ON THE FORMULATION OF WALD TESTS OF NONLINEAR RESTRICTIONS

BY P. C. B. PHILLIPS AND JOON Y. PARK

This paper utilizes asymptotic expansions of the Edgeworth type to investigate alternative forms of the Wald test of nonlinear restrictions. Some formulae for the asymptotic expansion of the distribution of the Wald statistic are provided for a general case that should include most econometric applications. When specialized to the simple cases that have been studied recently in the literature, these formulae are found to explain rather well the discrepancies in sampling behavior that have been observed by other authors. It is further shown how the corrections delivered by Edgeworth expansions may be used to find transformations of the restrictions which accelerate convergence to the asymptotic distribution.

KEYWORDS: Accelerated convergence, asymptotic series, Edgeworth expansions, tensor product representations, Wald tests.

1. INTRODUCTION

The numerical value of the Wald test depends not only on the restriction to be tested but also on its algebraic formulation. Under general conditions Wald statistics which are based upon different but algebraically equivalent forms all have the same asymptotic distribution under the null hypothesis that the restriction holds. However, numerical outcomes of the tests and their finite sample distributions can be substantially different for different forms of the same restrictions. Thus, the adequacy of the usual asymptotic $\chi^2$ approximation may also vary substantially as we change the algebraic form of the restriction.

Gregory and Veal (1985) recently studied this phenomenon by simulation. They observed that the distributions of alternative Wald statistics for a simple nonlinear restriction can be widely divergent in small samples; and they concluded that the algebraic form of the restrictions to be tested is likely to be important in many different empirical applications of the Wald test. Lafontaine and White (1986) go further and argue that it is even possible to obtain any value of the Wald statistic by suitably reformulating the restriction. Breusch and Schmidt (1985) make a similar point in a closely related paper.

The purpose of the present paper is to study this phenomenon by direct analytical methods. Our approach is to develop an Edgeworth expansion of the distribution of the Wald statistic in a form that is sufficiently general to permit different formulations of the restrictions. Higher order terms in the expansion then provide a mechanism by which deviations from the common asymptotic theory may be measured for alternative forms of the Wald test. In this sense the asymptotic expansion provides more complete distributional information than crude asymptotic theory and, in general, leads to distinct higher order terms for

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different algebraic representations of the restrictions. As is common in finite sample theory, these terms involve parametric dependencies, so that the adequacy of the first order asymptotics in each case will depend on the relevant region of the parameter space. This information can be used to improve decision making since it provides advice concerning the behavior of alternative forms of the Wald test in different regions of the parameter space.

Our results indicate that terms in the asymptotic expansion up to $O(T^{-1})$ where $T$ is the sample size provide enough additional information to capture the main distributional effects that are incurred by using alternative forms of the Wald test. The correction terms may be used to determine which version of a Wald test has a sampling distribution that is more closely approximated by the asymptotic. In simple cases the correction terms themselves suggest transformations of the restrictions which will improve the asymptotic approximation by eliminating the correction to a certain order and thereby accelerate convergence to the asymptotic distribution. An example is provided in Section 3. This idea is inspired by earlier work by Phillips (1979) and Konishi (1981) on the use of Edgeworth expansions to determine the form of normalizing transformations.

We emphasize that these conclusions are reached without reliance on the numerical quality of Edgeworth corrections. Edgeworth expansions by no means always improve the quality of first order asymptotic approximations. Indeed, as documented in other work (see Phillips (1977a, 1984)), they are often of very uneven quality and their performance is always parameter dependent. It might be said that Edgeworth expansions behave like the little girl with the curl in the famous nursery rhyme: when they are good they are very, very good and when they are bad they are horrid. Moreover, as shown in Phillips (1984), the corrections tend to work well when the error on the crude asymptotic is small (when they are least needed) and are poor when that error is large (when they are most needed). Nevertheless, even though their numerical quality is often unreliable, Edgeworth corrections still provide a valuable source of information about the adequacy of asymptotic theory. This is because they clearly signal those regions of the parameter space where the corrections are small and those where the corrections will be large. These, in turn, signal the regions where the crude asymptotic does well and those where it does poorly. This is precisely the type of information we seek in the present paper.

Some comments on related work and on our new algebraic approach are in order. Asymptotic expansions of the Edgeworth type have been extensively studied in the recent statistical and econometric literature. With respect to statistical criteria that are asymptotically chi-squared, this work has dealt both with formal expansions (Peers (1971), Hayakawa (1975, 1977), Harris and Peers (1980), Rothenberg (1984)) and with the theory of validity of these expansions (Chandra and Ghosh (1979, 1980), Sargan (1980), Mauléon (1981)). While much of the statistical literature (e.g. Hayakawa (1977) and Chandra and Ghosh (1979, 1980)) has emphasized the case of statistics which depend on underlying independent and identically distributed variates, some papers (e.g. Tunguchi (1985)) have extended the formal calculations to certain time series settings.
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Work in econometrics by Sargan (1976), Phillips (1977b) and Mauléon (1981) and Sargan and Satchell (1986) made general extensions of the validity theory to encompass time series applications. Our approach in the present paper is to provide formal calculations of the asymptotic expansions which in the general case have a similar range of applications.

The algebraic method that we adopt is different from that of earlier work, which has almost invariably used tensor notation. The latter is certainly the simplest and most economical for expansions to an arbitrary order. However, in the case of expansions to $O(T^{-1})$ many of the formulae can be derived and represented in matrix form. Our approach illustrates this new format which, in our view, helps to simplify the algebraic structure of Edgeworth expansions. En route, we provide some new general formulae for the matrix of sixth moments of the multivariate normal distribution.

The paper is organized as follows. Section 2 develops formulae for the asymptotic expansion of the distribution of the Wald test in a general setting. These formulae are specialized in Section 3 to study the examples given recently by other authors and to examine transformations which accelerate convergence. Some brief conclusions are given in Section 4. Proofs are provided in Appendix A. Appendix B details some useful additional formulae for the Edgeworth expansion of the Wald test which should be general enough to cover most major econometric applications.

2. EDGEWORTH EXPANSION IN A GENERAL CASE

We start by assuming that the data generating mechanism depends on a set of parameters represented by the $p \times 1$ vector $\beta$ whose true value we denote by $\beta^0$. The hypothesis to be tested takes the form

$$(1) \quad H_0: g(\beta^0) = 0$$

where $g: R^p \rightarrow R^r$, $r \leq p$, is a vector valued function that is continuously differentiable at least to the third order. Given a sample of size $T$, let $\hat{\beta}$ denote an estimator of $\beta^0$ and define $q = \hat{\beta} - \beta^0$ and $\tilde{q} = \sqrt{T}q$.

Under general conditions we can expect that

$$(2) \quad \tilde{q} \Rightarrow N(0, \Omega)$$

as $T \uparrow \infty$, where the symbol "$\Rightarrow$" represents weak convergence of the associated probability measures and $\Omega$ is some positive semi-definite matrix, possibly dependent on $\beta$.

