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Article

# 5th Order Multivariate Edgeworth Expansions for Parametric Estimates

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**Abstract:** The only cases where exact distributions of estimates are known is for samples from exponential families, and then only for special functions of the parameters. So statistical inference was traditionally based on the asymptotic normality of estimates. To improve on this we need the *Edgeworth expansion* for the distribution of the standardized estimate. This is an expansion in  $n^{-1/2}$  about the normal distribution, where n is typically the sample size. The 1st few terms of this expansion were originally given for the special case of a sample mean. In earlier work we derived it for *any* standard estimate, hugely expanding its application. We call an estimate  $\hat{w}$  of an unknown vector  $w \in R^p$ , a standard estimate, if  $E \hat{w} \to w$  as  $n \to \infty$ , and for  $r \ge 1$  the rth order cumulants of  $\hat{w}$  have magnitude  $n^{1-r}$  and can be expanded in  $n^{-1}$ . Here we give another huge extension. We give the expansion of the distribution of *any smooth function* of  $\hat{w}$ , say  $t(\hat{w}) \in R^q$ , giving its distribution to  $n^{-5/2}$ . We do this by showing that  $t(\hat{w})$ , is a standard estimate of t(w). This provides far more accurate approximations for the distribution of  $t(\hat{w})$  than its asymptotic normality.

**Keywords:** Edgeworth expansions; parametric inference; standard estimates; chain rules for cumulant coefficients; channel capacity

# 1. Introduction and summary

Suppose that  $\hat{w}$  is a *standard* or *Type A* estimate of an unknown  $w \in R^p$  with respect to a given parameter n. That is,  $E \hat{w} \to w$  as  $n \to \infty$  and for  $r \ge 1$ , its rth order cumulants have magnitude  $n^{1-r}$  and can be expanded as

$$\bar{k}^{1-r} = \kappa(\hat{w}^{i_1}, \dots, \hat{w}^{i_r}) = \sum_{e=r-1}^{\infty} n^{-e} \, \bar{k}_e^{1-r} \text{ for } 1 \le i_1, \dots, i_r \le p, \tag{1}$$

where the *cumulant coefficients*  $\bar{k}_e^{1-r} = k_e^{j_1-j_r}$  do not depend on n, or at least are bounded as  $n \to \infty$ . So  $\bar{k}_0^1 = w^{i_1}$ . For example (1) holds for  $\hat{w}$  a function of a sample mean. We show that if  $t(\hat{w})$  is a smooth function of a standard estimate  $\hat{w}$ , then it is a standard estimate of t(w). This is done for  $\hat{w}$  unbiased in Theorem 3.1, and for  $\hat{w}$  biased in Theorem 4.1. More generally we call  $\hat{w}$  a to 0 setimate if to 0 as to 0, and for to 0, and for to 0, and for to 0 setimate if to 0 as to 0.

$$\bar{k}^{1-r} = \sum_{d-2r-2}^{\infty} n^{-d/2} \bar{b}_d^{1-r} \text{ for } 1 \leq i_1, \ldots, i_r \leq p, \ \bar{b}_d^{1-r} = b_d^{i_1 \ldots i_r}.$$

For example this type arises when considering 1-sided confidence regions. If  $t(\hat{w})$  is a smooth function of a Type B estimate, then it is a Type B estimate of t(w). So for a Type A estimate,  $\bar{b}_d^{1-r}$  is  $\bar{k}_e^{1-r}$  for d=2e and 0 for d odd. n is typically the sample size or the minimum sample size if there is more than one sample.

§3 and §4 show that a smooth function of  $\hat{w}$ , say  $t(\hat{w})$ , is a standard estimate of t=t(w). They give the cumulant coefficients of  $t(\hat{w})$  in terms of those of  $\hat{w}$  and the derivatives of t(w). §3 does this for  $\hat{w}$  unbiased and §4 for  $\hat{w}$  biased. So they can be thought of as *chain rules* for obtaining the cumulant coefficients for  $t(\hat{w})$  from those of  $\hat{w}$ . We give the cumulant coefficients needed for Edgeworth expansions of  $\hat{t}$  to  $O(n^{-5/2})$ . Those to  $O(n^{-1})$  were given in Withers and Nadarajah (2022). Those to  $O(n^{-r/2})$  use the rth derivatives of t(w). §5 specialises to univariate t(w) with examples. Theorem 4.1

and Corollary 5.4 correct  $\bar{a}_2^{12} = K_2^{j_1 j_2}$  and  $a_{22}$  on p67 and p59 of Withers (1982). §2 extends the *shorthand* bar notation above and gives the foundation theorem.

We now summarise the expressions for Edgeworth expansions of  $\hat{w}$  for standard and Type B estimates in terms of the cumulant coefficients  $\bar{k}_e^{1-r}$  and  $\bar{b}_d^{1-r}$  given in Withers and Nadarajah (2010b, 2012a, 2014a):

$$Prob.(Y_{nw} \le x) = \sum_{r=0}^{\infty} n^{-r/2} P_r(x), \ p_{Y_{nw}}(x) = \sum_{r=0}^{\infty} n^{-r/2} p_r(x), \tag{2}$$

where 
$$Y_{nw} = n^{1/2}(\hat{w} - w - b_1 n^{-1/2}), (b_1)_i = b_1^i, P_0(x) = \Phi_V(x),$$
 (3)

$$P_r(x) = \tilde{B}_r(e(-\partial/\partial x)) \Phi_V(x) \text{ for } r \ge 1,$$
 (4)

$$e_{j}(t) = \sum_{r=1}^{j+2} \bar{b}_{r+j}^{1\dots r} t_{i_{1}} \dots t_{i_{r}} / r!, \ \bar{b}_{r+j}^{1\dots r} = b_{r+j}^{i_{1}\dots i_{r}},$$

$$(5)$$

 $\Phi_V(x)$  is the multivariate normal distribution with zero mean and covariance  $V=(\bar{b}_2^{12})$ ,  $\tilde{B}_r(e)$  is the complete ordinary Bell polynomial of Comtet (1974):

$$\tilde{B}_1(e) = e_1$$
,  $\tilde{B}_2(e) = e_2 + e_1^2$ ,  $\tilde{B}_3(e) = e_3 + 2e_1e_2 + e_1^3$ ,  $\tilde{B}_4(e) = e_4 + 2e_1e_3 + e_2^2 + 3e_1^2e_2 + e_1^4$ .

This gives the 5th order Edgeworth expansion for the distribution of  $Y_{nw}$ , that is, it gives (2) to  $O(n^{-5/2})$ . Note that (5) uses *the tensor summation convention* of implicitly summing  $i_1, \ldots, i_r$  over their range  $1, \ldots, p$ . For example

for 
$$\partial_i = \partial/\partial x_i$$
 and  $\bar{\partial}_k = \partial_{i_k}$ , 
$$P_1(x) = e_1(-\partial/\partial x)) \Phi_V(x) = \sum_{r=1}^3 \bar{b}_{r+1}^{1...r} (-\bar{\partial}_1) \dots (-\bar{\partial}_r) \Phi_V(x)$$
$$= \bar{k}_1^1 (-\bar{\partial}_1) \Phi_V(x) + \bar{k}_2^{1-3} (-\bar{\partial}_1) (-\bar{\partial}_2) (-\bar{\partial}_3) \Phi_V(x)$$

for a standard estimate. For a standard estimate,  $b_1 = 0$  in (3) and the cumulant coefficients needed for  $P_r(x)$ ,  $p_r(x)$  of (2) are  $\bar{k}_0^1 = w^{i_1}$ ,

for 
$$r = 0$$
:  $\bar{k}_1^{12}$ ; for  $r = 1$ :  $\bar{k}_1^1$ ,  $\bar{k}_2^{1-3}$ ; for  $r = 2$ :  $\bar{k}_2^{12}$ ,  $\bar{k}_3^{1-4}$ ; (6)

for 
$$r = 3 : \bar{k}_2^1, \bar{k}_3^{1-3}, \bar{k}_4^{1-5};$$
 for  $r = 4 : \bar{k}_3^{12}, \bar{k}_4^{1-4}, \bar{k}_5^{1-6}.$  (7)

So to obtain the 5th order Edgeworth expansion for the distribution of  $n^{1/2}(t(\hat{w}) - t(w))$  for  $\hat{w}$  a standard estimate, we just need to replace the coefficients in (6) and (7) where they appear in  $P_r(x)$ ,  $r \le 4$ , by those of  $t(\hat{w})$  given in §3-§5.

(9) of Withers and Nadarajah (2010b) gives  $P_r(x)$  for the more general case where  $P_0(x)$  is the distribution function of  $Y \in \mathbb{R}^p$  which depends on n but is asymptotic to  $\Phi_V(x)$  and has a Type B expansion. One can choose  $P_0(x)$  so that the number of terms in each  $P_r(x)$  greatly reduces: see Withers and Nadarajah (2012d, 2014c, 2015). When  $\hat{w}$  is lattice, further terms need to be added: see for example Chapter 5 of Bhattacharya and Rao (1976), Cai (2005), and for the density of  $Y_{nw}$ , p211 of Barndoff-Nielsen and Cox (1989), §5 of Daniels (1983), and §6 of Daniels (1987). Corollary 1 of Withers and Nadarajah (2010b) gives the *tilted* Edgeworth expansion for  $t(\hat{w})$ , sometimes called the *saddlepoint approximation*, or the *small sample* expansion as it is a series in  $n^{-1}$  not just  $n^{-1/2}$ . It is very useful for the tails of the distribution where Edgeworth expansions perform poorly. Cumulant coefficients are also needed for bias reduction, Bayesian inference, confidence regions and power. See Withers (1984)

and Withers and Nadarajah (2008, 2010a, 2011b, 2011c, 2012b, 2012f, 2014b, 2014c, 2015) for examples. For a history of Edgeworth expansions, see §7.

In summary, this paper gives high order expansions for the distribution of a vast range of estimates, by obtaining the cumulant coefficients needed for any smooth function of a standard estimate. This provides unprecedented accuracy for these distributions and avoids the need for simulation methods.

# 2. Foundations

Given  $w=(w^1,\cdots,w^p)\in R^p$  and an estimate  $\hat{w}$ , suppose that  $E\ \hat{w}\to w$  as  $n\to\infty$  and that for  $r\ge 1$ , its rth order cumulants have magnitude  $n^{1-r}$ . Given  $i_1,\cdots,i_r$  in  $1,2,\cdots,p$ , we write these cumulants in shorthand as

$$\bar{k}^{1-r} = k^{i_1 \cdots i_r} = \kappa(\hat{w}^{i_1}, \dots, \hat{w}^{i_r}) = O(n^{1-r}) \text{ as } n \to \infty.$$
 (1)

For example if  $\hat{w} = \bar{X}$  is the mean of a random sample of size n, then (1) holds since  $\bar{k}^{1-r} = n^{1-r}\kappa(X^{i_1},\ldots,X^{i_r})$  where  $X^i$  is the ith component of X. By Theorem 2.1, (1) holds if  $\hat{w}$  is a smooth function of one or more sample means. Let  $t: R^p \to R^q$  be a smooth function in a neighbourhood of w with jth component  $t^j = t^j(w)$ ,  $j = 1, \cdots, q$  and finite partial derivatives

$$\bar{t}_{sr...}^k = t_{.i_s i_r...}^{j_k} = \partial_{i_s} \partial_{i_r} \cdots t^{j_k}(w), \ \bar{t}_{s-r}^k = t_{.i_s...i_r}^{j_k} \text{ for } s \leq r$$

where  $\partial_i = \partial/\partial w^i$ . We reserve i's as superscripts for the cumulants of  $\hat{w}$  and subscripts for partial derivatives of t(w). We reserve j's as superscripts for the components of t(w) and for the joint cumulants of  $\hat{t} = t(\hat{w})$ . This bar shorthand allows us to shorten expressions by suppressing the i's and j's. We write the cumulants of  $\hat{t} = t(\hat{w})$  as

$$\bar{K}^{1-r} = K^{j_1 \cdots j_r} = \kappa(\hat{t}^{j_1}, \dots, \hat{t}^{j_r}) \text{ where } \hat{t} = t(\hat{w}), \ \hat{t}^j = t^j(\hat{w}).$$
 (2)

For example

$$\bar{k}^{12} = k^{i_1 i_2}$$
,  $\bar{K}^{12} = K^{j_1 j_2} \Rightarrow covar(\hat{w}) = (\bar{k}^{12})$ , and  $covar(\hat{t}) = (\bar{K}^{12})$ .

are  $O(n^{-1})$ . We now show that

$$\bar{K}^{12} = {}_1\bar{K}^{12} + O(n^{-2}) \text{ where } {}_1\bar{K}^{12} = \bar{t}_1^1\bar{t}_2^2\,\bar{k}^{12}, \text{ that is, } {}_1K^{j_1j_2} = t_{.i_1}^{j_1}t_{.i_2}^{j_2}k^{i_1i_2},$$

using the tensor sum convention. The rest of this section and all proofs can be skipped on a 1st reading. Theorem 2.1 gives the cumulants of  $\hat{t} = t(\hat{w})$  when  $\hat{w}$  is unbiased.

We shall use the notation  $\sum^N f^{j_1 j_2 \cdots}$  to mean summing over all N permutations of  $j_1, j_2, \cdots$  giving distinct terms.

