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**LOCAL POWER OF THREE CLASSIC CRITERIA  
IN GENERALISED LINEAR MODELS  
WITH UNKNOWN DISPERSION**

*by*

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# Local power of three classic criteria in generalised linear models with unknown dispersion

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## SUMMARY

Cordeiro, Ferrari & Paula (1994) obtained asymptotic expansions for the nonnull distribution functions of the likelihood ratio, score and Wald test statistics in generalised linear models under Pitman alternatives assuming known dispersion. This paper generalises some of their results by relaxing this assumption.

*Some key words:* Asymptotic expansion; Chi-squared distribution; Composite hypothesis; Exponential family; Generalised linear models; Likelihood ratio test; Local power; Pitman alternatives; Score test; Wald test.

## 1. INTRODUCTION

Three commonly used test criteria are the likelihood ratio, score and Wald tests. In a recent *Biometrika* paper, Cordeiro, Botter & Ferrari (1994) obtained asymptotic expansions to order  $n^{-1/2}$ , where  $n$  is the sample size, for the distribution functions of these statistics in generalised linear models under Pitman alternatives. They considered tests on a subset of linear parameters, on the dispersion parameter, and on all of the linear parameters. For the test on a subset of linear parameters, which is usually the most useful for practical applications, the authors have assumed that the dispersion parameter is known. This assumption is quite restrictive since for gamma, normal and inverse Gaussian models the dispersion is usually unknown. In this paper, we relax the assumption of known dispersion, and obtain nonnull asymptotic expansions for the likelihood ratio, score and Wald tests in generalised linear models for the test on a subset of parameters of the linear predictor. Our results show that these expansions are not sensitive to the uncertainty involved in the estimation of the dispersion parameter to order  $n^{-1/2}$ .

## 2. BACKGROUND

Let  $y = (y_1, \dots, y_n)'$  be a set of independent but not necessarily identically distributed observations,  $l = l(\theta)$  be the log-likelihood function, where  $\theta$  is an  $m$ -vector of unknown parameters, and  $U$  and  $K$  be the score function and Fisher's information matrix, respectively. The parameter vector  $\theta$  is partitioned as  $\theta = (\theta_1', \theta_2')'$ , where the dimensions of  $\theta_1$  and  $\theta_2$  are  $q$  and  $m - q$ , respectively. The null hypothesis under test is  $H_0 : \theta_1 = \theta_1^{(0)}$  against a two-sided alternative, where  $\theta_1^{(0)}$  is a known  $q$ -vector, and  $\theta_2$  is thus a vector of nuisance parameters. The following partitions are induced by the partition of the parameter vector:

$$K = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix}, \quad K^{-1} = \begin{pmatrix} K^{11} & K^{12} \\ K^{21} & K^{22} \end{pmatrix}, \quad A = \begin{pmatrix} 0 & 0 \\ 0 & K_{22}^{-1} \end{pmatrix}.$$

We also define  $M = K^{-1} - A$ . The unrestricted and restricted maximum likelihood estimates of  $\theta$  are denoted by  $\hat{\theta} = (\hat{\theta}_1', \hat{\theta}_2')'$  and  $\tilde{\theta} = (\tilde{\theta}_1^{(0)'} , \tilde{\theta}_2')'$ , respectively. Functions evaluated at  $\hat{\theta}$  are distinguished by the addition of a '^' whereas functions evaluated at  $\tilde{\theta}$  are distinguished by the addition of a '~'. The likelihood ratio, Wald and score statistics are denoted by  $S_1$ ,  $S_2$  and  $S_3$ , respectively. Under mild regularity conditions, these three statistics have a limiting null  $\chi_q^2$  distribution, that is,  $\text{pr}(S_i \leq c) = \text{pr}(\chi_q^2 \leq c) + o(1)$ ,  $i = 1, 2, 3$ .

Consider now a sequence of Pitman alternative hypotheses given by  $H_n : \theta_1 = \theta_1^{(0)} + \xi$ , where  $\xi = (\xi_1, \dots, \xi_q)'$  with  $\xi_i = O(n^{-1/2})$ ,  $i = 1, \dots, q$ . In what follows, all suffices range from 1 to  $m$ . Let  $U_i = \partial l / \partial \theta_i$ ,  $U_{ij} = \partial^2 l / \partial \theta_i \partial \theta_j$ , etc. We shall use the following notation for cumulants of log-likelihood derivatives (Lawley, 1956):  $\kappa_{ij} = E(U_i U_j)$ ,  $\kappa_{ijk} = E(U_i U_j U_k)$ ,  $\kappa_{ij} = E(U_i U_j)$ ,  $\kappa_{ij,k} = E(U_{ij} U_k)$ ,  $\kappa_{ij,kr} = E(U_{ij} U_{kr}) - \kappa_{ij} \kappa_{kr}$ ,  $\kappa_{i,j,kr} = E(U_i U_j U_{kr}) - \kappa_{i,j} \kappa_{kr} - \kappa_{i,j,k,r} = E(U_i U_j U_k U_r) - \kappa_{i,j} \kappa_{k,r} - \kappa_{i,k} \kappa_{j,r} - \kappa_{i,r} \kappa_{j,k}$ . We have that the elements of  $K$  are  $\kappa_{i,j} = -\kappa_{ij}$ , and the corresponding elements of  $K^{-1}$  are  $\kappa^{ij} = -\kappa^{ij}$ . Next, define  $\delta$  as an  $m$ -vector given by

