathbf{X}, \theta)$'s by

$$v_{I_{R_1}, \dots, I_{R_v}}(\theta) = \text{cum}_{\theta}[\ell_{I_{R_1}}(\mathbf{X}, \theta), \dots, \ell_{I_{R_v}}(\mathbf{X}, \theta)],$$

where descending order $R_1 \geq \dots \geq R_v \geq 1$ on the size $R_i = |I_{R_i}|$ is assumed, since $v_{I_{R_1}, \dots, I_{R_v}}(\theta)$ is symmetric under permutation of $\{I_{R_1}, \dots, I_{R_v}\}$. We assume that

$$\begin{aligned} v_{j_1}(\theta) &= 0, & v_{j_1 j_2}(\theta) + v_{j_1 j_2}(\theta) &= 0, & v_{j_1 j_2 j_3}(\theta) + \langle 3 \rangle v_{j_1 j_2 j_3}(\theta) + v_{j_1 j_2 j_3}(\theta) &= 0, \\ v_{j_1 j_2 j_3 j_4}(\theta) + \langle 4 \rangle v_{j_1 j_2 j_3 j_4}(\theta) + \langle 3 \rangle v_{j_1 j_2 j_3 j_4}(\theta) + \langle 6 \rangle v_{j_1 j_2 j_3 j_4}(\theta) + v_{j_1 j_2 j_3 j_4}(\theta) &= 0 \end{aligned} \quad (1)$$

for all $\theta \in \Theta$, where $\langle n \rangle$ before a term with indices is a sum of n similar terms obtained by index permutation. We make use of the Bartlett identities (1) to eliminate $v_{j_1 j_2}(\theta)$, $v_{j_1 j_2 j_3}(\theta)$ and $v_{j_1 j_2 j_3 j_4}(\theta)$ in subsequent calculations. Also, according to the partition $\theta = (\theta'_{(1)}, \theta'_{(2)})'$, we stack the element

$$Z_j^{(N)}(\theta) = \frac{1}{N^{1/2}} \sum_{i=1}^N \ell_j(\mathbf{X}_i, \theta) \quad \text{and} \quad v_{j,k}(\theta) = -v_{jk}(\theta)$$

as follows:

$$[Z_j^{(N)}(\theta)]_{j=1, \dots, p} = \begin{pmatrix} \mathbf{Z}_{(1)}^{(N)}(\theta) \\ \mathbf{Z}_{(2)}^{(N)}(\theta) \end{pmatrix}, \quad [v_{j,k}(\theta)]_{j,k \in \{1, \dots, p\}} = \begin{pmatrix} \mathbf{v}_{(11)}(\theta) & \mathbf{v}_{(12)}(\theta) \\ \mathbf{v}_{(21)}(\theta) & \mathbf{v}_{(22)}(\theta) \end{pmatrix}.$$

They are the $p \times 1$ score vector $\mathbf{Z}^{(N)}(\theta)$ and the $p \times p$ Fisher information matrix $\mathbf{v}(\theta) = \text{var}_{\theta}^{(N)}[\mathbf{Z}^{(N)}(\theta)]$, respectively. It should be emphasized that we have no assumption regarding the so-called global parameter orthogonality, i.e., $\mathbf{v}_{(12)}(\theta) \equiv \mathbf{O}_{p_1, p_2}$, where \mathbf{O}_{p_1, p_2} is the $p_1 \times p_2$ zero matrix. Further, we write

$$Z_{j_1 \dots j_R}^{(N)}(\theta) = \frac{1}{N^{1/2}} \sum_{i=1}^N \{\ell_{j_1 \dots j_R}(\mathbf{X}_i, \theta) - v_{j_1 \dots j_R}(\theta)\} \quad (R = 2, 3, \dots).$$

Unless otherwise stated, we use the letters $\{j, k\}$ as indices of θ that run from 1 to p , the letters $\{a, b\}$ as indices of $\theta_{(1)}$ that run from 1 to p_1 and the letters $\{r, s\}$ as indices of $\theta_{(2)}$ that run from $p_1 + 1$ to p . We adopt the Einstein summation

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