**Theorem 1.** Suppose that  $E \hat{w} = w$  and that (1) holds. Then for  $r \ge 1$  and  $1 \le j_1, \ldots, j_r \le q$ ,  $\bar{K}^{1-r}$  of (2) satisfies

$$\bar{K}^{1-r} = \sum_{e=r-1}^{\infty} {}_{e}\bar{K}^{1-r} \text{ where } {}_{e}\bar{K}^{1-r} = {}_{e}K^{j_{1}\cdots j_{r}} = O(n^{-e}) \text{ as } n \to \infty,$$
 (3)

and the leading  $_{e}\bar{K}^{1-r}$  are as follows.

$$\begin{array}{l} {}_{0}\bar{K}^{1}=\bar{t}^{1},\ that\ is,\ {}_{0}K^{j_{1}}=t^{j_{1}},\\ {}_{1}\bar{K}^{1}=\bar{t}^{1}_{12}\bar{k}^{12}/2,\ that\ is,\ {}_{1}K^{j_{1}}=t^{j_{1}}_{.i_{1}i_{2}}\ k^{i_{1}i_{2}}/2=\sum\limits_{i_{1},i_{2}=1}^{p}t^{j_{1}}_{.i_{1}i_{2}}\ k^{i_{1}i_{2}}/2,\\ {}_{2}\bar{K}^{1}=\bar{t}^{1}_{1-3}\ \bar{k}^{1-3}/6+\bar{t}^{1}_{1-4}\ \bar{k}^{12}\bar{k}^{34}/8,\\ that\ is,\ {}_{2}K^{j_{1}}=t^{j_{1}}_{.i_{1}i_{2}i_{3}}k^{i_{1}i_{2}i_{3}}/6+t^{j_{1}}_{.i_{1}-i_{4}}\bar{k}^{i_{1}i_{2}}\bar{k}^{i_{3}i_{4}}/8,\\ {}_{3}\bar{K}^{1}=\bar{t}^{1}_{1-4}\bar{k}^{1-4}/24+\bar{t}^{1}_{1-5}\bar{k}^{1-3}\bar{k}^{45}/12+\bar{t}^{1}_{1-6}\ \bar{k}^{12}\bar{k}^{34}\bar{k}^{56}/48,\\ {}_{4}\bar{K}^{1}=\bar{t}^{1}_{1-5}\bar{k}^{1-5}/120+\bar{t}^{1}_{1-6}\ (\bar{k}^{1-4}\bar{k}^{56}/48+\bar{k}^{1-3}\bar{k}^{4-6}/72)+\bar{t}^{1}_{1-7}\ \bar{k}^{1-3}\bar{k}^{45}\bar{k}^{67}/48\\ +\bar{t}^{1}_{1-8}\ \bar{k}^{12}\bar{k}^{34}\bar{k}^{56}\bar{k}^{78}/384, \end{array}$$

$$\begin{split} &_1\vec{k}^{12} = \vec{t}_1^1 \vec{t}_2^2 \, \vec{k}^{12}, \, _2\vec{k}^{12} = T_{1-3}^{12} \, \vec{k}^{1-3} / 2 + T_{1-4}^{12} \, \vec{k}^{12} \vec{k}^{34} / 2 \, where \\ &T_{1-3}^{12} = \sum_{l=1}^{2} \vec{t}_1^1 \vec{t}_3^2 \, \vec{t}_1^{1-4} = \sum_{l=1}^{2} \vec{t}_{1-3}^1 \vec{t}_4^2 + \vec{t}_{13}^1 \vec{t}_{24}^2, \, \sum_{l=1}^{2} \vec{t}_{1-d}^1 \vec{t}_{2-d}^2 = \vec{t}_{1-b}^1 \vec{t}_{2-d}^2 + \vec{t}_{2-b}^1 \vec{t}_{1-d}^2, \\ &_3\vec{k}^{12} = U_{1-4}^{12} \, \vec{k}^{1-4} + T_{1-5}^{12} \, \vec{k}^{1-3} \vec{k}^{45} + T_{1-6}^{12} \, \vec{k}^{12} \vec{k}^{34} \vec{k}^{56} / 4 \, where \\ &U_{1-4}^{12} = \sum_{l=1}^{2} \vec{t}_{1-3}^1 \vec{t}_4^2 / 6 + \vec{t}_{12}^1 \vec{t}_3^2 / 4 + \vec{t}_{124}^1 \vec{t}_3^2 / 2 + \vec{t}_{145}^1 \vec{t}_2^2 / 4 ), \\ &T_{1-5}^{12} = \sum_{l=1}^{2} (\vec{t}_{1-4}^1 \vec{t}_5^2 / 6 + \vec{t}_{1245}^1 \vec{t}_3^2 / 4 + \vec{t}_{124}^1 \vec{t}_3^2 / 2 + \vec{t}_{145}^1 \vec{t}_2^2 / 4 ), \\ &T_{1-6}^{12} = \sum_{l=1}^{2} \vec{t}_{1-5}^1 \vec{t}_6^2 / 2 + \sum_{l=1}^{2} \vec{t}_{1235}^1 \vec{t}_4^2 + \vec{t}_{1-3}^1 \vec{t}_4^2 - 2 + \vec{t}_{135}^1 \vec{t}_2^2 \vec{t}_4 ), \\ &Z^{\vec{k}^{1-3}} = \vec{t}_1^1 \vec{t}_2^2 \vec{t}_3^3 \, \vec{k}^{1-3} + T_{1-4}^{1-3} \, \vec{k}^{12} \vec{k}^{34} \, where \, T_{1-4}^{1-3} = \sum_{l=1}^{3} \vec{t}_{135}^1 \vec{t}_2^2 \vec{t}_4^3 \\ &Z^{\vec{k}^{1-3}} = \vec{t}_{1-4}^1 \vec{t}_3^2 \vec{t}_3^2 / 2 + \vec{t}_{1-3}^1 \vec{t}_4^2 \vec{t}_3^2 / 2 + \sum_{l=1}^{6} \vec{t}_{124}^1 \vec{t}_3^2 \vec{t}_3^2 / 2 + \sum_{l=1}^{6} \vec{t}_{124}^1 \vec{t}_3^2 \vec{t}_3^2 / 2 + \sum_{l=1}^{6} \vec{t}_{124}^1 \vec{t}_3^2 \vec{t}_3^2 / 2 + \sum_{l=1}^{6} \vec{t}_{135}^1 \vec{t}_2^2 \vec{t}_3^3 / 2 + \sum_{l=1}^{6} \vec{t}_{14}^1 \vec{t}_2^2 \vec{t}_3^3 / 2 + \sum_{l=1}^{6} \vec{t}_{135}^1 \vec{t}_3^2 \vec{t}_3^2 \vec{t}_3^2 / 2 + \sum_{l=1}^{6} \vec{t}_{135}^1 \vec{t}_2^2 \vec{t}_3^3 / 2 + \sum_{l=1}^{6} \vec{t}_{13}^1 \vec{t}_2^2 \vec{t}_3^3 / 2 + \sum_{l=1}^{6} \vec{t}_{14}^1 \vec{t}_2^2 \vec{t}_3^3 \vec{t}_3^4 / 2 + \sum_{l=1}^{6} \vec{t}_{135}^1 \vec{t}_2^2 \vec{t}_3^3 \vec{t}_3^4 / 2 + \sum_{l=1}^{6} \vec{t}_{13}^1 \vec{t}_2^2 \vec{t$$

$$\begin{split} &+\sum_{12}^{12} \bar{t}_{135}^{1} \bar{t}_{27}^{2} \bar{t}_{48}^{3} \bar{t}_{6}^{4} / 2 + \sum_{13}^{3} \bar{t}_{13}^{1} \bar{t}_{25}^{2} \bar{t}_{47}^{3} \bar{t}_{68}^{4}, \\ &+\bar{K}^{1-5} = \bar{t}_{1}^{1} \cdots \bar{t}_{5}^{5} \bar{k}^{1-5} + T_{1-6}^{1-5} \bar{k}^{1-4} \bar{k}^{56} + U_{1-6}^{1-5} \bar{k}^{1-3} \bar{k}^{4-6} + T_{1-7}^{1-5} \bar{k}^{1-3} \bar{k}^{45} \bar{k}^{67} \\ &+ T_{1-8}^{1-5} \bar{k}^{12} \bar{k}^{34} \bar{k}^{56} \bar{k}^{78} \ where \\ &T_{1-6}^{1-5} = \sum_{20}^{20} \bar{t}_{15}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{4}^{4} \bar{t}_{5}^{5}, U_{1-6}^{1-5} = \sum_{15}^{15} \bar{t}_{14}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{5}^{4} \bar{t}_{5}^{5}, \\ &T_{1-7}^{1-5} = \sum_{30}^{30} \bar{t}_{146}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{5}^{4} \bar{t}_{5}^{5} + \sum_{0}^{60} \bar{t}_{14}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{5}^{4} \bar{t}_{5}^{5}, \\ &T_{1-8}^{1-5} = \sum_{5}^{5} \bar{t}_{1357}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{5}^{4} \bar{t}_{5}^{5} + \sum_{0}^{60} \bar{t}_{135}^{1} \bar{t}_{27}^{2} \bar{t}_{3}^{3} \bar{t}_{5}^{4} \bar{t}_{5}^{5}, \\ &T_{1-8}^{1-5} = \sum_{5}^{5} \bar{t}_{1357}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{5}^{4} \bar{t}_{5}^{5} + \sum_{0}^{60} \bar{t}_{135}^{1} \bar{t}_{27}^{2} \bar{t}_{3}^{3} \bar{t}_{5}^{4} \bar{t}_{5}^{5}, \\ &5 \bar{K}^{1-6} = \bar{t}_{1}^{1} \cdots \bar{t}_{5}^{5} \bar{k}^{1-6} + T_{1-7}^{1-6} \bar{k}^{1-3} \bar{k}^{4-5} \bar{k}^{67} + U_{1-7}^{1-6} \bar{k}^{1-3} \bar{k}^{4-5} \bar{k}^{57} + T_{1-8}^{1-6} \bar{k}^{1-4} \bar{k}^{56} \bar{k}^{78} \\ &+ U_{1-8}^{1-6} \bar{k}^{1-3} \bar{k}^{4-6} \bar{k}^{78} + T_{1-9}^{1-6} \bar{k}^{1-3} \bar{k}^{45} \bar{k}^{67} \bar{k}^{9} + T_{1-10}^{1-6} \bar{k}^{12} \bar{k}^{34} \bar{k}^{56} \bar{k}^{78} \\ &+ U_{1-8}^{1-6} = \sum_{0}^{60} \bar{t}_{16}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{4}^{4} \bar{t}_{5}^{5} \bar{t}_{5}^{7}, U_{1-7}^{1-6} = \sum_{0}^{60} \bar{t}_{15}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{4}^{4} \bar{t}_{5}^{5} \bar{t}_{6}^{6} + \sum_{0}^{16} \bar{t}_{14}^{1} \bar{t}_{27}^{2} \bar{t}_{3}^{3} \bar{t}_{4}^{4} \bar{t}_{5}^{5} \bar{t}_{6}^{6}, \\ &T_{1-8}^{1-6} = \sum_{0}^{60} \bar{t}_{147}^{1} \bar{t}_{2}^{2} \bar{t}_{3}^{3} \bar{t}_{4}^{4} \bar{t}_{5}^{5} \bar{t}_{6}^{6} + \sum_{0}^{16} \bar{t}_{14}^{1} \bar{t}_{27}^{2} \bar{t}_{3}^{3} \bar{t}_{4}^{4} \bar{t}_{5}^{5} \bar{t}_{6}^{6}, \\ &T_{1-7}^{1-6} = \sum_{0}^{60} \bar{t}_{147}^{1}$$

**NOTE 2.1.** For N in  $\Sigma^N$ , see p48 of James and Mayne (1962). The understanding here is that  $\Sigma^N$  in terms like  $T_{1-s}^{1-r}$  only make sense for N < r! in the context where they occur. For example, writing  $(abc) = \overline{t}_{13}^a \overline{t}_2^b \overline{t}_4^c$  and recalling that  $\Sigma^N$  only permutes superscripts but leaves subscripts alone, we have

$$T_{1-4}^{1-3} = \sum_{i=1}^{N} (123) = (123) + (213) + (321)$$
 (4)

with N = 3 not 3! since

$$\sum_{k=1}^{3!} (123) = (123) + (132) + (213) + (231) + (321) + (312) = \sum_{k=1}^{6} S_k$$

say, when multiplied by  $\bar{k}^{12}\bar{k}^{34}$ , as in  $_2\bar{K}^{1-3}$ , gives  $\sum_{k=1}^6 S_k'$  say, where for k=1,2,3,  $S_{2k}'=S_{2k-1}'$ . For example  $T_{1-4}^{1-3}\,\bar{k}^{12}\bar{k}^{34}$  in  $_2\bar{K}^{1-3}$  above is shorthand for  $\sum^3 \bar{t}_{13}^1 \bar{t}_2^2 \bar{t}_4^3\,\bar{k}^{12}\bar{k}^{34}$ . For,

$$S_2' = \bar{t}_4^2 \bar{k}^{43} \bar{t}_{31}^1 \bar{k}^{12} \bar{t}_2^3 = \bar{t}_1^2 \bar{k}^{12} \bar{t}_{13}^1 \bar{k}^{34} \bar{t}_4^3 = S_1' \Rightarrow T_{1-4}^{1-3} \bar{k}^{12} \bar{k}^{34} = S_1' + S_3' + S_5'.$$

PROOF This follows by replacing  $\bar{A}^j_{1-r}=A^j_{i_1\cdots i_r}$  by  $\bar{t}^1_{1-r}/r!=t^{j_1}_{.i_1\cdots i_r}/r!$  in James and Mayne (1962).

Similarly one may easily obtain  $_4\bar{K}^{12}$ ,  $_4\bar{K}^{1-3}$  from p51–53 of James and Mayne (1962). The tensor form  $2\bar{K}_1^1 = \bar{t}_{12}^1\bar{k}_1^{12}$  can be viewed as a *molecule* or *molecular form* of 2 *atoms*,  $\bar{t}_{12}^1$  and  $\bar{k}_1^{12}$ , linked by the *double bond* 1,2, that is,  $i_1, i_2$ .  $_2\bar{K}^1$  is a linear combination of  $\bar{t}_{1-3}^1\bar{k}^{1-3}$ , 2 atoms linked by the triple bond 1,2,3, and secondly  $\bar{k}^{12}$   $\bar{t}_{1-4}^1\bar{k}^{34}$ . The last expression has the structure of  $CO_2$ , with 2 identical atoms each linked by a double bond to a central atom. Just as such bonds are depicted in chemistry to illustrate the

structure of a molecule, they can be very useful here to illustrate the difference in structure of similar mathematical expressions.  $S_1'$  of Note 2.1 is a linear molecular form with the 4 single bonds 1,2,3,4 and 4 distinct atoms,  $\bar{t}_1^2$ ,  $\bar{t}_1^1$ ,  $\bar{t}_{12}^1$ , and  $\bar{k}^{12}$ . Other expressions have more complex structures. Twice the last term in  $_2\bar{K}^{12}$  is  $T_{1-4}^{12}\bar{k}^{12}\bar{k}^{34}=S^{12}+S^{21}+S$  where  $S^{12}=\bar{k}^{12}\,\bar{t}_{1-3}^1\,\bar{k}^{34}\,\bar{t}_4^2$  is linear with a double bond, 1 and 2, then 2 single bonds, 3 and 4;  $S=\bar{t}_{31}^1\,\bar{k}^{12}\,\bar{t}_{24}^2\,\bar{k}^{43}$  forms a square or rectangle with 4 single bonds 1,2,4,3 on successive edges of the square. These pictorial forms are a very useful way to distinguish similar expressions in  $\sum^N f^{j_1j_2\cdots}$ .