$$\delta = \begin{pmatrix} I_q \\ -K_{22}^{-1} K_{21} \end{pmatrix} \xi,$$

where  $I_q$  is the identity matrix of order  $q$ , and define the scalar parameter  $\lambda = \delta' K \delta$ .

Under a sequence of Pitman alternatives  $H_n : \theta_1 = \theta_1^{(0)} + \xi$  and regularity conditions which exclude lattice problems, the asymptotic expansions of the distribution functions of the three test statistics can be written as (Harris & Peers, 1980; Hayakawa, 1975)

$$\text{pr}(S_i \leq c) = G_{q,\lambda}(c) + \sum_{j=0}^3 b_{ij} G_{q+2j,\lambda}(c) + o(n^{-1/2}),$$

where  $G_{v,\lambda}(\cdot)$  is the distribution function of a noncentral chi-squared variate with  $v$  degrees of freedom and noncentrality parameter  $\lambda$ . The  $b$ 's are written as follows:  $b_{11} = (1/6) \sum_{i,j,k=1}^q \{(\kappa_{ijk} - 2\kappa_{i,j,k})\delta_i \delta_j \delta_k + 3(\kappa_{ijk} + 2\kappa_{i,j,k})a_{ij} \delta_k\} - (1/2) \sum_{i=1}^q \sum_{j,k=1}^q (\kappa_{ijk} + \kappa_{i,j,k}) \xi_i \delta_j \delta_k$ ,  $b_{12} = (1/6) \sum_{i,j,k=1}^q \kappa_{i,j,k} \delta_i \delta_j \delta_k$ ,  $b_{13} = 0$ ,  $b_{21} = (1/2) \sum_{i,j,k=1}^q \{(\kappa_{ijk} + 2\kappa_{i,j,k})\delta_i \delta_j \delta_k - 2\kappa_{i,j,k} m_{ij} \delta_k + (\kappa_{ijk} + 2\kappa_{i,j,k})\kappa^{ij} \delta_k\} - (1/2) \sum_{i=1}^q \sum_{j,k=1}^q (\kappa_{ijk} + \kappa_{i,j,k}) \xi_i \delta_j \delta_k$ ,  $b_{22} = -(1/2) \sum_{i,j,k=1}^q (\kappa_{i,j,k} \delta_i \delta_j \delta_k +$

$\kappa_{ijk}m_{ij}\delta_k$ ),  $b_{23} = -(1/6)\sum_{i,j,k=1}^p \kappa_{ijk}\delta_i\delta_j\delta_k$ ,  $b_{31} = (1/6)\sum_{i,j,k=1}^p \{(\kappa_{ijk} - 2\kappa_{i,j,k})\delta_i\delta_j\delta_k - 3\kappa_{i,j,k}m_{ij}\delta_k + 3(\kappa_{ijk} + 2\kappa_{i,j,k})a_{ij}\delta_k\} - (1/2)\sum_{i=1}^q \sum_{j,k=1}^p (\kappa_{ijk} + \kappa_{i,j,k})\xi_i\delta_j\delta_k$ ,  $b_{32} = (1/2)\sum_{i,j,k=1}^p \kappa_{i,j,k}m_{ij}\delta_k$ ,  $b_{33} = (1/6)\sum_{i,j,k=1}^p \kappa_{i,j,k}\delta_i\delta_j\delta_k$ . Also,  $b_{i0} = -(b_{i1} + b_{i2} + b_{i3})$ ,  $i = 1, 2, 3$ . All quantities, except for  $\xi$ , are evaluated under  $H_0$ .