The vector $\tilde{q}$ may be regarded as being composed of low order sample moments of the underlying data or simple functions (often rational functions, as in the case of instrumental variables estimation) of such sample moments. As shown in Phillips (1977) and Sargan and Satchell (1986) the distribution of $\tilde{q}$ will admit a valid asymptotic expansion under very general conditions. Details of conditions which are sufficient to ensure the existence of a valid expansion are given in these articles and a constructive process by which the expansion may be
obtained is detailed in Phillips (1982). The asymptotic expansion of the density of \( \tilde{q} \) has the form:

\[
\text{pdf}(\tilde{q}) = (2\pi)^{-p/2}(\det \Omega)^{-1/2} \exp \left\{ -\left( \frac{1}{2} \tilde{q}'\Omega^{-1}\tilde{q} \right) \right. \\
\times \left. \left[ 1 + \sum_{j=1}^{r-1} P_j(\tilde{q}) T^{-j/2} \right] + O(T^{-(r+1)/2}) \right\}
\]

where \( P_j(\tilde{q}) \) is a real polynomial in the elements of \( \tilde{q} \) of degree \( 3j \) and where we have assumed that \( \Omega \) is positive definite. The coefficients in the polynomials \( P_j \) in (3) are determined by the cumulants of the underlying sample moments upon which \( \tilde{q} \) depends and the derivatives of the functions which define that dependence, both to a sufficiently high order that is determined in turn by the order, \( r \), of the expansion. \( P_j \) is an odd (respectively, even) function when the index \( j \) is odd (even).

For the main purpose of this paper we shall be working at a sufficient level of generality if we require that standardizing transformations have been carried out which ensure that \( \Omega = I_p \) and that, instead of (3), we have quite simply:

\[
\tilde{q} = N(0, I)
\]

where the symbol \( \equiv \) signifies equality in distribution. Our results may be extended to the fully general case of (3) by transformation and by carrying the additional terms induced by (3) in our subsequent computations of the asymptotic expansion of the Wald test. Since the calculations are much heavier in this case, the formulae that apply are derived in Appendix B. They should be useful in many additional contexts.

We now define \( G = \partial g/\partial \beta' \) and use carets over such functions of \( \beta \) to signify their evaluation at \( \hat{\beta} \), so that \( \hat{G} = G(\hat{\beta}) \) whereas \( G = G(\beta^0) \). In what follows we assume that \( G \) (respectively, \( \hat{G} \)) is of full rank \( r \) (with probability one).

The Wald statistic for testing (1) now has the form

\[
W = T \hat{g}' (\hat{G}\hat{G}')^{-1} \hat{g}
\]

and as \( T \to \infty \)

\[
W \to \chi^2_r
\]

under the stated conditions. To refine (6) we first develop the following Taylor representation of \( \sqrt{T} \hat{g} \) and \( \hat{\psi}^{-1} = (\hat{G}\hat{G}')^{-1} = (\hat{\psi}^{ia}) \). Thus, to \( O_p(T^{-1}) \) we have (using the tensor summation convention of a repeated suffix):

\[
\sqrt{T} \hat{g}_i = g_{ij} \hat{q}_j + \frac{1}{2\sqrt{T}} g_{ijk} \hat{q}_j \hat{q}_k + \frac{1}{6T} g_{ijklm} \hat{q}_j \hat{q}_k \hat{q}_l \hat{q}_m + O_p(T^{-3/2})
\]

\[
\hat{\psi}^{ia} = \psi^{ia} + \frac{1}{\sqrt{T}} \psi^{ia}_{m} \hat{q}_m + \frac{1}{2T} \psi^{ia}_{mn} \hat{q}_m \hat{q}_n + O_p(T^{-3/2}).
\]

Here subcripts \( j, k, l \) of \( g \) and \( m, n \) of \( \psi \) imply differentiation with respect to the corresponding components of \( \hat{\beta} \). Using these expansions in (5) and collecting
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terms up to $O_p(T^{-1})$ we obtain:

$$W = \left( g_{ij} \psi^{ia} g_{ab} \right) \bar{q}_j \bar{q}_b + \frac{1}{\sqrt{T}} \left( g_{ij} \psi^{ia} g_{ab} + g_{ijm} \psi^{ia} g_{ab} \right) \bar{q}_j \bar{q}_b \bar{q}_m$$

$$+ \frac{1}{T} \left[ \left( \frac{1}{2} \right) g_{ij} \psi^{ia} g_{ab} + g_{ijm} \psi^{ia} g_{ab} \right] \bar{q}_j \bar{q}_b \bar{q}_m$$

$$+ \frac{1}{T} \left( \frac{1}{2} g_{ij} \psi^{ia} g_{ab} + \left( \frac{1}{2} \right) g_{ijm} \psi^{ia} g_{ab} \right) \bar{q}_j \bar{q}_b \bar{q}_m$$

$$+ O_p(T^{-3/2})$$

(7)

$$= \bar{q}' G'(GG')^{-1} G \bar{q} + T^{-1/2} u(\bar{q}) + T^{-1} v(\bar{q}) + O_p(T^{-3/2}), \quad \text{say.}$$

We rewrite $u(\bar{q})$ and $v(\bar{q})$ as:

$$u(\bar{q}) = \text{vec}(J)'(\bar{q} \otimes \bar{q} \otimes \bar{q}) ,$$

$$v(\bar{q}) = \text{tr}\left\{ L(\bar{q} \bar{q}' \otimes \bar{q} \bar{q}') \right\} ,$$

where $J$ is the $p^2 \times p$ matrix which stacks the $p \times p$ matrices

(8) $\quad G'(GG')^{-1} G + G'(GG')^{-1} G$ \quad (i = 1, \ldots, p),

$\text{vec}(J)$ stacks rows of $J$, and $L$ is the $p^2 \times p^2$ matrix whose $(i, j)$th $p \times p$ block is given by

(9) $\quad \left( \frac{1}{2} \right) G'(GG')^{-1} G + G'(GG')^{-1} G + \left( \frac{1}{2} \right) G'(GG')^{-1} G,$

$$+ \left( \frac{1}{2} \right) G'(GG')^{-1} G \quad (i, j = 1, \ldots, p).$$

In (8) and (9) we use the notation $A_{(i)} = \partial A/\partial \beta_i$, $A_{(ij)} = \partial^2 A/\partial \beta_i \partial \beta_j$ for any matrix $A = A(\beta)$. Note that (7) is a convenient matrix representation of the stochastic expansion of $W$ in terms of the component variates $\bar{q}$.