§6 provides the 'more complicated' terms referred to (but not given) on p49 of James and Mayne (1962) when  $\hat{w}$  is biased. It can be used for an alternative proof of Theorem 4.1 below. From Theorem 2.1, Edgeworth expansions can be obtained for the distribution and density of the standardized form of  $t(\hat{w})$ ,

$$Y_{nt} = n^{1/2}(\hat{t} - t) = n^{1/2}(t(\hat{w}) - t(w)), \tag{5}$$

of the form

$$Prob.(Y_{nt} \le x) = \sum_{r=0}^{\infty} P_{rn}(x), \ p_{Y_{nt}}(x) = \sum_{r=0}^{\infty} p_{rn}(x),$$
 (6)

where  $P_{rn}(x)$ ,  $p_{rn}(x)$  are  $O(n^{-r/2})$ . The  $e^{\bar{K}^{1-r}}$  of Theorem 2.1 needed for  $P_{rn}(x)$ ,  $p_{rn}(x)$  are as follows.

For 
$$P_{0n}(x)$$
,  $p_{0n}(x)$ :  ${}_{0}\bar{K}^{1} = \bar{t}^{1}$ ,  ${}_{1}\bar{K}^{12}$ . For  $P_{1n}(x)$ ,  $p_{1n}(x)$ :  ${}_{1}\bar{K}^{1}$ ,  ${}_{2}\bar{K}^{1-3}$ . For  $P_{2n}(x)$ ,  $p_{2n}(x)$ :  ${}_{2}\bar{K}^{12}$ ,  ${}_{3}\bar{K}^{1-4}$ . For  $P_{3n}(x)$ ,  $p_{3n}(x)$ :  ${}_{2}\bar{K}^{1}$ ,  ${}_{3}\bar{K}^{1-3}$ ,  ${}_{4}\bar{K}^{1-5}$ . For  $P_{4n}(x)$ ,  $p_{4n}(x)$ :  ${}_{3}\bar{K}^{1-2}$ ,  ${}_{4}\bar{K}^{1-4}$ ,  ${}_{5}\bar{K}^{1-6}$ .

# **3.** Cumulant Coefficients for $t(\hat{w})$ when $E \hat{w} = w$

We now show that for  $r \ge 1$ ,  $1 \le j_1, \ldots, j_r \le q$ ,  $\bar{K}^{1-r}$  of (3) can be expanded as

$$\bar{K}^{1-r} = K^{j_1 - j_r} = \kappa(\hat{t}^{j_1}, \dots, \hat{t}^{j_r}) = \sum_{e=r-1}^{\infty} n^{-e} \bar{K}_e^{1-r}$$
(1)

Replacing  $\{\bar{k}^{1-r}\}$  by  $\{\bar{K}^{1-r}\}$  in the righthand side of (4), written RHS(4), gives the Edgeworth expansion for  $Y_{nt}$  of (5). For  $\pi$  a product of cumulants of type (1), let  $(\pi)_e$  be the coefficient of  $n^{-e}$  in the expansion of  $\pi$ . For example  $(\bar{k}^{1-r})_e = \bar{k}_e^{1-r}$ ,

$$(\bar{k}^{12}\bar{k}^{34})_3 = \bar{k}_1^{12}\bar{k}_2^{34} + \bar{k}_2^{12}\bar{k}_1^{34}, (\bar{k}^{12}\bar{k}^{34})_4 = \bar{k}_1^{12}\bar{k}_3^{34} + \bar{k}_2^{12}\bar{k}_2^{34} + \bar{k}_3^{12}\bar{k}_1^{34}, (\bar{k}^{1-3}\bar{k}^{45})_4 = \bar{k}_2^{1-3}\bar{k}_2^{45} + \bar{k}_3^{1-3}\bar{k}_1^{45}, (\bar{k}^{12}\bar{k}^{34}\bar{k}^{56})_4 = \bar{k}_1^{12}\bar{k}_1^{34}\bar{k}_2^{56} + \bar{k}_1^{12}\bar{k}_2^{34}\bar{k}_1^{56} + \bar{k}_2^{12}\bar{k}_1^{34}\bar{k}_1^{56}.$$
(2)

We now give the elements of the expansion (1) when  $E \hat{w} = w$ .

**Theorem 2.** Suppose that  $\hat{w}$  is an unbiased estimate of w satisfying (1) and that t(w) has finite derivatives. Then (1) holds with bounded cumulant coefficients

$$\bar{K}_e^{1-r} = K_e^{j_1 - j_r} = \sum_{k=r-1}^e {}_k \bar{K}_e^{1-r} : 
\bar{K}_{r-1}^{1-r} = {}_{r-1} \bar{K}_{r-1}^{1-r}, \ \bar{K}_r^{1-r} = {}_{r-1} \bar{K}_r^{1-r} + {}_r \bar{K}_r^{1-r}, \dots$$
(3)

The leading coefficients needed for  $P_r(x)$ ,  $p_r(x)$  of (4) for the distribution of  $Y_{nt}$  of (5) are given in the T, U, V notation of Theorem 2.1 as follows.

$$\begin{split} & \vec{R}_0^1 = \vec{l}^1, \, that \, is, \, K_0^{j_1} = t^{j_1} = t^{j_1} (w). \, _0 \vec{k}_c^1 = 0 \, for \, e \geq 1. \\ & For \, P_0(x) : \, \vec{k}_1^{12} = i \vec{k}_1^{12} = \vec{l}_1^1 \vec{l}_2^2 \, \vec{k}_1^{12}, \, that \, is, \, K_1^{j_1j} = t^{j_1}_{.j_1} t^{j_2}_{.j_1} k^{j_1j_2}_{1}. \\ & For \, P_1(x) : \, \vec{k}_1^1 = i \vec{k}_1^1 = \vec{l}_1^1 \vec{l}_2^1 \vec{k}_1^{12}/2, \, that \, is, \, K_1^{j_1} = t^{j_1}_{.j_1} k^{j_1j_2}/2, \\ & \vec{k}_2^{1-3} = 2 \vec{k}_2^{1-3} = \vec{l}_1^1 \vec{l}_2^2 \vec{l}_3^2 \vec{k}_2^{1-3} + T_{1-3}^{1-3} \, \vec{k}_1^{12} \vec{k}_1^{34}. \\ & For \, P_2(x) : \, \vec{k}_2^{12} = 1 \vec{k}_2^{12} + 2 \vec{k}_2^{12} \, for \, _1 \vec{k}_2^{12} = \vec{l}_1^{12} \vec{k}_2^{12}, \\ & \vec{k}_3^{1-4} = 3 \vec{k}_3^{1-4} = (\vec{l}_1 \cdots \vec{l}_4^4) \, \vec{k}_3^{1-4} + T_{1-5}^{1-4} \, \vec{k}_2^{1-3} \vec{k}_1^{45} + T_{1-6}^{1-4} \, \vec{k}_1^{12} \vec{k}_1^{34} \vec{k}_1^{56}. \\ & For \, P_3(x) : \, \vec{k}_2^1 = 1 \vec{k}_2^1 + 2 \vec{k}_2^1 \, for \, _1 \vec{k}_2^1 = \vec{l}_1^1 \vec{k}_2^{12}/2, \, _2 \vec{k}_2^1 = \vec{l}_{1-5}^1 \vec{k}_2^{1-3}/6 \\ & + \vec{l}_{1-4}^1 \, \vec{k}_1^{12} \vec{k}_1^{34}/8, \, that \, is, \, K_2^{j_1} = t^{j_1}_{11} \, x^{j_2}_2 x^{j_2}/2 + t^{j_1}_{11} \, x^{j_2}_1 x^{j_2}_3 x^{j_3}/6 + t^{j_1}_{11-4} \, k^{j_1i_2}_1 k^{j_1i_2}_1 k^{j_3i_4}/8, \\ & \vec{k}_3^{1-3} = \sum_{k=2}^3 \, k \vec{k}_3^{3-3} \, for \, _2 \vec{k}_3^{3-3} = \vec{l}_1^1 \vec{l}_2^2 \vec{k}_3^2 \, \vec{k}_3^{1-3} + T_{1-3}^{1-3} \, \vec{k}_1^{12} \vec{k}_1^{34} \vec{k}_1^{56}, \\ & \vec{k}_4^{1-5} = 4 \vec{k}_4^{1-5} = \vec{l}_1^1 \cdots \vec{l}_5^5 \, \vec{k}_4^{1-5} + T_{1-5}^{1-5} \, \vec{k}_3^{1-4} \vec{k}_1^{15} + U_{1-6}^{1-5} \, \vec{k}_2^{1-3} \vec{k}_2^{15} - 4 \\ & + T_{1-7}^{1-7} \, \vec{k}_2^{1-3} \, \vec{k}_1^{15} \vec{k}_1^{17} + T_{1-8}^{1-5} \, \vec{k}_1^{12} \vec{k}_1^{34} \vec{k}_1^{56} \vec{k}_1^{10}. \\ & For \, P_4(x) : \, \vec{k}_3^{12} = \sum_{3}^3 \, k \vec{k}_3^{3} \, for \, _1 \vec{k}_3^{12} = \vec{l}_1^1 \vec{l}_2^2 \, k_3^{13}, \\ & 2 \vec{k}_3^{12} = T_{1-3}^{12} \, \vec{k}_3^{1-3} /2 + T_{1-4}^{12} \, (\vec{k}_1^{12} \vec{k}_1^{34} \vec{k}_1^{56} \vec{k}_1^{10}, \\ & For \, P_4(x) : \, \vec{k}_3^{13} = \sum_{3}^3 \, k \vec{k}_3^{3} \, for \, _1 \vec{k}_3^{12} = \vec{l}_1^1 \vec{l}_2^2 \, \vec{k}_3^{13}, \\ & 2 \vec{k}_3^{13} = T_{1-3}^{12} \, \vec{k}_3^{1-4} + T_{1-5}^{12} \, \vec{k}_1^{1-3$$

$$\begin{split} \bar{K}_{3}^{1} &= \sum_{k=1}^{3} {}_{k} \bar{K}_{3}^{1} \, for \, {}_{1} \bar{K}_{3}^{1} = \bar{t}_{12}^{1} \, \bar{k}_{3}^{12}/2, \, {}_{2} \bar{K}_{3}^{1} = \bar{t}_{1-3}^{1} \, \bar{k}_{3}^{1-3}/6 + \bar{t}_{1-4}^{1} \, \bar{k}_{1}^{12} \bar{k}_{2}^{34}/4, \\ 3 \bar{K}_{3}^{1} &= \bar{t}_{1-4}^{1} \, \bar{k}_{3}^{1-4}/24 + \bar{t}_{1-5}^{1} \, \bar{k}_{2}^{1-3} \bar{k}_{1}^{45}/12 + \bar{t}_{1-6}^{1} \, \bar{k}_{1}^{12} \bar{k}_{1}^{34} \bar{k}_{1}^{56}/48, \\ \bar{K}_{4}^{1} &= \sum_{k=1}^{4} {}_{k} \bar{K}_{4}^{1} \, for \, {}_{1} \bar{K}_{4}^{1} &= \bar{t}_{12}^{1} \bar{k}_{4}^{12}/2, \\ 2 \bar{K}_{4}^{1} &= \bar{t}_{1-3}^{1} \bar{k}_{4}^{1-3}/6 + \bar{t}_{1-4}^{1} \, (2 \bar{k}_{1}^{12} \bar{k}_{3}^{34} + \bar{k}_{2}^{12} \bar{k}_{2}^{34})/8, \\ 3 \bar{K}_{4}^{1} &= \bar{t}_{1-4}^{1} \, \bar{k}_{4}^{1-4}/24 + \bar{t}_{1-5}^{1} \, (\bar{k}_{2}^{1-3} \bar{k}_{2}^{45} + \bar{k}_{3}^{1-3} \bar{k}_{1}^{45})/12 + \bar{t}_{1-6}^{1} \, \bar{k}_{1}^{12} \bar{k}_{1}^{34} \bar{k}_{2}^{56}/16, \\ 4 \bar{K}_{4}^{1} &= \bar{t}_{1-5}^{1} \, \bar{k}_{4}^{1-5}/120 + \bar{t}_{1-6}^{1} \, (\bar{k}_{3}^{1-4} \bar{k}_{1}^{56}/48 + \bar{k}_{2}^{1-3} \bar{k}_{2}^{4-6}/72) + \bar{t}_{1-7}^{1} \, \bar{k}_{2}^{1-3} \bar{k}_{1}^{45} \bar{k}_{1}^{67}/48 \\ &+ \bar{t}_{1-8}^{1} \, \bar{k}_{1}^{12} \bar{k}_{1}^{34} \bar{k}_{1}^{56} \bar{k}_{1}^{78}/384. \end{split}$$

PROOF Substituting (1) into 
$$_{k}\bar{K}^{1-r}$$
 of Theorem 2.1 gives  $_{k}\bar{K}^{1-r} = \sum_{e=k}^{\infty} {_{k}\bar{K}_{e}^{1-r}} n^{-e}$  say. So by (3), (1) and (3) hold.  $_{k}\bar{K}_{e}^{1-r} = {_{k}K_{e}^{j_{1}\cdots j_{r}}}$  is  $_{k}\tilde{V}_{e}^{j_{1}\cdots j_{r}}$  of Withers (1982).