### 3. ORTHOGONAL PARAMETERS

Suppose  $\theta = (\beta'_1, \beta'_2, \phi)'$  is our unknown parameter vector, where  $\phi$  is a scalar parameter which is globally orthogonal to  $\beta = (\beta'_1, \beta'_2)'$  in the sense of Cox & Reid (1987), with  $\beta_1 = (\beta_1, \dots, \beta_q)'$  and  $\beta_2 = (\beta_{q+1}, \dots, \beta_p)'$ . Suppose also that the null hypothesis under test is  $H_0: \beta_1 = \beta_1^{(0)}$ , where  $\beta_1^{(0)}$  is a known  $q$ -vector. We can then write the information matrix as  $K = \text{diag}\{K_\beta, \kappa_{\phi,\phi}\}$ , where  $K_\beta = E\{(\partial l/\partial\beta)(\partial l/\partial\beta)'\}$ ,  $\kappa_{\phi,\phi} = -E(\partial^2 l/\partial\phi^2)$  and  $l = l(\beta, \phi)$  is the log-likelihood function.  $K_\beta$  can be partitioned following the partition of the  $\beta$  vector as

$$K_\beta = \begin{pmatrix} K_{\beta 11} & K_{\beta 12} \\ K_{\beta 21} & K_{\beta 22} \end{pmatrix}.$$

We also have that

$$A = \begin{pmatrix} A_\beta & 0 \\ 0 & \kappa_{\phi,\phi}^{-1} \end{pmatrix} \quad \text{and} \quad M = \begin{pmatrix} M_\beta & 0 \\ 0 & 0 \end{pmatrix}, \quad \text{where} \quad A_\beta = \begin{pmatrix} 0 & 0 \\ 0 & K_{\beta 22}^{-1} \end{pmatrix}$$

and  $M_\beta = K_\beta^{-1} - A_\beta$ . It then follows that  $m_{r\phi} = m_{\phi r} = m_{\phi\phi} = 0$ ,  $a_{r\phi} = a_{\phi r} = 0$ , for  $r = 1, \dots, p$ , and  $a_{\phi\phi} = \kappa_{\phi,\phi}^{-1}$ , where  $m_{\phi\phi}$  and  $a_{\phi\phi}$  are the  $(p+1, p+1)$ th elements of the matrices  $M$  and  $A$ , respectively, and  $m_{r\phi}$  and  $a_{r\phi}$  are the  $(r, p+1)$ th elements of the matrices  $M$  and  $A$ , respectively.

Given the results above, it is possible to show that

$$\delta = \begin{pmatrix} \delta_\beta \\ 0 \end{pmatrix} \xi, \quad \text{where} \quad \delta_\beta = \begin{pmatrix} I_q \\ -K_{\beta 22}^{-1}K_{\beta 21} \end{pmatrix},$$

and  $\lambda = \delta'_\beta K_\beta \delta_\beta$ . That is, the noncentrality parameter  $\lambda$  is the same, regardless of whether or not  $\phi$  is known. It is also possible to show using the general expressions for the  $b$ 's in the previous section that  $b_{ij} = b_{ij,\beta} + b_{ij,\beta\phi}$ , where  $b_{ij,\beta}$  equals the term  $b_{ij}$  for the case where  $\phi$  is known. That is, the  $b$ 's can be additively decomposed into their counterparts for the case where  $\phi$  is known plus some extra terms that account for the uncertainty involved in the estimation of the parameter  $\phi$ . It is also possible to show that  $b_{ij,\beta\phi} = 0$  for  $i = 1, 2, 3$  and  $j = 2, 3$  and

$$b_{i1,\beta\phi} = \frac{1}{2} \sum_{k=1}^p (\kappa_{\phi\phi k} + 2\kappa_{\phi,\phi k}) \kappa_{\phi,\phi}^{-1} \delta_k \quad (1)$$

for  $i = 1, 2, 3$ , where  $\kappa_{\phi\phi k} = E(\partial^3 l/\partial^2 \phi \partial \beta_k)$  and  $\kappa_{\phi,\phi k} = E\{(\partial l/\partial \phi)(\partial^2 l/\partial \phi \partial \beta_k)\}$ .

#### 4. NONNULL ASYMPTOTIC EXPANSIONS IN GENERALISED LINEAR MODELS

Let each observation  $y_i$ ,  $i = 1, \dots, n$ , be the realization of a continuous random variable  $Y_i$  with density function

$$\pi(y; \theta, \phi) = \exp\{\phi\{y\theta_i - b(\theta_i)\} + a(y, \phi)\}, \quad (2)$$

where  $a(\cdot, \cdot)$ ,  $b(\cdot)$  and  $c(\cdot)$  are known functions and  $\theta_i$  and  $\phi$  are possibly unknown parameters. Here,  $E(Y_i) = \mu_i = b'(\theta_i)$  and  $\text{var}(Y_i) = V_i/\phi$ , where  $\phi^{-1}$  is the dispersion parameter,  $V = V(\mu) = d\mu/d\theta$  is the variance function, and  $\theta = \int (1/V) d\mu = q(\mu)$  is a strictly monotonic function of the mean. The linear predictor is given by  $\eta = \sum_{j=1}^p \beta_j x_j = X\beta$ , where  $X$  is an  $n \times p$  ( $p \leq n$ ) matrix with rank  $p$  and  $\beta$  is a  $p$ -vector of unknown parameters. A generalised linear model is defined by a distribution in (2) and by a strictly monotonic, twice-differentiable link function  $d(\mu) = \eta$  relating the mean to the linear predictor.