The characteristic function of $W$ may now be written:

(10) $\quad \text{cf}_W(t) = (2\pi)^{-p/2} \int_{R^p} e^{i \omega \bar{q}} e^{-\frac{1}{2} \bar{q}' G^{-1} \bar{q}} d\bar{q}$

$$= (2\pi)^{-p/2} \int_{R^p} \exp \left\{ -\left( \frac{1}{2} \right) \bar{q}' \left[ I - 2itG'(GG')^{-1} G \right] \bar{q} \right\}$$

$$\times \left[ 1 + \frac{i \omega}{T} u(\bar{q}) + \frac{i \omega}{T} v(\bar{q}) - \frac{T^2}{2T} u^2(\bar{q}) \right] d\bar{q} + o(T^{-1}).$$

To compute the integral in (10) we first transform $\bar{q} \rightarrow z = R^{-1} \bar{q}$ where we define $R$ by writing

$$S = I - 2itP_G = \bar{P}_G + (1 - 2it)P_G'$$

and

$$R = \bar{P}_G + (1 - 2it)^{-1/2}P_G = S^{-1/2}.$$ 

Here we use the notations $P_A = A(A'A)^{-1}A'$ and $\bar{P}_A = I - P_A$ for any matrix $A$ of full column rank.
Noting that the Jacobian of the transformation $\bar{q} \rightarrow z$ is $\det R = (1 - 2it)^{-r/2}$ we find that (10) reduces to:

$$c_f_r(t) = (1 - 2it)^{-r/2} \left( 1 + (it)T^{-1/2}E(u) + (ut)T^{-1}E(v) - \left( \frac{t^2}{2} \right) T^{-1}E(u^2) \right) + o(T^{-1})$$

where $E$ is the integral expectation operator with respect to the density $(2\pi)^{-p/2} \exp\{-\frac{1}{2}z'z\}$ and

$$u = u(Rx), \quad v = v(Rx).$$

Note that the leading term in (11) is just the characteristic function of the $\chi^2_r$ distribution, corresponding to the usual first order asymptotics of $W$. Higher order terms in (11) are computed simply by taking expectations of polynomials in independent standard normal variates. We observe that $E(u) = 0$ since $u$ involves only a polynomial of odd degree in $z$. Moreover,

$$v = \text{tr} \left( (R \otimes R) L (R \otimes R) (zz' \otimes zz') \right)$$

and

$$u^2 = \text{vec} (J)'(R \otimes R \otimes R)(zz' \otimes zz' \otimes zz')(R \otimes R \otimes R) \text{vec} (J).$$

Expectations of $v$ and $u^2$ are now obtained through the following Lemmas which provide convenient representations of the matrices of fourth and sixth moments of the multivariate normal distribution. We use $K_{mn}$ to denote the commutation matrix of order $mn \times mn$, i.e. the matrix for which $\text{vec} A' = K_{mn} \text{vec} A$ where $A$ is any $m \times n$ matrix.

**Lemma 2.1:** If $z = N(0, I_p)$ then

13. $E(zz' \otimes zz') = I + K_{pp} + (\text{vec} I) (\text{vec} I)'$;

14. $E(zz' \otimes zz \otimes zz')$ = $I \otimes I \otimes I$

\[+ \left( \frac{1}{2} \right) \sum_{i, j} \left[ (I \otimes T_{ij} \otimes T_{ij}) + (T_{ij} \otimes I \otimes T_{ij}) + (T_{ij} \otimes T_{ij} \otimes I) \right] \]

\[+ \sum_{i, j, k} (T_{ij} \otimes T_{ik} \otimes T_{jk}). \]

Alternatively,

15. $E(zz' \otimes zz' \otimes zz')$ = $I + K_{pp} + K_{pp}^2$

\[+ (I + K_{pp} + K_{pp}^2)(I \otimes (\text{vec} I)(\text{vec} I)')(I + K_{pp} + K_{pp}^2) \]

\[+ I \otimes K_{pp} + K_{pp} \otimes I + K_{pp}^2 \left( I \otimes K_{pp} \right) K_{pp}^2, \]

where

$$T_{ij} = E_{ij} + E_{ji}, \quad E_{ij} = e_i e_j',$$

and $e_i$ is the $i$'th unit vector in $\mathbb{R}^p$. 
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LEMMA 2.2: If \( x = N_p(0, V) \) then

\[
E(xx' \otimes xx') = (I + K_{pp})(V \otimes V) + (\text{vec} V')(\text{vec} V),
\]

\[
E(xx' \otimes xx') = (I + K_{pp} + K_{pp'}) (V \otimes V \otimes V)
+ (I + K_{pp} + K_{pp'})(V \otimes (\text{vec} V')(\text{vec} V)'(I + K_{pp} + K_{pp'})
+ V \otimes \{ K_{pp}(V \otimes V) \} \otimes V
+ K_{pp'}[V \otimes \{ K_{pp}(V \otimes V) \}] K_{pp'}.
\]

LEMMA 2.3:

\[
E(v) = a_0 + a_1(1 - 2it)^{-1} + a_2(1 - 2it)^{-2},
\]

\[
E(u^2) = b_1(1 - 2it)^{-1} + b_2(1 - 2it)^{-2} + b_3(1 - 2it)^{-3},
\]

where

\[
a_i = \text{tr}(A_i), \quad (i = 0, 1, 2),
\]

\[
A_0 = L \left\{ (I + K_{pp})(\vec{P}_G \otimes \vec{P}_G) + (\text{vec} \vec{P}_G)(\text{vec} \vec{P}_G) \right\},
\]

\[
A_1 = L \left\{ (I + K_{pp}) \left( (\vec{P}_G \otimes p_G + p_G \otimes \vec{P}_G) \right) + (\text{vec} \vec{P}_G)(\text{vec} p_G)' \right. \left. + (\text{vec} p_G)(\text{vec} \vec{P}_G)' \right\},
\]

\[
A_2 = L \left\{ (I + K_{pp})(p_G \otimes p_G) + (\text{vec} p_G)(\text{vec} p_G) \right\},
\]

\[
b_i = (\text{vec} J)'B_i(\text{vec} J), \quad (i = 1, 2, 3),
\]

\[
B_0 = H(\vec{P}_G \otimes \vec{P}_G \otimes \vec{P}_G) + H(\text{vec} \vec{P}_G)(\text{vec} \vec{P}_G) H
\]

\[
+ \vec{P}_G \otimes K_{pp}(\vec{P}_G \otimes \vec{P}_G) + K_{pp}(\vec{P}_G \otimes \vec{P}_G) \otimes \vec{P}_G
\]

\[
+ K_{pp'}(\vec{P}_G \otimes K_{pp}(\vec{P}_G \otimes \vec{P}_G)) K_{pp'} = C_0(\vec{P}_G), \quad \text{say},
\]

\[
B_1 = H \{ p_G \otimes (\text{vec} p_G)' + (\text{vec} p_G) \otimes \text{vec} \vec{P}_G \}' H
\]

\[
+ p_G \otimes K_{pp}(\vec{P}_G \otimes \vec{P}_G) + \vec{P}_G \otimes K_{pp}(p_G \otimes \vec{P}_G)
\]

\[
+ \vec{P}_G \otimes K_{pp}(\vec{P}_G \otimes p_G) + K_{pp}(p_G \otimes \vec{P}_G) \otimes \vec{P}_G
\]

\[
+ K_{pp'}(\{ p_G \otimes K_{pp}(\vec{P}_G \otimes \vec{P}_G) \} + \{ \vec{P}_G \otimes K_{pp}(p_G \otimes \vec{P}_G) \})
\]

\[
+ \{ \vec{P}_G \otimes K_{pp}(\vec{P}_G \otimes p_G) \} K_{pp'} = C_1(\vec{P}_G, p_G), \quad \text{say},
\]

\[
B_2 = C_1(\vec{P}_G, \vec{P}_G),
\]

\[
B_2 = C_0(p_G),
\]
and where
\[ H = I + K_{p'p} + K_{p'p}. \]

Using (18) and (19) in (11) we find the following expression for the characteristic function of \( W \):

\[
\begin{align*}
\varphi_W(t) &= (1 - 2it)^{-r/2} \left[ 1 + \frac{1}{T} \left( (a_0 - \frac{1}{4}b_1)(it) 
+ (a_1 + \frac{1}{4}b_1 - \frac{1}{2}b_2)it(1 - 2it)^{-1} 
+ (a_2 + \frac{1}{4}b_2 - \frac{1}{2}b_3)it(1 - 2it)^{-2} 
+ \frac{1}{2}b_3it(1 - 2it)^{-3} \right) \right] + o(T^{-1}).
\end{align*}
\]

Upon inversion of (20) we obtain asymptotic expansions of the density and distribution function of \( W \) up to \( O(T^{-1}) \). The final result is given in the following theorem.