**NOTE 3.1.** (4) made explicit the 3 terms needed in  $T_{1-4}^{1-3}$  needed for  $P_1(x)$  of Theorem 3.1. Similarly  $P_2(x)$  needs the 12 terms

$$T_{1-5}^{1-4} = \sum_{-5}^{12} (1234) = (1234) + (1243) + (2413) + (2431) + (3124) + (3142) + (3241) + (3412) + (4123) + (4132) + (4231) + (4321)$$

where  $(abcd) = \bar{t}_{14}^a \bar{t}_2^b \bar{t}_3^c \bar{t}_5^d$ . It also needs the 4 + 12 terms  $T_{1-6}^{1-4} = A + B$  where

$$A = \sum_{12}^{4} (1234) = (1234) + (2134) + (3124) + (4123) \text{ for } (abcd) = \bar{t}_{135}^a \bar{t}_2^b \bar{t}_4^c \bar{t}_6^d,$$

$$B = \sum_{12}^{4} (1234) = (1234) + (1423) + (1432) + (1324) + (2134) + (2314) + (2413) + (3124) + (3214) + (3412) + (4213) + (4312) \text{ for } (abcd) = \bar{t}_{13}^a \bar{t}_{25}^b \bar{t}_4^c \bar{t}_6^d.$$

# **4.** Cumulant Coefficients for $t(\hat{w})$ when $E \hat{w} \neq w$

We now remove the assumption that  $\hat{w}$  is unbiased. We use  $\bar{K}_e^{1-r}$  of Theorem 3.1, and the shorthand  $\bar{f}_{.m} = \partial_{i_m} f$  where again  $\partial_i = \partial/\partial w^i$ . There is a key difference with Theorem 3.1: there  $\bar{k}_e^{1-r}$  was treated as an algebraic expression. But now we must view each of them as a function of w. So we assume that the distribution of  $\hat{w}$  is determined by w. This is needed to obtain higher order confidence intervals for t(w) when q=1: see Withers (1989). We show that for  $Y_{nt}$  of (5),  $P_2(x)$ ,  $p_2(x)$  need the 1st derivatives  $\bar{k}_{1.i}^{12} = \partial_i \bar{k}_1^{12}$  for  $\partial_i = \partial/\partial w^i$ ,  $P_3(x)$ ,  $p_3(x)$  need the 1st derivatives  $\bar{k}_{2.4}^{1-3}$ , and so on. The derivatives of  $\bar{k}_e^{1-r}$  are given by Leibniz's rule for the derivatives of a product. For example

$$\begin{split} \vec{K}_{1.3}^{12} &= (\vec{t}_1^1 \vec{t}_2^2 \, \vec{k}_1^{12})_{.3} = (\sum_{12}^2 \vec{t}_1^1 \vec{t}_{23}^2) \, \vec{k}_1^{12} + \vec{t}_1^1 \vec{t}_2^2 \, \vec{k}_{1.3}^{12} \, \text{for} \, \sum_{12}^2 \vec{t}_1^1 \vec{t}_{23}^2 = \vec{t}_{13}^1 \vec{t}_2^2 + \vec{t}_1^1 \vec{t}_{23}^2, \\ \vec{K}_{1.3}^1 &= (\vec{t}_{12}^1 \vec{k}_1^{12})_{.3}/2 = \vec{t}_{1-3}^1 \vec{k}_1^{12}/2 + \vec{t}_{12}^1 \vec{k}_{1.3}^{12}/2, \\ \vec{K}_{1.34}^{12} &= \sum_{12}^2 [(\vec{t}_{14}^1 \vec{t}_{23}^2 + \vec{t}_1^1 \vec{t}_{2-4}^2) \vec{k}_1^{12} + \vec{t}_1^1 \vec{t}_{23}^2 \vec{k}_{1.4}^{12} + \vec{t}_{14}^1 \vec{t}_2^2 \vec{k}_{1.3}^{12}] + \vec{t}_1^1 \vec{t}_2^2 \vec{k}_{1.34}^{12}, \\ (\vec{t}_1^1 \vec{t}_2^2 \vec{t}_3^3 \, \vec{k}_2^{1-3})_{.4} &= (\vec{t}_1^1 \vec{t}_2^2 \vec{t}_3^3)_{.4} \vec{k}_2^{1-3} + \vec{t}_1^1 \vec{t}_2^2 \vec{t}_3^3 \, \vec{k}_{2.4}^{1-3}, \, (\vec{t}_1^1 \vec{t}_2^2 \vec{t}_3^3)_{.4} = \vec{t}_1^1 \vec{t}_2^2 \vec{t}_3^3 + \vec{t}_1^1 \vec{t}_2^2 \vec{t}_3^3, \\ (T_{1-4}^{1-3} \, \vec{k}_1^{12} \vec{k}_1^{34})_{.5} &= T_{1-4.5}^{1-3} \, \vec{k}_1^{12} \vec{k}_1^{34} + T_{1-4}^{1-3} \, (\vec{k}_1^{12} \vec{k}_1^{34})_{.5}, \\ T_{1-4}^{1-3} &= \sum_{13}^3 (\vec{t}_{135}^1 \vec{t}_2^2 \vec{t}_3^2 + \vec{t}_{13}^1 \vec{t}_2^2 \vec{t}_3^3 + \vec{t}_{13}^1 \vec{t}_2^2 \vec{t}_3^3), \, (\vec{k}_1^{12} \vec{k}_1^{34})_{.5} &= \vec{k}_{15}^{12} \vec{k}_1^{34} + \vec{k}_1^{12} \vec{k}_1^{34}. \end{split}$$

**Theorem 3.** Let  $\hat{w} \in R^p$  be a biased standard estimate of w satisfying (1) where  $\bar{k}_e^{1-r}$  depend on w. Then  $\hat{t} = t(\hat{w}) \in R^q$  is a standard estimate of t(w):

$$\kappa(\hat{t}^{j_1}, \dots, \hat{t}^{j_r}) = \sum_{e=r-1}^{\infty} n^{-e} \, \bar{a}_e^{1-r} \, \text{for } r \ge 1, \ 1 \le j_1, \ \dots, \ j_r \le q, \tag{1}$$

where 
$$\bar{a}_e^{1-r} = \bar{K}_e^{1-r} + \bar{D}_e^{1-r}$$
,  $\bar{D}_{r-1}^{1-r} = 0$ , (2)

for  $\bar{K}_e^{1-r}$  of Theorem 3.1, and the other  $\bar{D}_e^{1-r} = D_e^{j_1...j_r}$  needed for  $P_r(x)$ ,  $p_r(x)$  of (4) for  $Y_{nt}$  of (5) are as follows.

$$\begin{split} &For \ P_0(x): \ \bar{D}_1^{12} = 0 \Rightarrow \bar{a}_1^{12} = \bar{K}_1^{12} = K_1^{j_1 j_2} = \bar{t}_1^1 \bar{t}_2^2 \ \bar{k}_1^{12}, \\ &For \ P_1(x): \bar{D}_1^1 = \bar{t}_1^1 \bar{k}_1^1 \Rightarrow \bar{a}_1^1 = \bar{K}_1^1 + \bar{D}_1^1 = \bar{t}_1^1 \bar{k}_1^1 + \bar{t}_{12}^1 \bar{k}_1^{12}/2, \\ &For \ P_2(x): \ \bar{D}_2^{12} = \bar{K}_{1.3}^{12} \ \bar{k}_1^3 = [(\bar{t}_{13}^1 \bar{t}_2^2 + \bar{t}_1^1 \bar{t}_{23}^2) \ \bar{k}_1^{12} + \bar{t}_1^1 \bar{t}_2^2 \ \bar{k}_{1.3}^{12}] \ \bar{k}_1^3 \\ &\Rightarrow \bar{a}_2^{12} = \bar{t}_1^1 \bar{t}_2^2 \ \bar{k}_2^{12} + T_{1-3}^{12} \ \bar{k}_2^{1-3}/2 + T_{1-4}^{12} \ \bar{k}_1^{12} \bar{k}_1^{34}/2 \\ &+ [(\bar{t}_{13}^1 \bar{t}_2^2 + \bar{t}_1^1 \bar{t}_{23}^2) \ \bar{k}_1^{12} + \bar{t}_1^1 \bar{t}_2^2 \ \bar{k}_{1.3}^{12}] \ \bar{k}_1^3. \\ &For \ P_3(x): \ \bar{D}_2^1 = \bar{K}_{1,1}^1 + \bar{K}_{0,2}^1, \ \bar{K}_{1,1}^1 = \bar{K}_{1.3}^1 \ \bar{k}_1^3, \ \bar{K}_{0,2}^1 = \bar{t}_1^1 \bar{k}_2^1 + \bar{t}_{12}^1 \bar{k}_1^1 \bar{k}_1^2/2 \Rightarrow \\ \bar{a}_2^1 = \bar{t}_1^1 \bar{k}_2^1 + \bar{t}_{12}^1 (\bar{k}_2^{12} + \bar{k}_1^1 \bar{k}_1^2 + \bar{k}_{1.3}^1 \bar{k}_1^3)/2 + \bar{t}_{1-3}^1 (\bar{k}_2^{1-3}/6 + \bar{k}_1^1 \bar{k}_1^{23}/2) + \bar{t}_{1-4}^1 \bar{k}_1^{12} \bar{k}_1^{34}/8, \\ \bar{D}_3^{1-3} = \bar{K}_{2.4}^{1-3} \ \bar{k}_1^4 = (\bar{t}_1^1 \bar{t}_2^2 \bar{t}_3^3 \ \bar{k}_2^{1-3}).4 \ \bar{k}_1^4 + (T_{1-3}^{1-3} \ \bar{k}_1^{12} \bar{k}_1^{34}).5 \ \bar{k}_1^5. \end{split}$$

For 
$$P_4(x)$$
:  $\bar{D}_3^{12} = \bar{K}_{2,1}^{12} + \bar{K}_{1,2}^{12}$ ,  $\bar{K}_{2,1}^{12} = \bar{K}_{2,3}^{12} \, \bar{k}_1^3$ ,  $\bar{K}_{1,2}^{12} = \bar{K}_{1,3}^{12} \, \bar{k}_2^3/2 + \bar{K}_{1,34}^{12} \, \bar{k}_1^3 \bar{k}_1^4$ ,  $\bar{D}_4^{1-4} = \bar{K}_{3,5}^{1-4} \, \bar{k}_1^5$ ,  $\bar{D}_5^{1-6} = 0$ .

For E  $t^{j_1}(\hat{w})$  to  $O(n^{-5})$  we also need  $\bar{D}_j = \bar{D}^1_j$ , j = 3, 4, given by

$$\begin{split} \bar{D}_3 &= \bar{K}_{2,1} + \bar{K}_{1,2} + \bar{K}_{0,3}, \ \bar{K}_{2,1} &= \bar{K}_{2,1} \bar{k}_1^1, \\ \bar{K}_{2,1} &= (\bar{t}_{1-3} \bar{k}_2^{23} + \bar{t}_{23} \bar{k}_{2,1}^{23})/2 + (\bar{t}_{1-4} \bar{k}_2^{2-4} + \bar{t}_{2-4} \bar{k}_{2,1}^{2-4})/6 + \bar{t}_{1-5} \bar{k}_1^{23} \bar{k}_1^{45}/8 + \bar{t}_{2-5} \bar{k}_1^{23} \bar{k}_{1,1}^{45}/4, \\ \bar{K}_{1,2} &= \bar{K}_{1,1} \bar{k}_2^1 + \bar{K}_{1,12} \bar{k}_1^1 \bar{k}_1^2/2, \\ 2\bar{K}_{1,1} &= \bar{t}_{1-3} \bar{k}_1^{23} + \bar{t}_{23} \bar{k}_{1,1}^{23}, \ 2\bar{K}_{1,12} &= \bar{t}_{1-4} \bar{k}_1^{34} + \sum_{12}^2 \bar{t}_{2-4} \bar{k}_{1,1}^{34} + \bar{t}_{34} \bar{k}_{1,12}^{34}, \\ \bar{K}_{0,3} &= \bar{t}_1 \bar{k}_3^1 + \bar{t}_{12} \bar{k}_1^1 \bar{k}_2^2 + \bar{t}_{1-3} \bar{k}_1^1 \bar{t}_1^2 \bar{k}_1^3/6, \\ \bar{D}_4 &= \bar{K}_{3,1} + \bar{K}_{2,2} + \bar{K}_{1,3} + \bar{K}_{0,4}, \ \bar{K}_{3,1} &= \bar{K}_{3,1} \bar{k}_1^1, \\ \bar{K}_{3,1} &= \bar{t}_{1-3} \bar{k}_3^{23}/2 + \bar{t}_{23} \bar{k}_{3,1}^{23}/2 + \bar{t}_{1-4} \bar{k}_3^{2-4}/6 + \bar{t}_{2-4} \bar{k}_{3,1}^{2-4}/6 + \bar{t}_{1-5} \bar{t}_1^{23} \bar{k}_2^{45}/4 \\ &+ \bar{t}_{2-5} \bar{k}_1^{23} \bar{k}_2^{45}/2 + (\bar{t}_{1-5} \bar{k}_3^{2-5} + \bar{t}_{2-5} \bar{k}_{3,1}^{2-5})/24 + (\bar{t}_{1-6} \bar{k}_2^{2-4} \bar{k}_1^{56} + \bar{t}_{2-6} \bar{k}_{2,1}^{2-4} \bar{k}_1^{56} \\ &+ \bar{t}_{2-6} \bar{k}_2^{2-4} \bar{k}_{1,0}^{56}/12 + \bar{k}_1^{23} \bar{k}_1^{45} (\bar{t}_{1-7} \bar{k}_1^{67}/48 + \bar{t}_{2-7} \bar{k}_{1,1}^{67}/16), \\ \bar{K}_{2,2} &= \bar{K}_{2,1} \bar{k}_2^1 + \bar{K}_{2,12} \bar{k}_1^1 \bar{k}_1^2/2, \\ \bar{K}_{2,1} &= \bar{t}_{1-3} \bar{k}_2^{23}/2 + \bar{t}_{23} \bar{k}_{2,1}^{23}/2 + \bar{t}_{1-4} \bar{k}_2^{2-4}/6 + \bar{t}_{2-4} \bar{k}_{2,1}^{2-4}/6 + \bar{t}_{1-5} \bar{k}_1^{23} \bar{k}_1^{45}/8 + \bar{t}_{2-5} \bar{k}_1^{23} \bar{k}_1^{45}/4, \\ 2\bar{K}_{2,12} &= \bar{t}_{1-4} \bar{k}_2^{34} + \sum_{12}^2 \bar{t}_{13-6} \bar{k}_1^{34} \bar{k}_{1,2}^{56}/2 + \bar{t}_{3-6} (\bar{k}_{1,1}^{34} \bar{k}_{1,1}^{56} + \bar{k}_1^{34} \bar{k}_{1,2}^{56})/2, \\ \bar{K}_{1,3} &= \bar{K}_{1,1} \bar{k}_3^1 + \bar{K}_{1,12} \bar{k}_1^1 \bar{k}_2^2 + \bar{K}_{1,123} \bar{k}_1^1 \bar{k}_1^2 \bar{k}_1^3/6, \\ 2\bar{K}_{1,1} &= \bar{t}_{1-3} \bar{k}_1^{23} + \bar{t}_{23} \bar{k}_{1,1}^{23}, \ 2\bar{K}_{1,12} &= \bar{t}_{1-4} \bar{k}_1^{34} + \sum_{12}^2 \bar{t}_{134} \bar{k}_{1,2}^{34} + \bar{t}_{34} \bar{k}_{1,12}^{34}, \\ 2\bar{K}_{1,123} &= \bar{t}_{1-5} \bar{k}_1^{45} + \bar{t}_{134}$$