Next, we partition the  $p$ -vector  $\beta$  as in the previous section, that is,  $\beta = (\beta_1', \beta_2')'$ , where  $\beta_1 = (\beta_1, \dots, \beta_q)'$  and  $\beta_2 = (\beta_{q+1}, \dots, \beta_p)'$  for  $q \leq p$ , which induces the partition of the model matrix as  $X = (X_1 \ X_2)$ . The null hypothesis is  $H_0 : \beta_1 = \beta_1^{(0)}$  and the alternative  $H_n : \beta_1 = \beta_1^{(0)} + \xi$ , with  $\xi_i = O(n^{-1/2})$ , as in Section 3. The likelihood ratio, Wald and score statistics can be written as  $S_1 = 2\{l(\hat{\beta}, \hat{\phi}) - l(\hat{\beta}_0, \hat{\phi}_0)\}$ ,  $S_2 = (\hat{\beta}_1 - \beta_1^{(0)})'(K^{(11)})^{-1}(\hat{\beta}_1 - \beta_1^{(0)})$  and  $S_3 = \hat{U}' \hat{K}^{11} \hat{U}_1$ , respectively, where  $\hat{\phi}$  is the restricted maximum likelihood estimate of the unknown precision parameter  $\phi$ .

For two-parameter full exponential family distributions with canonical parameters  $\phi$  and  $\phi\theta$ , the term  $a(y, \phi)$  in (2) can be written as  $a(y, \phi) = d_1(\phi) + d_2(y)$ . It then follows that in generalised linear models  $\kappa_{\phi\phi k} = \kappa_{\phi, \phi k} = 0$ , and hence equation (1) reduces to  $b_{i1, \beta\phi} = 0$ ,  $i = 1, 2, 3$ . This implies that the  $b$ 's are given by  $b_{11} = (1/2) \sum [\phi\{(f-g)_1 e_1 t_i^2 - f_1 t_i^2\} - f_1 t_1 z_{21}]$ ,  $b_{12} = (\phi/6) \sum (f-g)_1 t_i^3$ ,  $b_{13} = 0$ ,  $b_{21} = (1/2) \sum [\phi\{(f+g)_1 e_1 t_i^2 - f_1 t_i^2\} - 2g_1 t_1 (z_{11} - z_{21}) - f_1 t_1 z_{21}]$ ,  $b_{22} = (1/2) \sum \{-\phi g_1 t_i^3 + (f+2g)_1 t_1 (z_{11} - z_{21})\}$ ,  $b_{23} = (\phi/6) \sum (f+2g)_1 t_i^3$ ,  $b_{31} = (1/2) \sum [\phi\{(f+g)_1 e_1 t_i^2 - f_1 t_i^2\} - (f-g)_1 t_1 (z_{11} - z_{21}) - f_1 t_1 z_{21}]$ ,  $b_{32} = (1/2) \sum (f-g)_1 t_1 (z_{11} - z_{21})$  and  $b_{33} = (\phi/6) \sum (f-g)_1 t_i^3$ , where summations are over the  $n$  observations and  $Z = \{z_{lm}\} = X(X'WX)^{-1}X'$ ,  $Z_2 = \{z_{2lm}\} = X_2(X_2'WX_2)^{-1}X_2'$ ,  $W = \text{diag}\{w_1, \dots, w_n\}$  with  $w_1 = V_1^{-1}(d\mu_1/d\eta)^2$ ,  $t = (t_1, \dots, t_n)' = X\delta$ ,  $e = (e_1, \dots, e_n) = X_1\delta$ ,

$$f = \frac{1}{V} \frac{d\mu}{d\eta} \frac{d^2\mu}{d\eta^2} \quad \text{and} \quad g = \frac{1}{V} \frac{d\mu}{d\eta} \frac{d^2\mu}{d\eta^2} - \frac{1}{V^2} \frac{dV}{d\mu} \left( \frac{d\mu}{d\eta} \right)^3.$$

Also, it can be shown that the noncentrality parameter is  $\lambda = \phi \sum t_i^2 w_i$ . That is, we obtain the same  $b$ 's and  $\lambda$  as did Cordeiro, Botter and Ferrari (1994) for the case of known dispersion.

Some previous results have shown that the uncertainty involved in the estimation of the dispersion parameter does change the null distributions of the likelihood ratio and score statistics in generalised linear models to order  $n^{-1}$ ; see Cordeiro (1987) and Cribari-Neto & Ferrari (1995), respectively. However, the results above show that the nonnull distributions of the likelihood ratio, Wald and score statistics to order  $n^{-1/2}$  in generalised linear models do not change when one introduces unknown dispersion.

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