**Theorem 2.4:** The asymptotic expansion of the distribution function of \( W \) up to \( O(T^{-1}) \) as \( T \uparrow \infty \) is given by:

\[
\begin{align*}
\text{cdf}(w) &= F_r(w) - \frac{1}{T} c(w) f_r(w) + o(T^{-1}) \\
&= F_r(w - T^{-1}c(w)) + o(T^{-1})
\end{align*}
\]

where \( f_r \) and \( F_r \) denote the density and distribution function, respectively, of a \( \chi^2_r \) variate and where

\[
\begin{align*}
c(w) &= \sum_{n=0}^{3} \alpha_n w^n
\end{align*}
\]

with
\[
\begin{align*}
\alpha_0 &= (4a_0 - b_1)/4, \\
\alpha_1 &= (4a_1 + b_1 - b_2)/4r, \\
\alpha_2 &= (4a_2 + b_2 - b_3)/4r(r + 2), \\
\alpha_3 &= b_3/4r(r + 2)(r + 4).
\end{align*}
\]

Finally, we observe that if \( w_\alpha^* \) is the critical value of the \( \chi^2_r \) distribution at the level \( \alpha \), then the corresponding critical value of \( \text{cdf}(w) \) correct to \( O(T^{-1}) \) is given by the solution of

\[
w_\alpha^* - T^{-1}c(w_\alpha^*) = w_\alpha.
\]

This may be approximated by

\[
w_\alpha^* = w_\alpha + T^{-1}c(w_\alpha)
\]

which is correct to the same accuracy of \( O(T^{-1}) \).
3 Specializations

We shall examine various examples and Monte Carlo results that have appeared in the recent literature in the light of the previous section.

(i) Gregory and Veall (1985)

These authors study by simulation methods the behavior of alternative Wald tests of the algebraically equivalent restrictions:

(I) $\beta_1 - 1/\beta_2 = 0,$

(II) $\beta_1 \beta_2 - 1 = 0,$

where $\beta_1$ and $\beta_2$ are coefficients in a classical linear regression model. Their evidence suggests that the Wald test based on formulation (I) performs poorly in finite samples, even in samples as large as $T = 500$, when $\beta_2$ is in the vicinity of the origin (more specifically, $\beta_2 = 0.2, 0.1$ in their experiments).

Asymptotic expansions of the distributions of these alternative Wald tests may be deduced from Theorem 2.4 by applying the formulae of the previous section. In particular, the coefficients of the correction factor $c(w)$ in (22) are displayed in Table I for easy comparison.

The first two coefficients $\alpha_0$ and $\alpha_1$ agree under the null hypothesis. Discrepancies to $O(T^{-1})$ occur in $\alpha_2$ and $\alpha_3$. For small $\beta_2$ both $\alpha_2$ and $\alpha_3$ are of $O(\beta_2^{-2})$ under (I); but, under (II), $\alpha_2 = O(\beta_2^2)$ and $\alpha_3 = O(\beta_2^2)$ as $\beta_2 \to 0$. Thus, substantial deviations from the nominal asymptotic size are to be expected for the Wald test under (I), whereas rather good approximations from first order asymptotics are to be expected under (II), at least in this region of the parameter space. For example, when $\beta_2 = 0.1$ we find that:

$\alpha_2 = -400, \quad \alpha_3 = 100 \quad$ under (I)

while

$\alpha_2 = -0.01, \quad \alpha_3 = 10^{-6} \quad$ under (II).

TABLE I

<table>
<thead>
<tr>
<th>Coefficients of the Correction Term $c(w)$</th>
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<tbody>
<tr>
<td>(I) $\beta_1 - 1/\beta_2 = 0$</td>
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<tr>
<td>(II) $\beta_1 \beta_2 - 1 = 0$</td>
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<tr>
<td>$\alpha_0$</td>
</tr>
<tr>
<td>$\alpha_1$</td>
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<tr>
<td>$\alpha_2$</td>
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<tr>
<td>$\alpha_3$</td>
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</table>
It is clear from the polynomial correction factor
\[(1/T) c(w) = (1/T)(-10^{-6}w - 400w^2 + 100w^3)\] under (I)
that the Edgeworth expansion produces large corrections (often above 50 percent) to nominal asymptotic probabilities even for sample sizes as big as $T = 100$. This is strong evidence that the asymptotic distribution of $W$ under (I) is a poor approximation in this region of the parameter space ($\beta_1 = 10, \beta_2 = 0.1$) and that the gap between the finite sample and asymptotic distributions of $W$ is substantial. This confirms the experimental results of Gregory and Veall. Theory also strongly supports their recommendation that (II), the multiplicative form of the Wald test, is to be preferred on the basis of the accuracy of nominal size if we have reason to believe that $\beta_2$ may be in the vicinity of the origin. For in this case the polynomial correction factor
\[(1/T) c(w) = (1/T)(-10^{-6}w - 0.01w + 10^{-6}w^3)\] under (II)
delivers only very minor corrections to the asymptotic even when samples are as small as $T = 10$.

(ii) Lafontaine and White (1986)

These authors consider the algebraically equivalent restrictions:

(I) $\beta = 1$,

(II)$_k$ $\beta^k = 1$, $k \in \mathbb{Z}$.

In this case the coefficients in our correction factor $c(w)$ for the distribution of the Wald test based on (II)$_k$ are given by:

$\alpha_0 = \alpha_1 = 0,$
$\alpha_2 = -\frac{1}{2}(k-1)(k-2),$
$\alpha_3 = \frac{1}{4}(k-1)^2.$

Note that all coefficients vanish when $k = 1$, as is to be expected since the Wald statistic has an exact $\chi^2$ distribution in this case. Moreover, $\alpha_5 = O(|k|^2), \alpha_3 = O(|k|^3)$ as $|k| \to \infty$. Thus, the correction on the first order asymptotics becomes more substantial as $|k|$ increases. In other words, as the nonlinearity of the formulation (II)$_k$ increases, the adequacy of the asymptotic $\chi^2$ approximation deteriorates and the nominal critical values for the test that are delivered from asymptotic theory become less reliable. For large $|k|$ the deviations from the asymptotic theory may be expected to be large. This confirms the finding of Lafontaine and White.