PROOF 
$$\bar{K}^{1-r}(w) = \bar{K}^{1-r}$$
 and  $\bar{K}^{1-r}_{\ell}(w) = \bar{K}^{1-r}_{\ell}$  are functions of  $w$ . By (1)

 $\bar{K}_{0.4} = \bar{t}_1 \bar{k}_1^1 + \bar{t}_{12} (\bar{k}_1^1 \bar{k}_3^2 + \bar{k}_2^1 \bar{k}_2^2 / 2) + \bar{t}_{1-3} \bar{k}_1^1 \bar{k}_2^1 \bar{k}_2^3 / 2 + \bar{t}_{1-4} \bar{k}_1^1 \bar{k}_1^2 \bar{k}_3^3 \bar{k}_1^4 / 24.$ 

$$\bar{K}^{1-r}(w_n) = \sum_{e=r-1}^{\infty} n^{-e} \bar{K}_e^{1-r}(w_n) \text{ for } w_n = E \hat{w} = w + d_n,$$

where by (1),  $d_n$  has  $i_1$ th component  $\bar{d}_n^1 = d_n^{i_1} = \sum_{e=1}^{\infty} n^{-e} \bar{k}_e^1$ . Consider the Taylor series expansion

$$\bar{K}_{k}^{1-r}(w+d_{n}) = \bar{K}_{k}^{1-r} + \bar{K}_{k,1}^{1-r} \, \bar{d}_{n}^{1} + \bar{K}_{k,12}^{1-r} \, \bar{d}_{n}^{1} \bar{d}_{n}^{2} / 2! + \dots = \sum_{e=0}^{\infty} \bar{K}_{k,e}^{1-r} \, n^{-e}$$

say. Substituting into (1) gives (1) with

$$\bar{a}_c^{1-r} = \sum_{k+e=c} \bar{K}_{k,e}^{1-r} = \sum_{e=0}^{c-r+1} \bar{K}_{c-e,e}^{1-r}.$$
 (3)

Also  $\bar{K}_{k,0}^{1-r} = \bar{K}_k^{1-r}$  so that (2) holds with

$$\bar{D}_{c}^{1-r} = \sum_{e=1}^{c-r+1} \bar{K}_{c-e,e}^{1-r} :$$

$$\bar{D}_{r}^{1-r} = \bar{K}_{r-1,1}^{1-r}, \ \bar{D}_{r+1}^{1-r} = \sum_{e=1}^{2} \bar{K}_{r+1-e,e}^{1-r}, \dots \qquad \Box$$

$$(4)$$

An alternative proof can be obtained using §6. This corrects  $C_e = \bar{a}_e^1$  given in Appendix B of Withers (1987). Withers (1982) uses  $K_k^{j_1\cdots j_r}=\bar{K}_k^{1-r}$  for  $\bar{a}_k^{1-r}$  but the expression for  $K_2^{ab}$  on p67, lines 2-3 omitted the term  $A_i^a A_j^b k_{1,k}^{ij} k_1^k$ . That is, the last term in  $\bar{a}_2^{12}$  of Theorem 4.1 was omitted. Similarly the results on p67 for r=3,4 are only true when the  $\hat{w}$  is unbiased or the cumulant coefficients of  $\hat{w}$  do not depend on w, as they omit the derivatives of  $\bar{k}_e^{1-r}$ . The examples given there are not affected as  $\hat{w}$  is unbiased. Nor are the nonparametric examples of Withers (1983, 1988) affected, as the empirical distribution is an unbiased estimate of a distribution. Likewise  $\hat{w}$  is unbiased for the examples of Withers (1989). M-estimates are biased but the results of Withers and Nadarajah (2010a) are not affected as only  $K_1^{j_1j_2}$ ,  $K_1^{j_1}$ ,  $K_2^{j_1j_2j_3}$  are given. No changes are needed for Withers and Nadarajah (2010b, 2011a, 2011b, 2012a, 2012b, 2014b). Applications to non-parametric and parametric confidence intervals were given in Withers (1983, 1989, 1989) and to ellipsoidal confidence regions and power in Withers and Nadarajah (2012a) and Kakizawa (2015). For nonparametric problems, F(x) and its empirical distribution  $F_n(x)$ play the role of w and  $\hat{w}$ ; since it is unbiased, no corrections are needed. For q=1,  $a_{ri}=\bar{a}_i^{1-r}$  were given for parametric and non-parametric problems in Withers (1982, 1983, 1988), and expressions for the classic Edgeworth expansion of  $Y_{nw}$  in terms of  $a_{ri}$  were given in Withers (1984). For  $q \ge 1$ ,  $\bar{a}_i^{1-r}$  for parametric problems were given in Withers (1982), and can be obtained easily from  $a_{ri}$  given when q = 1 for 1 sample and multi sample non-parametric problems in Withers (1983, 1988), and for semi-parametric problems in Withers and Nadarajah (2010a, 2011a, 2014b). All these results can be extended to samples with independent non-identically distributed residuals, as done in Withers and Nadarajah (2010 §6, 2011b, 2012b). The extension to matrix  $\hat{w}$  just needs a slight change in notation. For example in Withers and Nadarajah (2011b, 2011c, 2012b),  $\hat{w}$  can be viewed as a function of the mean of *n* independent *complex random matrices*, although *n* is actually the number of transmitters or receivers. Extensions to dependent random variables are also possible: see Withers and Nadarajah (2012c).

# **5.** Cumulant Coefficients for Univariate $t(\hat{w})$

Now **suppose that** q = 1. Let  ${}_kK_{re}$  be the coefficient of  $n^{-e}$  in  ${}_kK^{1-r}$ . We write  $\bar{K}_e^{1-r}$  as  $K_{re}$ . For  $E \hat{w} = w$ , (1), (3) and (4) become

$$K^{r} = \kappa_{r}(\hat{t}) = \sum_{e=r-1}^{\infty} n^{-e} K_{re}, \ r \ge 1; \ K_{re} = \sum_{k=r-1}^{e} {}_{k} K_{re} :$$

$$K_{r,r-1} = {}_{r-1} K_{r,r-1}, \ K_{rr} = \sum_{k=r-1}^{r} {}_{k} K_{rr}, \ K_{r,r+1} = \sum_{k=r-1}^{r+1} {}_{k} K_{r,r+1}, \cdots$$

$$(1)$$

For  $E \hat{w} \neq w$ , (1), (2) and (3) become

$$K^{r} = \kappa_{r}(\hat{t}) = \sum_{e=r-1}^{\infty} n^{-e} a_{re}, \ r \ge 1; \ a_{re} = K_{re} + D_{re}, \ D_{rc} = \sum_{e=1}^{c-r+1} K_{r,c-e,e} :$$

$$D_{r,r-1} = 0, \ D_{rr} = K_{r,r-1,1}, \ D_{r,r+1} = \sum_{e=1}^{2} K_{r,r+1-e,e}, \cdots$$

Here we give the cumulant coefficients  $K_{re}$  needed for the Edgeworth expansion of  $Y_{nt}$  of (5) for  $P_r(x)$ ,  $r \le 4$ . We do this when  $E \hat{w} = w$  in Corollary 5.1 and when  $E \hat{w} \ne w$  in Corollaries 5.3 and 5.4. To show more clearly the expressions we need in molecular form, we introduce the following *ions*, (expressions with unpaired suffixes),

$$s^{i_{1}} = \bar{s}^{1} = \bar{k}_{1}^{12}\bar{t}_{2}, \ \bar{u}_{1} = \bar{t}_{12}\bar{s}^{2} = \bar{t}_{12}\bar{k}_{1}^{23}\bar{t}_{3}, \ \bar{X}^{34} = \bar{k}_{1}^{31}\bar{t}_{12}\bar{k}_{1}^{24}, \ \bar{z}_{12} = \bar{t}_{1-3}\bar{s}^{3}, \\ \bar{v}^{1} = \bar{k}_{1}^{12}\bar{u}_{2} = \bar{X}^{14}\bar{t}_{4}, \ \bar{x}_{1} = \bar{t}_{12}\bar{v}^{2}, \ \bar{S}^{1} = \bar{k}_{1}^{22}\bar{t}_{2}, \ \bar{y}^{1} = \bar{k}_{2}^{1-3}\bar{t}_{2}\bar{t}_{3}, \ \bar{Y}_{1} = \bar{t}_{12}\bar{y}^{2}.$$
 (2)

Where a suffix does not have a match then summation does not occur. For example the RHS of  $\bar{s}^1 = \bar{k}_1^{12}\bar{t}_2$  sums over  $i_2$  but not  $i_1$ . Let  $v, c_{01}, c_{02}, c_{21}, c_{22}, c_{23}, c_{11}, \cdots, c_{1,10}, c_{31}, \cdots, c_{3,11}$  be the 27 functions of  $\omega$  given on p4234–4235 of Withers (1989), labelled there as  $I_2({}^2_0)$ ,  $I_1({}^1_0)$ ,  $\cdots$ ,  $I_{301}({}^{222}_{000})$ . By Corollaries 5.1, 5.3 below, those needed for  $P_r(x)$ ,  $r \leq 2$ , of (4), that is, for the Edgeworth expansion of  $Y_{nt}$  of (5) to  $O(n^{-3/2})$ , are the following molecules.

For 
$$P_0(x): v = K_{21} = \bar{t}_1 \bar{k}_1^{12} \bar{t}_2$$
.  
For  $P_1(x)$ ,  $K_{11}: c_{02} = \bar{t}_{12} \bar{k}_1^{12}$ ;  $D_{11}: c_{01} = \bar{t}_1 \bar{k}_1^{1}$ ; for  $K_{32}: c_{21} = \bar{t}_1 \bar{t}_2 \bar{t}_3 \bar{k}_2^{1-3} = \bar{t}_1 \bar{y}^1$ ,  $c_{23} = \bar{s}^1 \bar{t}_{12} \bar{s}^2 = \bar{s}^1 \bar{u}_1$ .  
For  $P_2(x)$ ,  $K_{22}: c_{11} = \bar{t}_1 \bar{k}_2^{12} \bar{t}_2 = \bar{t}_1 \bar{s}^1$ ,  $c_{15} = \bar{t}_1 \bar{k}_2^{1-3} \bar{t}_{23}$ ,  $c_{19} = \bar{t}_{12} \bar{X}^{12}$ ,  $c_{1,10} = \bar{s}^1 \bar{t}_{1-3} \bar{k}_1^{23} = \bar{z}_{23} \bar{k}_1^{23}$ ; for  $D_{22}: c_{12} = \bar{k}_1^1 \bar{k}_{1,1}^{23} \bar{t}_2 \bar{t}_3$ ,  $c_{16} = \bar{k}_1^1 \bar{u}_1 = \bar{k}_1^1 \bar{t}_{.12} \bar{k}_1^{23} \bar{t}_3$ , for  $K_{43}: c_{31} = \bar{t}_1 \bar{t}_2 \bar{t}_3 \bar{t}_4 \bar{k}_3^{1-4}$ ,  $c_{36} = \bar{y}^3 \bar{u}_3$ ,  $c_{3,10} = \bar{u}_1 \bar{k}_1^{12} \bar{u}_2$ ,  $c_{3,11} = \bar{s}^1 \bar{s}^2 \bar{s}^3 \bar{t}_{1-3}$ .

Each molecule can be written as a shape. For example  $c_{19}$  is a rectangle. We now give the molecules  $L_j$ ,  $L_{ij}$  needed for the Edgeworth expansion to  $O(n^{-5/2})$ , that is, for  $P_r(x)$  for r = 3, 4. Note that  $P_r(x)$  needs the derivatives of t(w) up to order r + 1.

For 
$$P_3(x)$$
,  $K_{12}: L_1 = \bar{t}_{12}\bar{k}_2^{12}$ ,  $L_2 = \bar{t}_{1-3}\bar{k}_2^{1-3}$ ,  $L_3 = \bar{t}_{1-4}\,\bar{k}_1^{12}\bar{k}_1^{34}$ ; for  $K_{33}: L_4 = \bar{t}_1\bar{t}_2\bar{t}_3\,\bar{k}_3^{1-3}$ ,  $L_5 = \bar{u}_1\bar{S}^1$ ,  $L_6 = \bar{t}_{13}\bar{t}_2\bar{t}_4\,\bar{k}_3^{1-4}$ ,  $L_{71} = \bar{z}_{12}\,\bar{k}_2^{1-3}\bar{t}_3$ ,  $L_{72} = \bar{y}^1\bar{t}_{145}\,\bar{k}_1^{45}$ ,  $L_{73} = \bar{t}_{12}\bar{k}_2^{1-3}\bar{u}_3$ ,  $L_{74} = \bar{t}_{14}\bar{k}_1^{45}\bar{t}_{52}\bar{k}_2^{1-3}\bar{t}_3$ ,  $L_{81} = \bar{k}_1^{12}\bar{t}_{1-4}\bar{s}^3\bar{s}^4$ ,  $L_{82} = \bar{k}_1^{12}\bar{t}_{1-3}\bar{v}^3$ ,  $L_{83} = \bar{X}^{34}\bar{z}_{34}$ ,  $L_{84} = \bar{X}^{14}\bar{t}_{45}\bar{k}_1^{56}\bar{t}_{61}$ , a sexagon, for  $K_{54}: L_9 = \bar{t}_1 \cdots \bar{t}_5\,\bar{k}_4^{1-5}$ ,  $L_{10} = \bar{u}_1\bar{t}_2\bar{t}_3\bar{t}_4\,\bar{k}_3^{1-4}$ ,  $L_{11} = \bar{y}^1\bar{Y}_1 = \bar{y}^1\bar{t}_{12}\bar{y}^2$ ,  $L_{121} = \bar{y}^1\bar{t}_{1-3}\bar{s}^2\bar{s}^3$ ,  $L_{122} = \bar{t}_1\bar{k}_2^{1-3}\bar{u}_2\bar{u}_3$ ,  $L_{123} = \bar{Y}_2\bar{v}^2$ ,  $L_{131} = \bar{s}^1 \cdots \bar{s}^4\,\bar{t}_{1-4}$ ,  $L_{132} = \bar{s}^1\bar{s}^2\bar{t}_{1-3}\bar{v}^3$ ,  $L_{133} = \bar{v}^1\bar{t}_{12}\bar{v}^2 = \bar{v}^1\bar{x}_1$ . For  $P_4(x)$ ,  $K_{23}: L_{14} = \bar{t}_1\bar{t}_2\,\bar{k}_3^{12}$ ,  $L_{15} = \bar{t}_{12}\bar{k}_2^{1-3}\bar{t}_3$ ,  $L_{161} = \bar{S}^1\bar{t}_{1-3}\bar{k}_1^{23}$ ,  $L_{191} = \bar{k}_2^{1-3}\bar{t}_{1-4}\bar{s}^4$ ,  $L_{182} = \bar{t}_{12}\bar{k}_3^{1-4}\bar{t}_{34}$ ,  $L_{191} = \bar{k}_2^{1-3}\bar{t}_{1-4}\bar{s}^4$ ,

$$\begin{split} L_{192} &= \bar{k}_1^{12} \bar{t}_{1-4} \bar{k}_2^{3-5} \bar{t}_5, \ L_{193} &= \bar{t}_{1-3} \bar{k}_2^{2-4} \bar{t}_{45} \bar{k}_1^{51}, \ L_{194} &= \bar{t}_{12} \bar{k}_1^{2-3} \bar{t}_{3-5} \bar{k}_1^{45}, \\ L_{201} &= \bar{k}_1^{12} \bar{k}_1^{34} \bar{t}_{1-5} \bar{s}^5, \ L_{202} &= \bar{k}_1^{12} \bar{t}_{1-4} \bar{k}^{34}, \ L_{203} &= \bar{k}_1^{12} \bar{t}_{1-3} \bar{k}_1^{34} \bar{t}_{4-6} \bar{k}_1^{56}, \\ L_{204} &= \bar{t}_{135} \left( \bar{k}_1^{12} \bar{k}_1^{34} \bar{k}_1^{56} \right) \bar{t}_{246}; \\ \text{for } K_{44} : L_{21} &= \bar{t} \cdots \bar{t}_4 \ \bar{k}_4^{1-4}, \ L_{221} &= \bar{Y}_2 \bar{k}_2^{23} \bar{t}_3, \ L_{222} &= \bar{u}_1 \bar{t}_2 \bar{t}_3 \ \bar{k}_3^{1-3}, \\ L_{231} &= \bar{s}^1 \bar{z}_{12} \bar{s}^2. \ L_{241} &= \bar{x}_2 \bar{s}^2, \ L_{242} &= \bar{u}_1 \bar{k}_2^{12} \bar{u}_2. \\ L_{25} &= \bar{t}_{12} \bar{k}_4^{1-5} \bar{t}_3 \bar{t}_4 \bar{t}_5, \ L_{261} &= \bar{t}_1 \bar{t}_2 \bar{k}_3^{1-4} \bar{t}_{3-5} \bar{s}^5, \ L_{262} &= \bar{t}_1 \bar{t}_2 \bar{t}_3 \bar{k}_3^{1-4} \bar{t}_{4-6} \bar{k}_1^{56}, \\ L_{263} &= \bar{t}_{12} \bar{k}_3^{1-4} \bar{t}_3 \bar{u}_4, \ L_{264} &= \bar{t}_1 \bar{t}_2 \bar{k}_3^{1-4} \left( \bar{t}_{35} \bar{t}_{46} \right) \bar{k}_1^{56}, \\ L_{271} &= \bar{t}_1 \bar{k}_2^{1-3} \bar{t}_{2-4} \bar{y}^4, \ L_{272} &= \bar{t}_{12} \bar{k}_2^{1-3} \bar{y}_3, \ L_{273} &= \bar{t}_1 \bar{k}_2^{1-3} \left( \bar{t}_{24} \bar{t}_{35} \right) \bar{k}_3^{4-6} \bar{t}_6, \\ L_{281} &= \bar{t}_1 \bar{k}_2^{1-3} \bar{t}_{2-4} \bar{y}^4, \ L_{285} &= \bar{y}^1 \bar{k}_1^{23} \bar{t}_{1-4} \bar{y}^4, \ L_{286} &= \bar{t}_{12} \bar{k}_2^{1-3} \bar{z}_{23}, \\ L_{287} &= \bar{t}_1 \bar{k}_2^{1-3} \bar{t}_{24} \bar{k}_1^{45} \bar{z}_{53}, \ L_{288} &= \bar{y}^1 \bar{t}_{1-3} \bar{x}^{23}, \\ L_{289} &= \bar{t}_1 \bar{k}_2^{1-3} \bar{t}_{33}, \ L_{2810} &= \bar{u}_1 \bar{k}_2^{1-3} \left( \bar{t}_{24} \bar{t}_{35} \right) \bar{k}_1^{45}, \\ L_{293} &= \bar{x}^{34} \bar{t}_{3-6} \bar{s}^5 \bar{s}^6, \ L_{294} &= \bar{k}_1^{12} \bar{t}_{1-3} \bar{k}_1^{34} \bar{t}_{4-6} \bar{s}^5 \bar{s}^6, \ L_{295} &= \bar{k}_1^{12} \bar{t}_{1-4} \bar{v}^3 \bar{s}^4, \\ L_{296} &= \bar{k}_1^{12} \bar{t}_{1-3} \bar{k}_1^{34} \bar{x}_4, \ L_{297} &= \bar{k}_1^{12} \bar{t}_{135} \bar{v}^5 \bar{t}_{24} \bar{k}_1^{35}, \\ L_{298} &= \bar{x}^{14} \bar{t}_{45} \bar{k}_1^{56} \bar{s}_{1}, \ L_{299} &= \bar{X}^{14} \bar{t}_{45} \bar{x}^{58}; \\ \text{for } K_{65} : \ L_{30} &= \bar{t}_1 \cdots \bar{t}_6 \ k_5^{1-6}, \ L_{31} &= \bar{u}_1 \bar{t}_2 \bar{t}_3 \bar{$$

These  $c_{rs}$  and  $L_i$  don't use derivatives of  $\bar{k}_e^{1-r}$ , the cumulant coefficients of  $\hat{w}$ .

**Corollary 1.** Suppose that  $\hat{w}$  is an unbiased standard estimate of  $w \in R^p$  with respect to n, and that q = 1. Then the cumulants of  $\hat{t} = t(\hat{w})$  can be expanded as (1) with bounded

cumulant coefficients  $K_{re}$ . The leading coefficients needed for  $P_r(x)$  of (4) for the distribution of  $Y_{nt}$  of (5) are as follows.

$$\begin{split} &K_{10} = \bar{t} = t(w). \ For \ P_0(x): \ K_{21} = v = \bar{t}_1 \ \bar{k}_1^{12} \ \bar{t}_2. \\ &For \ P_1(x): \ K_{11} = c_{02}/2, \ K_{32} = c_{21} + 3c_{23}. \\ &For \ P_2(x): \ K_{22} = \sum_{k=1}^2 {}_k K_{22}, \ {}_1 K_{22} = c_{11}, \ {}_2 K_{22} = c_{15} + c_{19}/2 + c_{1,10}, \\ &K_{43} = c_{31} + 12c_{36} + 12c_{3,10} + 4c_{3,11}. \\ &For \ P_3(x): \ K_{12} = \sum_{k=1}^2 {}_k K_{12}, \ {}_1 K_{12} = L_1/2, \ {}_2 K_{12} = L_2/6 + L_3/8; \\ &K_{33} = \sum_{k=2}^3 {}_k K_{33}, \ {}_2 K_{33} = L_4 + 6L_5, \ {}_3 K_{33} = 3L_6/2 + 3L_7 + L_8 \ where \end{aligned} \tag{3} \\ &L_7 = L_{71} + 3L_{72}/2 + 3 \sum_{k=3}^4 {}_L C_{7k}, \ L_8 = 3L_{81}/2 + 6L_{82} + 3L_{83} + 3L_{84}, \\ &K_{54} = L_9 + 20L_{10} + 15L_{11} + 30L_{12} + L_{13} \ where \\ &L_{12} = L_{121} + 2L_{122} + 2L_{123}, \ L_{13} = L_{131} + 60(L_{132} + L_{133}). \\ &For \ P_4(x): \ K_{23} = \sum_{k=1}^3 {}_k K_{23}, \ 1K_{23} = L_{14}, \ 2K_{23} = L_{15} + \sum_{k=1}^2 {}_L L_{16k} + L_{171}, \\ &3K_{23} = L_{181}/3 + L_{182}/4 + \sum_{k=19}^2 {}_L k \ where \ L_{19} = L_{191}/3 + \sum_{k=2}^4 {}_L L_{19k}, \\ &L_{20} = L_{201}/4 + L_{202}/2 + L_{203}/4 + L_{204}/6; \\ &K_{44} = \sum_{k=3}^4 {}_k K_{44}, \ {}_3 K_{44} = L_{21} + 12 \sum_{k=1}^2 {}_L L_{22k} + 12L_{231} + 24L_{241} + 12L_{242}. \\ &4K_{44} = 2L_{25} + 2L_{26} + 6 \sum_{k=1}^3 {}_L L_{77k} + L_{28} + L_{29} \ where \\ &L_{26} = 3L_{261} + L_{262} + 6L_{263} + 3L_{264}, \\ &L_{28} = \sum_{k=1}^{11} {}_{c_k} L_{28k}, \ c_1 = 6, c_3 = 24, c_8 = 4, c_k = 12 \ otherwise, \\ &L_{29} = \sum_{k=1}^9 {}_h k_{L_{29k}}, \ h_1 = 2, h_2 = h_6 = 12, h_7 = 24, h_9 = 3, h_k = 6 \ otherwise; \\ &K_{65} = L_{30} + 30L_{31} + 60vL_{32} + 60U_{33} + 90L_{34} + 60U_{35} + 6L_{36} \ for \end{aligned}$$

 $L_{33} = L_{331} + 3L_{332} + 2L_{333}, L_{34} = L_{341} + 4L_{342} + L_{343},$ 

$$L_{35} = \sum_{k=1}^{7} d_k L_{35k}, d_1 = 1, d_2 = d_3 = d_7 = 6, d_4 = 3, d_5 = 2, d_6 = 12,$$

$$L_{36} = \sum_{k=1}^{6} e_k L_{36k}, e_1 = 1, e_2 = 20, e_3 = 15, e_7 = 6, e_4 = e_5 = e_6 = 60.$$

$$Also, K_{13} = \bar{t}_{12} \bar{k}_3^{12} / 2 + \bar{t}_{1-3} \bar{k}_3^{1-3} / 6 + \bar{t}_{1-4} (\bar{k}^{12} \bar{k}^{34})_3 / 8 + \bar{t}_{1-4} \bar{k}_3^{1-4} / 24$$

$$+ \bar{t}_{1-5} \bar{k}_2^{1-3} \bar{k}_1^{45} / 12 + \bar{t}_{1-6} \bar{k}_1^{12} \bar{k}_1^{34} \bar{k}_1^{56} / 48,$$

$$K_{14} = \bar{t}_{12} \bar{k}_4^{12} / 2 + \bar{t}_{1-3} \bar{k}_4^{1-3} / 6 + \bar{t}_{1-4} [(\bar{k}^{12} \bar{k}^{34})_4 / 8 + \bar{k}_4^{1-4} / 24] + \bar{t}_{1-5} [\bar{k}_4^{1-5} / 120 + (\bar{k}^{1-3} \bar{k}^{45})_4 / 12] + \bar{t}_{1-6} [(\bar{k}^{12} \bar{k}^{34} \bar{k}^{56})_4 / 48 + \bar{k}_3^{1-4} \bar{k}_5^{56} / 48 + \bar{k}_2^{1-3} \bar{k}_2^{4-6} / 72]$$

$$+ \bar{t}_{1-7} \bar{k}_2^{1-3} \bar{k}_1^{45} \bar{k}_1^{67} / 48 + \bar{t}_{1-8} \bar{k}_1^{12} \bar{k}_1^{34} \bar{k}_5^{56} \bar{k}_1^{78} / 384.$$

PROOF Since q=1,  $\sum^N$  becomes N. We write  $T_{1-s}^{1-r}$ ,  $U_{1-s}^{1-r}$ ,  $V_{1-s}^{1-r}$  as  $T_{1-s}^r$ ,  $U_{1-s}^r$ ,  $V_{1-s}^r$ . By Theorem 3.1 we need the following.