(iii) Breusch and Schmidt (1985)

These authors give the example of the equivalent null hypotheses:

(I) $\beta = 0$,

(II) $h(\beta) = 0,$
where \( h \) satisfies \( h(0) = 0 \) and \( h'(\beta) > 0 \) for all \( \beta \). To test these equivalent hypotheses the following two Wald statistics are considered:

\[
W_1 = T\bar{X}^2, \quad W_2 = Th^2(\bar{X})/(h'(\bar{X}))^2,
\]

where \( \bar{X} \) is the sample mean of a random sample of iid \( N(\beta, 1) \) variates. Taking \( h \) to be continuously differentiable to the third order we deduce from Theorem 2.4 after some elementary manipulations that the distribution of \( W_2 \) has the following Edgeworth expansion to \( O(T^{-1}) \):

\[
(23) \quad cdf(w) = F_1(w) - \frac{1}{T} c(w) f_1(w) + o(T^{-1})
\]

where

\[
(24) \quad c(w) = -\frac{2h^{(3)}(0)}{3h'(0)} w^2 + \frac{1}{4}\left(\frac{h^{(2)}(0)}{h'(0)}\right)^2 w^3.
\]

For certain functions \( h \) in (II) the correction term (24) will lead to substantial deviations from the first order asymptotics. Breusch and Schmidt suggest the following function to illustrate possibilities:

\[
h(\beta) = \left\{ 1 + e^{-b(\beta-c)} \right\}^{-1} - \left\{ 1 + e^{bc} \right\}^{-1}, \quad \beta \geq 0,
\[
= -h(-\beta), \quad \beta < 0,
\]

where \( b > 0 \) and \( c \) is unrestricted. Simple calculations verify that

\[
\frac{h^{(3)}(0)}{h'(0)} = -\frac{b(1 - e^{bc})}{1 + e^{bc}}
\]

and

\[
\frac{h^{(3)}(0)}{h'(0)} = \frac{b^2(1 - 4e^{2bc} + e^{2bc})}{(1 + e^{bc})^2}.
\]

These expressions can be large for large \( b \). Thus, we find

\[
\frac{h^{(3)}(0)}{h'(0)} = O(b), \quad \frac{h^{(3)}(0)}{h'(0)} = O(b^2)
\]

as \( b \to \infty \). In such cases the distribution of \( W_2 \) can be expected to be poorly approximated by the asymptotic \( \chi^2 \).

By contrast, \( W_1 \) is quadratic in \( \bar{X}^2 \) and \( W_1 = \chi^2 \) for all \( T \). Thus, when \( h \) is linear the asymptotic distribution is exact.

The form of the Edgeworth expansion (23) and the correction factor (24) suggest that it may be possible to select formulations (II) which accelerate convergence to the asymptotic \( \chi^2 \) distribution by eliminating (or, at least reducing the magnitude of) the correction. Thus, any function \( h \) for which

\[
(25) \quad h^{(2)}(0) = h^{(3)}(0) = 0
\]
will be sufficient to eliminate the $O(T^{-1})$ correction term on the asymptotic and hence accelerate the convergence to an error of $O(T^{-2})$. Clearly a function $h$ satisfying (25) is approximately linear in the vicinity of the origin, so that this approach goes some way towards suggesting the formulation (1). In fact, if $h$ were assumed to be analytic in a fixed neighborhood of the origin, we could develop a complete asymptotic series for the distribution of $W_2$. This would involve an expansion of the form (23) taken to an arbitrary number of terms. Although we shall not report the details here, application of the argument given above on accelerated convergence in this case would lead to the choice of a function $h$ for which

$$h^{(j)}(0) = 0, \quad \text{all } j = 2, 3, \ldots.$$  (26)

Since $h$ is analytic and has a convergent power series representation, (26) would then imply a preferred choice of a linear function. In this case, therefore, the argument leads directly to the formulation (1) of the Wald test.

4 CONCLUDING REMARKS

Since the finite sample distribution of the Wald statistic for testing a nonlinear restriction can depend substantially on the algebraic representation of the restriction, the study of functional forms is of great importance in this context. The present paper provides a theoretical study of the problem using asymptotic expansions. The correction terms that are delivered by these expansions are shown to be extremely useful in evaluating algebraically equivalent, competing formulations of the Wald test. We have derived the explicit form of the asymptotic expansion of the distribution of the Wald statistic to $O(T^{-1})$ for a general class of parametric restrictions. The formulae obtained should be of independent interest. Moreover, upon specialization to the testing problems considered recently by other authors, these formulae explain well the discrepancies that have been observed in the finite sample behavior of alternative Wald tests. Finally, as we have seen in Section 3, Edgeworth expansions are not only useful in explaining phenomena such as the Wald test discrepancies. They may also be used to locate transformations which attenuate or even eliminate the effect of higher order terms, thereby accelerating convergence to the asymptotic distribution and helping to make it more reliable in finite samples.

This paper does not address the big question facing empirical researchers of what is the best way to formulate a Wald test of a given nonlinear restriction. But we can tell, using the approach we have adopted, how adequate the usual asymptotic theory of the Wald test is likely to be for different formulations of the restrictions in a given region of the parameter space. This is useful information. To tackle the bigger question we must first measure the loss associated with errors of different magnitude (and sign) in asymptotic significance testing and, presumably, be prepared to average this loss in some way over relevant regions of the parameter space for different formulations of the test. In principle, this is possible but, in practice, it is likely to place excessive demands on the quality of
the corrections delivered by Edgeworth expansions. We prefer the more open ended solution to the problem presented by our own approach. This puts less demands on the numerical quality of the corrections and serves in the role of a diagnostic device for assessing likely trouble spots in the formulation of Wald tests.

Finally, we should remark that algebraically equivalent formulations of non-linear restrictions generally lead to nonequivalent formulations of alternatives, including local alternatives. Thus

(I) \[ \beta_1 - 1/\beta_2 = d/T^{1/2} \]

is not equivalent to

(II) \[ \beta_1 \beta_2 - 1 = d/T^{1/2} \]

unless \(d(\text{II}) = \beta_2 \, d(\text{I})\). Such differences should be taken into account in power evaluations under local alternatives of different formulations of the Wald test. These may be studied by methods analogous to those of the present paper.