$$\begin{split} T_{1-4}^3/3 &= \bar{t}_{13}\bar{t}_2\bar{t}_4, \ T_{1-3}^2/2 = \bar{t}_{12}\bar{t}_3, \ T_{1-4}^2/2 = \bar{t}_{1-3}\bar{t}_4 + \bar{t}_{13}\bar{t}_{24}, \\ T_{1-5}^4/12 &= \bar{t}_{14}\bar{t}_2\bar{t}_3\bar{t}_5, \ T_{1-6}^4/4 = \bar{t}_{135}\bar{t}_2\bar{t}_4\bar{t}_6 + 3\bar{t}_{13}\bar{t}_{25}\bar{t}_4\bar{t}_6, \\ T_{1-5}^3/3 &= \bar{t}_{124}\bar{t}_3\bar{t}_5 + 3\bar{t}_{145}\bar{t}_2\bar{t}_3/2 + 3\bar{t}_{12}\bar{t}_3\bar{t}_5 + 3\bar{t}_{14}\bar{t}_25\bar{t}_3, \\ T_{1-6}^3/3 &= 3\bar{t}_{1235}\bar{t}_4\bar{t}_6/2 + 6\bar{t}_{1-3}\bar{t}_{45}\bar{t}_6 + 3\bar{t}_{135}\bar{t}_{24}\bar{t}_6 + \bar{t}_{13}\bar{t}_{25}\bar{t}_{46}, \\ T_{1-6}^5/20 &= \bar{t}_{15}\bar{t}_2\bar{t}_3\bar{t}_4\bar{t}_6, \ U_{1-6}^5/15 = \bar{t}_{14}\bar{t}_2\bar{t}_3\bar{t}_5\bar{t}_6, \\ T_{1-7}^5/30 &= \bar{t}_{146}\bar{t}_2\bar{t}_3\bar{t}_5\bar{t}_7 + 2\bar{t}_{14}\bar{t}_26\bar{t}_3\bar{t}_5\bar{t}_7 + 2\bar{t}_{14}\bar{t}_56\bar{t}_2\bar{t}_3\bar{t}_7, \\ T_{1-8}^5 &= \bar{t}_{1357}\bar{t}_2\bar{t}_4\bar{t}_6\bar{t}_8 + 60\bar{t}_{135}\bar{t}_27\bar{t}_4\bar{t}_6\bar{t}_8 + 60\bar{t}_{13}\bar{t}_25\bar{t}_47\bar{t}_6\bar{t}_8. \ U_{1-4}^2 = \bar{t}_{1-3}\bar{t}_4/3 + \bar{t}_{12}\bar{t}_{34}/4, \\ T_{1-5}^2 &= \bar{t}_{1-4}\bar{t}_5/3 + \bar{t}_{1245}\bar{t}_3/2 + \bar{t}_{124}\bar{t}_{35} + \bar{t}_{145}\bar{t}_{23}/2, \\ T_{1-6}^2 &= \bar{t}_{1-5}\bar{t}_6 + 2\bar{t}_{1235}\bar{t}_{46} + \bar{t}_{1-3}\bar{t}_{4-6} + 2\bar{t}_{135}\bar{t}_{246}/3, \\ T_{1-5}^4/12 &= \bar{t}_{14}\bar{t}_2\bar{t}_3\bar{t}_5, \ U_{1-5}^4/4 = \bar{t}_{12}\bar{t}_3\bar{t}_4\bar{t}_5, \\ U_{1-6}^4/2 &= 3\bar{t}_{125}\bar{t}_3\bar{t}_4\bar{t}_6 + 2\bar{t}_{156}\bar{t}_2\bar{t}_3\bar{t}_4 + 6\bar{t}_{12}\bar{t}_{35}\bar{t}_4\bar{t}_6 + 3\bar{t}_{15}\bar{t}_{26}\bar{t}_3\bar{t}_4, \\ V_{1-6}^4/6 &= \bar{t}_{124}\bar{t}_3\bar{t}_5\bar{t}_6 + \bar{t}_{12}\bar{t}_3\bar{t}_5\bar{t}_6 + \bar{t}_{12}\bar{t}_{34}\bar{t}_5\bar{t}_6 + \bar{t}_{14}\bar{t}_{25}\bar{t}_3\bar{t}_6, \\ T_{1-7}^4 &= 6\bar{t}_{1246}\bar{t}_3\bar{t}_5\bar{t}_7 + 3\bar{t}_{1456}\bar{t}_2\bar{t}_3\bar{t}_7 + 24\bar{t}_{124}\bar{t}_36\bar{t}_5\bar{t}_7 + 12\bar{t}_{14}\bar{t}_{26}\bar{t}_3\bar{t}_7 \\ + 12\bar{t}_{145}\bar{t}_2\bar{t}_3\bar{t}_7 + 12\bar{t}_{146}\bar{t}_23\bar{t}_5\bar{t}_7 + 24\bar{t}_{146}\bar{t}_25\bar{t}_3\bar{t}_7 + 4\bar{t}_{146}\bar{t}_{25}\bar{t}_3\bar{t}_7 \\ + 12\bar{t}_{145}\bar{t}_2\bar{t}_3\bar{t}_4\bar{t}_6\bar{t}_8 + 12\bar{t}_{1235}\bar{t}_47\bar{t}_6\bar{t}_8 + 6\bar{t}_{1357}\bar{t}_24\bar{t}_6\bar{t}_8 \\ + 6\bar{t}_{1335}\bar{t}_{27}\bar{t}_4\bar{t}_6\bar{t}_8 + 6\bar{t}_{135}\bar{t}_{247}\bar{t}_6\bar{t}_8 + 12\bar{t}_{123}\bar{t}_4\bar{t}_6\bar{t}_7\bar{t}_7$$

$$\begin{split} T_{1-8}^6/60 &= \bar{t}_1 \bar{s}_1 \bar{t}_2 \bar{t}_3 \bar{t}_4 \bar{t}_6 \bar{t}_8 + 3 \bar{t}_1 \bar{s}_1 \bar{t}_2 \bar{t}_3 \bar{t}_4 \bar{t}_6 \bar{t}_8 + 2 \bar{t}_1 \bar{s}_1 \bar{t}_6 \bar{t}_8 \bar{t}_3 \bar{t}_3 \bar{t}_4 \bar{t}_8, \\ U_{1-8}^6/90 &= \bar{t}_1 \bar{t}_2 \bar{t}_3 \bar{t}_6 \bar{t}_8 + 4 \bar{t}_1 \bar{t}_2 \bar{t}_3 \bar{t}_6 \bar{t}_8 + \bar{t}_1 \bar{t}_8 \bar{t}_2 \bar{t}_3 \bar{t}_5 \bar{t}_9, \\ &+ 3 \bar{t}_4 6 \bar{s}_1 \bar{t}_3 \bar{t}_2 \bar{t}_3 \bar{t}_5 \bar{t}_7 \bar{t}_9 + 6 \bar{t}_1 4 \bar{t}_2 \bar{t}_3 \bar{t}_3 \bar{t}_7 \bar{t}_9 + 6 \bar{t}_1 4 \bar{t}_3 \bar{t}_6 \bar{t}_3 \bar{t}_7 \bar{t}_9 + 6 \bar{t}_1 4 \bar{t}_3 \bar{t}_6 \bar{t}_3 \bar{t}_7 \bar{t}_9 + 6 \bar{t}_1 4 \bar{t}_3 \bar{t}_6 \bar{t}_7 \bar{t}_9 \bar{t}_9$$

**Example 1.** Suppose that  $E \hat{w} = w$  and t(w) is linear in w. Then  $K_{1e} = 0$  for  $e \ge 1$ . For  $r \le 4$ , the  $K_{ij}$  needed for  $P_r(x)$  of (4) for the distribution of  $Y_{nt}$  of (5) are as follows.

$$K_{10} = \bar{t} = t(w)$$
. For  $P_0(x)$ :  $K_{21} = v$ . For  $P_1(x)$ :  $K_{32} = c_{21}$ . For  $P_2(x)$ :  $K_{22} = {}_{1}K_{22} = c_{11}$ ,  $K_{43} = c_{31}$ . For  $P_3(x)$ :  $K_{33} = {}_{2}K_{33} = L_4$ ,  $K_{54} = L_9$ . For  $P_4(x)$ :  $K_{23} = {}_{1}K_{23} = L_{14}$ ,  $K_{44} = {}_{3}K_{44} = L_{21}$ ,  $K_{65} = L_{30}$ .

For,  $\bar{u}_1$ ,  $\bar{v}^1$ ,  $\bar{x}_1$ ,  $\bar{z}_{12}$  are 0, as are most  $c_{ij}$ ,  $L_k$  and  $_2K_{22}$ ,  $_3K_{33}$ ,  $_2K_{23}$ ,  $_3K_{23}$ ,  $_4K_{44}$ . So for  $r=0,\cdots,4$ , for  $P_r(x)$  we only need to calculate these 3  $c_{ij}$  and 5  $L_j$ .

Let  $G_{\gamma}$  be a gamma random variable with known mean  $\gamma$ . Its rth cumulant is  $(r-1)!\gamma$ .

**Example 2.** Linear combinations of scale parameters. Suppose that  $E \hat{w} = w$  and t(w) is linear, the components of  $\hat{w}$  are independent, and for  $1 \le i \le p$ ,  $\hat{w}^i/w^i$  has a distribution with known rth cumulant  $n^{1-r}\kappa_{ri}$ . Then  $K_{re} = 0$  for  $e \ne r-1$  and

$$K_{r,r-1} = \sum_{i=1}^p t_{\cdot i}^r \kappa_{ri}(w^i)^r.$$

For example if  $\hat{w}_i/w_i$  is a standard exponential variable then  $\kappa_{ri}=(r-1)!$ .

For  $s \leq r$  and any function  $f^{i_s \cdots i_r}$ , set  $\sum_{s-r} \bar{f}^{s-r} = \sum_{i_s, \cdots, i_r} f^{i_s \cdots i_r}$  summed over their range. In Example 5.3 their range is 1, 2; for example in  $L_{123}$ ,  $\sum_1 t_{.2i_1} \bar{v}^1 = \sum_{i_1=1}^k t_{.2i_1} \bar{v}^1$ . In Example 5.4 their range is 1,  $\cdots$ , k; for example  $\bar{u}_1 = \sum_2 \bar{t}_{12} \bar{s}^2 = \sum_{i_2=1}^k \bar{t}_{12} \bar{s}^2$ .

**Example 3.** Suppose that  $\hat{\mu} \sim \mathcal{N}_1(\mu, V/n)$  and  $\hat{V}/V = \chi_f^2/f = G_\gamma/\gamma$  are independent, where  $\gamma = f/2$  has magnitude n. Set  $\nu = \gamma/n$ . Then

 $\kappa_r(\hat{\mu}) = \mu \delta_{r1} + V n^{-1} \delta_{r2}$ ,  $\kappa_r(\hat{V}) = k_r n^{1-r}$  for  $k_r = (r-1)! \nu^{1-r} V^r$ , and cross-cumulants of  $\hat{w}$  are zero. Take p = 2,  $w_1 = \mu$ ,  $w_2 = V$ . Then by Corollary 5.1,  $K_{re}$  are given in terms of

$$s^1 = t_{.1}V$$
,  $s^2 = t_{.2}k_2$ ,  $u_1 = t_{.11}t_{.1}V + t_{.12}t_{.2}k_2$ ,  $u_2 = t_{.12}t_{.1}V + t_{.22}t_{.2}k_2$ ,  $v^1 = Vu_1$ ,  $v^2 = k_2u_2$ ,

as follows.

For 
$$P_0(x)$$
:  $K_{21} = v = t_{.1}^2 V + t_{.2}^2 k_2 / 2$ .  
For  $P_1(x)$ :  $c_{02} = t_{.11} + t_{.22} k_2$ ,  $c_{21} = t_{.2}^3 k_3$ ,  $c_{23} = \sum_{i=1}^2 t_{.ii} (s^i)^2$ .  
For  $P_2(x)$ :  $c_{11} = 0$ ,  $c_{15} = t_{.22} t_{.2} k_3$ ,  $c_{19} = (t_{.11} V)^1 + (t_{.22} k_2)^2 + 2t_{.12} V k_2$ ,  $c_{31} = t_{.2}^4 k_4$ ,  $c_{36} = t_{.2}^2 k_3 u_2$ ,  $c_{3,10} = u_1^2 V + u_2^2 k_2$ ,  $c_{3,11} = \sum_{i=1}^2 (s^i)^3 t_{.iii} + 3 \sum_{12}^2 (s^1)^2 s^2 t_{.112}$  where  $\sum_{12}^2 f_{12} = f_{12} + f_{21}$ .  
For  $P_3(x)$ :  $L_1 = L_4 = L_5 = 0$ ,  $L_2 = t_{.222} k_3$ ,  $L_3 = t_{.1111} V^2 + 2t_{.1122} V k_2 + t_{.2222} k_2^2$ ,  $L_6 = t_{.22} t_{.2}^2 k_4$ ,  $L_{71} = z_{22} k_3 t_2$ ,  $L_{72} = y^2 (t_{.211} V + t_{.222} k_3)$ ,  $L_{73} = t_{.22} k_3 u_2$ ,  $L_{74} = (V t_{.12}^2 + k_2 t_{.22}^2) k_3 t_{.2}$ ,  $L_{81} = (V t_{.10} t_{.3} t_4 + k_2 t_{.21234}) \bar{s}^3 \bar{s}^4$ ,

$$\begin{split} L_{82} &= \sum_{1} (Vt_{.11i_1} + k_2t_{.22i_1}) v^{i_1}, \ L_{83} = V^2t_{.11}z_{11} + 2Vk_2t_{.12}z_{12} + k_2^2t_{.22}z_{22}, \\ L_{84} &= \sum_{1-3} \bar{k}_1^{11} \bar{k}_1^{22} \bar{k}_1^{33} \bar{t}_{12} \bar{t}_{23} \bar{t}_{31}, \\ L_9 &= t_{.2}^5 k_5, \ L_{10} = u_2t_{.2}^3 k_4, \ L_{11} = (y^2)^2t_{.22}, \ L_{121} = y^2 \sum_{12} t_{.2i_1i_2} \bar{s}^1 \bar{s}^2, \\ L_{122} &= t_{.2}k_3u_2^2, \ L_{123} = y^2 \sum_{1} t_{.2i_1} \bar{v}^1, \ L_{13} = \sum_{1-3} \bar{s}^1 \bar{s}^2 \bar{t}_{1-3} \bar{k}_1^{33} \bar{u}_3. \end{split}$$

Similarly one can write down the Ls needed for  $P_4(x)$ .

**Example 4.** Suppose that we have the summary statistics from k samples of size  $n_i$  from normal populations with means and variances  $\mu_i$ ,  $V_i$ ,  $1 \le i \le k$ . Take p = 2k,  $w_i = \mu_i$ ,  $w_{i+k} = V_i$ ,  $1 \le i \le k$ . So we have p independent statistics,  $\hat{\mu}_i \sim \mathcal{N}(\mu_i, V_i/n_i)$  and  $\hat{V}_i \sim V_i \chi_{f_i}^2/f_i = V_i G_{\gamma_i}/\gamma_i$ ,  $1 \le i \le k$  where  $\gamma_i = f_i/2$  has magnitude n, the total sample size. Set

$$v_i = \gamma_i/n$$
,  $\lambda_i = n_i/n$ ,  $\tau_i = V_i/\lambda_i$ .