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APPENDIX A

PROOF OF LEMMA 21: Equations (13) and (14) are given by Magnus and Neudecker (1979, Theorem 4.1). To prove (15) we note that

\[
\sum_{i,j} (E_{ij} \otimes T_{ij}) = \sum_{i,j} (E_{ij} \otimes E_{ij}) + \sum_{i,j} (E_{ij} \otimes E_{ij})
\]

so that

\[
\sum_{i,j} (T_{ij} \otimes T_{ij}) = 2\{ K_{pp} + (\text{vec } I)(\text{vec } I) \}
\]

Also

\[
\sum_{i,j} (T_{ij} \otimes I \otimes T_{ij}) = \sum_{ij} K_{pp}^2 \{ I \otimes T_{ij} \otimes T_{ij} \} K_{pp}^2
\]

\[= 2 K_{pp}^2 \{ I \otimes \{ K_{pp} + (\text{vec } I)(\text{vec } I) \} \} K_{pp}^2. \]

Hence

(A.1) \[
\sum_{i,j} \{ (I \otimes T_{ij} \otimes T_{ij}) + (T_{ij} \otimes I \times T_{ij}) + (T_{ij} \otimes T_{ij} \otimes I) \}
\]

\[= 2 \{ I \otimes \{ K_{pp} + (\text{vec } I)(\text{vec } I) \} \}
\]+ \[K_{pp}^2 \{ I \otimes \{ K_{pp} + (\text{vec } I)(\text{vec } I) \} \} K_{pp}^2
\]

\[+ \{ K_{pp} + (\text{vec } I)(\text{vec } I) \} \otimes I \].

Now
\[ T_{ij} \otimes T_{ik} \otimes T_{jk} = (E_{ij} + E_{ik}) \otimes (E_{ik} + E_{jk}) \otimes (E_{jk} + E_{ij}) \]
and simple manipulations along the following lines yield.
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = \sum_{i, j, k} (e_i \otimes e_j)(e'_k \otimes e'_j) \otimes E_{jk}
\]
\[
= \sum_{i, j, k} \left\{ \text{vec}(E_{ij})(e'_k \otimes e'_j) \right\} \otimes E_{jk}
\]
\[
= \left\{ \sum_i \text{vec}(E_{ij}) \otimes \mathbf{I}_p \right\} \left\{ \sum_{j, k} \left\{ (e'_j \otimes e'_k) \otimes (e_j \otimes e'_j) \right\} \right\}
\]
\[
= \left\{ \text{vec} \mathbf{I} \otimes \mathbf{I}_p \right\} \sum_{j, k} \left\{ \left( e'_j \otimes 1 \otimes e_j \otimes 1 \right)(\mathbf{I}_p \otimes e'_k \otimes 1 \otimes e_k) \right\}
\]
\[
= \left\{ \text{vec} \mathbf{I} \otimes \mathbf{I}_p \right\} \left\{ \sum_{j} \left( e'_j \otimes e_j \right)(\mathbf{I} \otimes e'_k \otimes e_j) \right\}
\]
\[
= \left\{ \text{vec} \mathbf{I} \otimes \mathbf{I}_p \right\} \left\{ \sum_{j} \left( \sum_k (e'_j \otimes e'_k) \right) \right\}
\]
\[
= \left\{ \text{vec} \mathbf{I} \otimes \mathbf{I}_p \right\} \left\{ \mathbf{I} \otimes \sum_k (e'_j \otimes e'_k) \right\}
\]
\[
= \left\{ \text{vec} \mathbf{I} \otimes \mathbf{I}_p \right\} \left\{ \mathbf{I} \otimes (\text{vec} \mathbf{I})' \right\}
\]

Note that
\[
K_{pp^2}((\text{vec} \mathbf{I}) \otimes \mathbf{I})K_{pp^2} = K_{pp^2}((\text{vec} \mathbf{I}) \otimes \mathbf{I}) = \mathbf{I} \otimes (\text{vec} \mathbf{I})
\]
so that, using the fact that \( K_{pp^2} = K_{pp^2} \), we deduce that
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = K_{pp^2}(\mathbf{I} \otimes (\text{vec} \mathbf{I})(\text{vec} \mathbf{I})')
\]

In a similar way we find that,
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = K_{pp^2}(\mathbf{I} \otimes (\text{vec} \mathbf{I})(\text{vec} \mathbf{I})'),
\]
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = (\mathbf{I} \otimes \mathbf{I})K_{pp^2},
\]
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = K_{pp^2}(\mathbf{I} \otimes (\text{vec} \mathbf{I})(\text{vec} \mathbf{I})')K_{pp^2},
\]
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = K_{pp^2}(\mathbf{I} \otimes (\text{vec} \mathbf{I})(\text{vec} \mathbf{I})')K_{pp^2},
\]
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = (\mathbf{I} \otimes \mathbf{I})K_{pp^2}
\]
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = (\mathbf{I} \otimes (\text{vec} \mathbf{I})(\text{vec} \mathbf{I})')K_{pp^2},
\]
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = (\mathbf{I} \otimes (\text{vec} \mathbf{I})(\text{vec} \mathbf{I})')K_{pp^2},
\]
\[
\sum_{i, j, k} (E_{ij} \otimes E_{ik} \otimes E_{jk}) = (\mathbf{I} \otimes (\text{vec} \mathbf{I})(\text{vec} \mathbf{I})')K_{pp^2}
\]
It now follows from (14), (A1) and the above that:

\[(A2) \quad E(z'z \otimes zz') = -I + K_{pp} + K_{pp}' + (I \otimes K_{pp}) + (K_{pp} \otimes I) + K_{pp}'(I \otimes K_{pp}) K_{pp} + (K_{pp} \otimes I)(I \otimes (K_{pp}'))(I \otimes I)' \otimes I + K_{pp}' \{I \otimes (vec I)(vec I)'\} K_{pp} + (I \otimes (vec I)(vec I)')(I \otimes I)' \otimes I + K_{pp}' \{I \otimes (vec I)(vec I)'\} K_{pp} + (K_{pp} \otimes K_{pp}')(I \otimes (vec I)(vec I)')(I \otimes I)' \otimes I'.\]

Noting that

\[(vec I)(vec I)' \otimes I = K_{pp}^2 (I \otimes (vec I)(vec I)') K_{pp},\]

we deduce that the final six terms of (A2) sum to

\[(I + K_{pp}) K_{pp} \} (I \otimes (vec I)(vec I)')(I + K_{pp} + K_{pp}'),\]

leading to result (15) as stated in the lemma.

**Proof of Lemma 2.2:** Let \( z = V^{-1/2} x = N(0, I) \). Then

\[E(xx' \otimes xx') = (V^{1/2} \otimes V^{1/2}) E(z'z \otimes zz')(V^{1/2} \otimes V^{1/2}) = (I + K_{pp})(V \otimes V) + (vec V)(vec V)\]

as required for (16). Similarly

\[E(xx' \otimes xx' \otimes xx') = (V^{1/2} \otimes V^{1/2} \otimes V^{1/2}) E(z'z \otimes zz' \otimes zz')(V^{1/2} \otimes V^{1/2} \otimes V^{1/2})\]

and result (17) follows directly in view of the properties of the commutation matrices \( K_{pp}, K_{pp}', K_{pp} \).

**Proof of Lemma 2.3:** We note that \( \bar{q} = Rz = N(0, R^2) = N(0, S^{-1}) \) and

\[(A3) \quad E(g) = Tr \{ LE(\bar{q} \bar{q}) \} = Tr \{ L \{ (I + K_{pp})(S^{-1} \otimes S^{-1}) + (vec S^{-1})(vec S^{-1})' \} \}

Now

\[(A4) \quad S^{-1} = \bar{p}_{pp} + (1 - 2u)t^{-1} \bar{p}_{pp},\]

so that upon expansion of (A3) we obtain

\[E(g) = Tr(A_0) + Tr(A_1)(1 - 2u)t^{-1} + Tr(A_2)(1 - 2u)^{-2}\]

as required for (18). To prove (19) we write

\[(A5) \quad E(x^2) = (vec J)'E(\bar{q} \bar{q}' \otimes \bar{q} \bar{q}') (vec J) = (vec J)' \{ B_0 + (1 - 2u)B_1\} (vec J).\]

Upon evaluation using (17) with \( V = S^{-1} \) and noting that \( (vec J)' \bar{R}_{vec}(vec J) = 0 \), we find that (A5) yields the stated formula (19).

**Proof of Theorem 2.4:** Term by term Fourier inversion of (20) yields

\[(A6) \quad pdf(w) = f_0(w) + \frac{1}{T} \{ \phi_1(-1)f_1(w) + \psi_1(-1)f_{s1}^2(w)\]

\[+ \phi_2(-1)f_2^2(w) + \phi_4(-1)f_4^2(w) + o(T^{-1})\]

where \( f_0(w) \) is the constant term in the Fourier series of \( f(w) \), and \( f_1(w), f_2(w), f_3(w), f_4(w) \) are the Fourier coefficients of \( f(w) \).
where
\[ \phi_1 = a_0 - b_1/4, \]
\[ \phi_2 = a_1 + b_1/4 - b_2/4, \]
\[ \phi_3 = a_2 + b_2/4 - b_3/4, \]
\[ \phi_4 = b_3/4. \]

Upon integration of (A6) we obtain
\[ \text{cdff}(w) = F_r(w) + \frac{1}{T} \left\{ -\phi_1 f_r(w) - \phi_2 f_{r+2}(w) - \phi_3 f_{r+4}(w) \right\} \]
\[ -\phi_4 f_{r+6}(w) + o(T^{-1}) \]
and in view of the relationship
\[ f_{r+2n}(w) = 2^{-n} w^n f_r(w)/(r/2)_n. \]

(A7) can be rewritten in the form
\[ \text{cdff}(w) = F_r(w) - \frac{1}{T} f_r(w) \left[ \phi_1 + \frac{\phi_2 w}{r + 2} + \frac{\phi_3 w^2}{r + 2} + \frac{\phi_4 w^3}{(r + 2)(r + 4)} \right] + o(T^{-1}) \]
which reduces to (21) as stated upon translation of notation. Validity of the expansion as a proper asymptotic series follows from the theorem in Sargent (1980).

APPENDIX B

This appendix derives formulae for the Edgeworth expansion to \( O(T^{-1}) \) of the distribution of the Wald statistic (5) in full generality. Thus, in place of assumption (4) we require only that the distribution of the component variates \( \bar{q} \) have a valid Edgeworth expansion of the form (3). Let \( \bar{q} = \bar{U}^{-1/2} \bar{q} \) and carrying terms to \( O(T^{-1}) \) we write the Edgeworth expansion as:
\[ \text{pdf}(\bar{q}) = (2\pi)^{-r/2} e^{-\bar{q}^2/2} \left[ 1 + T^{-1/2} \left\{ f_0(\bar{q}) + f_1(\bar{q}) \bar{q} + f_2(\bar{q}) \bar{q}^2 \right\} \right. \]
\[ + T^{-1} \left\{ f_0 + \text{tr} \left( F_2 \bar{q} \bar{q}^T \right) \right\} \]
\[ \left. + \text{tr} \left( F_3 \bar{q} \bar{q}^T \bar{q} \bar{q}^T \right) \right] + o(T^{-1}). \]

This new form of (3) up to \( O(T^{-1}) \) has not been used in the literature before. But it is most convenient for our algebraic approach to Edgeworth expansions in explicit matrix form.

The Wald statistic for testing \( H_0 \) as given by (1) is now:
\[ W = T \hat{q} ^ T \left( \hat{G} \hat{H} \hat{G}^-1 \right) ^{-1} \hat{q} \]
where \( \hat{q} = \Omega(\hat{U}) \) is a consistent estimator of \( \Omega \) and where it is assumed that \( GB \) (respectively \( \hat{G} \hat{H} \)) is nonsingular (with probability one). The function \( \Omega(\cdot) \) representing the estimator of \( \Omega \) is assumed to be continuously differentiable to the third order. We write \( \bar{\Psi}^{-1} = (\hat{G} \hat{H} \hat{G}^-1)^{-1} = (\hat{\Psi})^{-1} \) and employ the Taylor representations
\[ \sqrt{T} \hat{q} = \hat{g}_0 + \frac{1}{\sqrt{T}} \hat{g}_1 + \frac{1}{2T} \hat{g}_2 + O_p(T^{-3/2}), \]
\[ \hat{\Psi} = \Psi + \frac{1}{\sqrt{T}} \hat{\Psi}_m + \frac{1}{2T} \hat{\Psi}_m^2 + O_p(T^{-3/2}), \]
where
\[ \hat{g}_0 = g_0 \omega_0, \quad \hat{g}_1 = g_{10} \omega_0 + \omega_0, \]
\[ \hat{g}_2 = g_{20} \omega_0^2 + g_{11} \omega_0 \omega_1, \quad \hat{\Psi}_m = \psi_m \omega_m, \]
\[ \hat{\Psi}_m^2 = \psi_m^2 \omega_m^2, \]
and where \( \Omega^{1/2} = (\omega_m) \) is the symmetric positive definite square root of \( \Omega \).
In place of (7) we now obtain the expansion
\[ W = \tilde{q}^T \Omega^{-1} G^{-1} (G \Omega G')^{-1} G \Omega \tilde{q} + T^{-1/2} u(\tilde{q}) + T^{-1} v(\tilde{q}) + O_p(T^{-3/2}) \]
where
\[ u(\tilde{q}) = \text{vec}(\tilde{J})(\tilde{q} \otimes \tilde{q} \otimes \tilde{q}), \]
\[ v(\tilde{q}) = \text{tr}\{ \tilde{L}(\tilde{q} \tilde{q}' \otimes \tilde{q} \tilde{q}') \} \]
Here vec(\tilde{J}) has \((p^2 \cdot j - 1) + p(b - 1) + m \)'th element:
\[ \tilde{g}_{ij} \tilde{q}_{iab} \tilde{g}_{jcm} \tilde{g}_{ab} + \tilde{g}_{ij} \tilde{g}_{jcm} \tilde{g}_{ab} + \tilde{g}_{ij} \tilde{g}_{jcm} \tilde{g}_{ab} + \tilde{g}_{ij} \tilde{g}_{jcm} \tilde{g}_{ab} \]
and \( \tilde{L} \) is \( p^2 \times p^2 \) with \((j(p - 1) + b, m(p - 1) + n)\)'th element:
\[ \frac{1}{2} \tilde{g}_{ij} \tilde{g}_{jcm} \tilde{g}_{ab} = \tilde{g}_{ij} \tilde{g}_{jcm} \tilde{g}_{ab} + \frac{1}{2} \tilde{g}_{ij} \tilde{g}_{jcm} \tilde{g}_{ab} + \frac{1}{2} \tilde{g}_{ij} \tilde{g}_{jcm} \tilde{g}_{ab} \]
In place of (10) the characteristic function of \( W \) therefore becomes:
\[ \text{cf}_W(t) = (2\pi)^{-p/2} \int_{\mathbb{R}^p} \exp \left\{ -\frac{1}{2} \tilde{q}' \tilde{S} \tilde{q} \right\} \]
\[ \times \left\{ 1 + \frac{it}{\sqrt{T}} u(\tilde{q}) + \frac{it}{T} v(\tilde{q}) - i \frac{t}{2T} u''(\tilde{q}) \right\} \]
\[ \times \left\{ 1 + T^{-1/2} \left[ f_{1} + f_{1}'(\tilde{q} \otimes \tilde{q} \otimes \tilde{q}) \right] \right. \]
\[ + T^{-1} \left[ f_{0} + \text{tr}(F_{1} \tilde{q} \tilde{q}''') + \text{tr}(F_{1} \tilde{q} \tilde{q}'' \otimes \tilde{q}''') \right] \]
\[ + \text{tr}(F_{1} \tilde{q} \tilde{q}' \otimes \tilde{q}'' \otimes \tilde{q}''') \right\} d\tilde{q} + o(T^{-1}) \]
where
\[ \tilde{S} = I - 2it \tilde{G} (G \tilde{G})^{-1} \tilde{G}, \quad \tilde{G} = G \Omega^{-1}. \]
Upon integration and simplification this reduces to:
\[ \text{cf}_W(t) = (1 - 2it)^{-p/2} \left\{ 1 + T^{-1} \left[ f_{0} + \text{tr}(F_{1} \tilde{S}^{-1}) + \text{tr}(F_{1} E(\tilde{q} \tilde{q}'' \otimes \tilde{q}''')) \right] \right. \]
\[ + \text{tr}(F_{1} E(\tilde{q} \tilde{q}' \otimes \tilde{q}'' \otimes \tilde{q}''')) \right\} \]
\[ + (it/T) \left\{ E(u) + \text{tr}(F_{1} E(\tilde{q} \tilde{q}' \otimes \tilde{q}'' \otimes \tilde{q}''')) \right\} \]
\[ + \text{vec}(\tilde{J} \tilde{q}' E(\tilde{q} \tilde{q}' \otimes \tilde{q}'' \otimes \tilde{q}''')) f_{0} \]
\[ - (t^2/2T) E(u^2) \] + o(T^{-1}) \]
where \( E \) denotes the expectation operator with respect to \( p(\tilde{q}) = N(0, \tilde{S}^{-1}) \) and where vec(\( F_{1} \)) = \( f_{1} \otimes \text{vec}(\tilde{J}) \).

Using Lemmas 2.2 and (2.3) we find that:
\[ E(\tilde{q} \tilde{q}' \otimes \tilde{q}'' \otimes \tilde{q}''') = D_{0} + D_{1}(1 - 2it)^{-1} + D_{2}(1 - 2it)^{-2}, \]
\[ E(\tilde{q} \tilde{q}'' \otimes \tilde{q}'' \otimes \tilde{q}''') = \tilde{b}_{1} + \tilde{b}_{1}(1 - 2it)^{-1} \]
\[ + \tilde{b}_{2}(1 - 2it)^{-2} + \tilde{b}_{2}(1 - 2it)^{-3}, \]
\[ E(u^2) = \text{vec}(\tilde{J} \tilde{q}) \left\{ \tilde{b}_{1} + \tilde{b}_{1}(1 - 2it)^{-1} \right. \]
\[ + \tilde{b}_{2}(1 - 2it)^{-2} + \tilde{b}_{2}(1 - 2it)^{-3} \right\} \text{vec}(\tilde{J}), \]
\[ E(v) = \text{tr}(\tilde{A}_{0}) + \text{tr}(\tilde{A}_{0}(1 - 2it)^{-1}) + \text{tr}(\tilde{A}_{0})(1 - 2it)^{-2}, \]
where \( \{ \tilde{B}_{j}, \tilde{A}_{k} : j = 0, 1, 2, 3; k = 0, 1, 2 \} \) are matrices identical in form to those defined in Lemma 2.3.
but with $P_{\mathcal{C}} = \mathcal{G}(\bar{\mathcal{G}}^{-1})^2 \mathcal{G}^{-1} \mathcal{G}(G_\mathcal{C})^{-1} G_{\mathcal{C}}$ and $\bar{L}$ in place of $P_{\mathcal{C}}$ and $L$, and where

$$D_0 = (I + K_{pp}) (\bar{P}_{\mathcal{C}} \otimes \bar{P}_{\mathcal{C}}) + (\text{vec} \bar{P}_{\mathcal{C}})(\text{vec} \bar{P}_{\mathcal{C}})' ,$$

$$D_1 = (I + K_{pp}) (\bar{P}_{\mathcal{C}} \otimes P_{\mathcal{C}} + P_{\mathcal{C}} \otimes \bar{P}_{\mathcal{C}}) + (\text{vec} \bar{P}_{\mathcal{C}})(\text{vec} P_{\mathcal{C}})' + (\text{vec} P_{\mathcal{C}})(\text{vec} \bar{P}_{\mathcal{C}})' ,$$

$$D_2 = (I + K_{pp}) (P_{\mathcal{C}} \otimes P_{\mathcal{C}}) + (\text{vec} P_{\mathcal{C}})(\text{vec} P_{\mathcal{C}})' .$$

We now obtain

$$c_{fw}(t) = (1 - 2ut)^{-r/2} \left[ 1 + \frac{1}{T} \sum_{j=0}^{3} c_j (1 - 2ut)^{-j} + \frac{r}{T} \sum_{j=0}^{3} \bar{c}_j (1 - 2ut)^{-j} \right] + o(T^{-1})$$

where

$$c_0 = f_0 + \text{tr}(F_2 \bar{P}_{\mathcal{C}}) + \text{tr}(F_2 D_0) + \text{tr}(F_2 \bar{A}_0) ,$$

$$c_1 = \text{tr}(F_2 P_{\mathcal{C}}) + \text{tr}(F_2 D_1) + \text{tr}(F_2 \bar{B}_1) ,$$

$$c_2 = \text{tr}(F_2 D_2) + \text{tr}(F_2 \bar{B}_2) ,$$

$$c_3 = \text{tr}(F_2 \bar{B}_3) .$$

$$e_j = \tilde{a}_j + d_j \quad (j = 0, 1, 2, 3) ,$$

$$d_j = (\text{vec} \bar{J}) \bar{B}_j f_j \quad (j = 0, 1, 2, 3) ,$$

$$\tilde{a}_j = \left\{ \begin{array}{l} \text{tr}(A_j) + \text{tr}(F_j D_j) \quad (j = 0, 1, 2) , \\ 0 \quad (j = 3) . \end{array} \right.$$
Formula (B4) is the Edgeworth expansion to $o(T^{-1})$ of the distribution of the Wald statistic (B1) in the general case. It should be useful in many situations besides those considered in the present paper.

REFERENCES