Then  $\kappa_r(\hat{\mu}_i) = \mu_i \delta_{r1} + V_i(\lambda_i n)^{-1} \delta_{r2}$ ,  $\kappa_r(\hat{V}_i) = k_{ri}(\lambda_i n)^{1-r}$  for  $k_{ri} = (r-1)! \nu_i^{1-r}$ , and cross-cumulants of  $\hat{w}$  are zero. Suppose that t(w) only depends on  $\mu_1, \dots, \mu_k$ , as in Example 3.3 of Withers (1982). (The notation there is slightly different.) Then

$$\bar{s}^1 = \bar{t}_1 \bar{\tau}_1$$
,  $\bar{u}_1 = \sum_2 \bar{t}_{12} \bar{s}^2$ ,  $\bar{v}^1 = \bar{\tau}_1 \bar{u}_1$ ,  $\bar{z}_{12} = \sum_3 \bar{t}_{1-3} \bar{s}^3$ ,

and by Corollary 5.1, the coefficients needed are as follows.

$$For \ P_{0}(x): \ K_{21} = v = \bar{t}_{1}\bar{s}^{1} = \sum_{1} \bar{t}_{1}^{2}\bar{\tau}_{1},$$

$$For \ P_{1}(x): \ c_{02} = \sum_{1} \bar{t}_{11}\bar{\tau}_{1}, \ c_{21} = 0, \ c_{23} = \bar{s}^{1}\bar{t}_{12}\bar{s}^{2} = \sum_{12} \bar{t}_{1}\bar{\tau}_{1}\bar{t}_{12}\bar{t}_{2}\bar{\tau}_{2},$$

$$For \ P_{2}(x), \ K_{22}: \ c_{11} = c_{15} = 0, \ c_{19} = \sum_{12} \bar{t}_{12}^{2}\bar{\tau}_{1}\bar{\tau}_{2}, c_{1,10} = \sum_{12} \bar{t}_{1}\bar{\tau}_{1}\bar{t}_{122}\bar{\tau}_{2};$$

$$For \ K_{43}: \ c_{31} = c_{36} = 0, \ c_{3,10} = \sum_{1} \bar{u}_{1}^{2}\bar{\tau}_{1}, \ c_{3,11} = \sum_{1-3} \bar{s}^{1}\bar{s}^{2}\bar{s}^{3}\bar{t}_{1-3}.$$

$$For \ P_{3}(x), \ K_{12} = L_{3}/6 = \sum_{12} \bar{t}_{1122}\bar{\tau}_{1}\bar{\tau}_{2}/6 \ as \ L_{1} = L_{2} = 0;$$

$$for \ K_{33}: \ L_{4} = L_{5} = L_{6} = L_{7} = 0, \ L_{81} = \sum_{1-3} \bar{\tau}_{1}\bar{t}_{1123}\bar{s}^{2}\bar{s}^{3}, \ L_{82} = \sum_{12} \bar{\tau}_{1}\bar{t}_{122}\bar{v}^{2},$$

$$L_{83} = \sum_{12} \bar{t}_{12}\bar{\tau}_{1}\bar{\tau}_{2}\bar{z}_{12} = \sum_{1-3} \bar{\tau}_{1}\bar{\tau}_{2}\bar{\tau}_{3}\bar{t}_{12}\bar{t}_{123}\bar{t}_{3}, \ L_{83} = \sum_{1-3} \bar{\tau}_{1}\bar{\tau}_{2}\bar{\tau}_{3}\bar{t}_{12}\bar{t}_{23}\bar{t}_{31};$$

$$for \ K_{54}: \ L_{k} = 0 \ for \ 10 \le k \le 14.$$

$$For \ P_{4}(x), \ K_{23}: \ L_{i,k} = 0 \ for \ i = 15, 16, 18, 19, \ L_{171} = \sum_{12} \bar{t}_{12}^{2}\bar{\tau}_{1}\bar{\tau}_{2},$$

$$L_{201} = \sum_{1-3} \bar{\tau}_{1}\bar{\tau}_{2}\bar{t}_{11223}\bar{s}^{5}, \ L_{202} = \sum_{1-3} \bar{\tau}_{1}\bar{\tau}_{2}\bar{\tau}_{3}\bar{t}_{1123}\bar{t}_{23}, \ L_{203} = \sum_{1-3} \bar{\tau}_{1}\bar{\tau}_{2}\bar{\tau}_{3}\bar{t}_{112}\bar{t}_{233},$$

$$L_{204} = \sum_{1-3} \bar{\tau}_{1}\bar{\tau}_{2}\bar{\tau}_{3}\bar{t}_{123};$$

$$\begin{split} &K_{44} = L_{29} \ as \ L_k = 0 \ for \ k = 21, (22,k), (23,1), (24,k), 25, (26,k), (27,k), (28,k), \\ &L_{291} = \sum_{1-4} \bar{\tau}_1 \bar{s}^2 \bar{s}^3 \bar{s}^4 \bar{t}_{11234}, \ L_{292} = \sum_{1-3} \bar{\tau}_1 \bar{t}_{1123} \bar{v}^2 \bar{s}^3, \ L_{293} = \sum_{1-4} \bar{\tau}_1 \bar{\tau}_2 \bar{s}^3 \bar{s}^4 \bar{t}_{1234}, \\ &L_{294} = \sum_{1-4} \bar{\tau}_1 \bar{\tau}_2 \bar{t}_{112} \bar{t}_{234} \bar{s}^3 \bar{s}^4, \ L_{295} = \sum_{12} \bar{\tau}_1 \bar{z}_{12}^2 \bar{\tau}_2, \ L_{296} = \sum_{1-3} \bar{\tau}_1 \bar{\tau}_2 \bar{\tau}_3 \bar{t}_{112} \bar{t}_{23}, \\ &L_{297} = \sum_{12} \bar{\tau}_1 \bar{x}_{12} \bar{t}_{12} \ for \ \bar{x}_{12} = \sum_{3} \bar{t}_{123} \bar{v}^3, \ L_{298} = \sum_{1-3} \bar{\tau}_1 \bar{\tau}_2 \bar{\tau}_3 \bar{t}_{12} \bar{t}_{23} \bar{z}_{31}, \\ &L_{299} = \sum_{1-4} \bar{\tau}_1 \bar{\tau}_2 \bar{\tau}_3 \bar{\tau}_4 \bar{t}_{12} \bar{t}_{23} \bar{t}_{34} \bar{t}_{41}; \\ &K_{65} = 6L_{36}, \ as \ the \ other \ components \ of \ K_{65} \ are \ 0. \ Also, \\ &K_{13} = \sum_{1-3} \bar{\tau}_1 \bar{\tau}_2 \bar{\tau}_3 \bar{t}_{112233} / 48, \ K_{14} = \sum_{1-4} \bar{\tau}_1 \bar{\tau}_2 \bar{\tau}_3 \bar{\tau}_4 \bar{t}_{11223344} / 384. \end{split}$$

**Corollary 2.** Set  $Y_n = v^{-1/2}Y_{nt} = (n/v)^{1/2}(t(\hat{w}) - t(w))$ . Then

$$Prob.(Y_n \le x) = \sum_{r=0}^{\infty} n^{-r/2} P_{rv}(x)$$

where  $P_{rv}(x)$  is  $P_r(x)$  of Corollary 5.1 with  $K_{re}$  replaced by  $K_{re}/v^{r/2}$ .

PROOF This is straightforward. □

Looking at  $K_{re}$ ,  $c_{ab}$ , v,  $\bar{s}^2$ ,  $\bar{u}_3$  as functions of w, we denote their partial derivatives with respect to  $\bar{w}_3 = w_{i_3}$ , say. by  $\bar{K}_{re,3}$ ,  $\bar{c}_{ab,3}$ ,  $\bar{v}_{,3}$ ,  $\bar{s}_{,3}^2$ ,  $\bar{u}_{3,3}$  and similarly for higher derivatives. We shall give the ones we need in Lemma 5.1. When constructing confidence regions, one needs to assume that the distribution of  $\hat{w}$  is determined by w. So far we've not assumed this. For  $\hat{w}$  biased, we need

**Corollary 3.** Let  $\hat{w} \in R^p$  be a biased standard estimate of w satisfying (1) where  $\bar{k}_e^{1-r}$  may depend on w. Then for q = 1,  $\hat{t} = t(\hat{w})$  is a standard estimate of  $t = t(w) \in R$ :

$$\kappa_r(\hat{t}) = \sum_{e=r-1}^{\infty} n^{-e} a_{re} \text{ for } r \ge 1, \text{ where } a_{re} = K_{re} + D_{re}, D_{r,r-1} = 0,$$
(6)

and the other  $D_{re}$  needed for  $P_i(x)$  of (4) for the distribution of  $Y_{nt}$  of (5) are as follows.

For 
$$P_0(x)$$
:  $D_{21} = 0 \Rightarrow a_{21} = K_{21} = v = \bar{t}_1 \bar{k}_1^{12} \bar{t}_2$ .  
For  $P_1(x)$ :  $D_{11} = c_{01} \Rightarrow a_{11} = c_{01} + c_{02}/2$ ,  $a_{32} = K_{32} = c_{21} + 3c_{23}$ ,

For 
$$P_2(x)$$
:  $D_{22} = \bar{v}_{.3} \, \bar{k}_1^3 = c_{12} + 2c_{16}$ ,  $a_{43} = K_{43}$ .  
For  $P_3(x)$ :  $D_{12} = K_{11,1} + K_{10,2}$ ,  $K_{11,1} = \bar{K}_{11.3} \, \bar{k}_1^3$ ,  $K_{10,2} = \bar{t}_1 \bar{k}_2^1 + \bar{k}_1^1 \bar{t}_{12} \bar{k}_1^2 / 2$ ,  $D_{33} = \bar{K}_{32.4} \, \bar{k}_1^4$ ,  $a_{54} = K_{54}$ .  
For  $P_4(x)$ :  $D_{23} = K_{22,1} + K_{21,2}$ ,  $K_{22,1} = \bar{K}_{22.3} \, \bar{k}_1^3$ ,  $K_{21,2} = \bar{v}_{.3} \, \bar{k}_2^3 / 2 + \bar{v}_{.34} \, \bar{k}_1^3 \bar{k}_1^4$ ,  $D_{44} = \bar{K}_{43.5} \, \bar{k}_1^5$ ,  $a_{65} = K_{65}$ .

PROOF This follows from Theorem 4.1.  $D_{rc} = \sum_{c=1}^{c-r+1} K_{r,c-r,e}$  where  $K_{rke}$  is the coefficient of  $n^{-e}$  in the expansion of  $K_{rk}(E \hat{w})$  about  $K_{rk}$ .  $\Box$  For  $r \leq s$ , and any  $\bar{X}_{r-s}$ , let  $\sum_{r-s}^{N} \bar{X}_{r-s}$  sums over all N permutations of  $i_r, \cdots, i_s$  giving distinct terms. For example

$$\sum_{3-5}^{3} \bar{t}_{13}\bar{t}_{245} = \bar{t}_{13}\bar{t}_{245} + \bar{t}_{14}\bar{t}_{235} + \bar{t}_{15}\bar{t}_{234}.$$

The derivatives of  $v = K_{21}$  and  $K_{re}$  needed for Corollary 5.3 are given by by

# Lemma 1.

$$\bar{v}_{.3} = 2\bar{u}_{3} + \bar{T}_{3}, \ \bar{v}_{.34} = \sum_{k=0}^{2} \bar{v}_{34k}, \ where \ \bar{T}_{3} = \bar{t}_{1}\bar{t}_{2}\bar{k}_{1.3}^{12},$$

$$\bar{v}_{340} = 2\bar{z}_{34} + 2\bar{t}_{31}\bar{k}_{1}^{12}\bar{t}_{24}, \ \bar{v}_{341} = 2\bar{t}_{1}(\bar{k}_{1.3}^{12}\bar{t}_{24} + \bar{k}_{1.4}^{12}\bar{t}_{23}), \ \bar{v}_{342} = \bar{t}_{1}\bar{t}_{2}\bar{k}_{1.34}^{12}.$$

$$\bar{v}_{.3-5} = \sum_{k=0}^{3} \bar{v}_{3-5k} for \ \bar{v}_{3-50} = 2\bar{k}_{1}^{62} \sum_{3-5}^{3} \bar{t}_{63}\bar{t}_{245} + \bar{s}^{6}\bar{t}_{3-6},$$

$$\bar{v}_{3-51} = 2\sum_{3-5}^{3} (\bar{t}_{1}\bar{k}_{1.3}^{12}\bar{t}_{245} + \bar{t}_{31}\bar{k}_{1.4}^{12}\bar{t}_{25}), \ \bar{v}_{3-52} = 2\bar{t}_{1} \sum_{3-5}^{3} \bar{t}_{23}\bar{k}_{1.45}^{12}, \ \bar{v}_{3-53} = \bar{t}_{1}\bar{t}_{2}\bar{k}_{1.3-5}^{12},$$

$$2\bar{k}_{11.3} = c_{02.3} = \bar{k}_{1}^{12}\bar{t}_{1-3} + \bar{t}_{12}\bar{k}_{1.3}^{12}, \ \bar{K}_{32.4} = c_{21.4} + 3c_{23.4}$$

$$for \ c_{21.4} = 3\bar{Y}_{4} + \bar{t}_{1}\bar{t}_{2}\bar{t}_{3}\bar{k}_{2.4}^{1-3}, \ c_{23.4} = \bar{s}^{1}\bar{s}^{2}\bar{t}_{124} + 2b_{4}, \ b_{4} = \bar{s}_{1.4}^{4}\bar{u}_{1} = \bar{x}_{4} + \bar{u}_{1}\bar{k}_{1.4}^{12}\bar{t}_{2}.$$

$$\bar{k}_{22.3} = c_{11.3} + c_{15.3} + c_{19.3}/2 + c_{1,10.3} for \ c_{11.3} = 2\bar{t}_{31}\bar{s}^{1} + \bar{t}_{1}\bar{t}_{2}\ \bar{k}_{2.3}^{12},$$

$$c_{15.4} = \sum_{k=1}^{3} c_{15k4}, \ c_{1514} = \bar{t}_{14}\bar{k}_{2}^{1-3}\bar{t}_{23}, \ c_{1524} = \bar{t}_{1}\bar{k}_{2}^{1-3}\bar{t}_{2-4}, \ c_{1534} = \bar{t}_{1}\bar{k}_{2.4}^{1-3}\bar{t}_{23},$$

$$c_{19.5} = 2c_{1915} + 2c_{1925}, \ c_{1915} = \bar{t}_{34}\bar{k}_{1}^{31}\bar{k}_{1}^{42}\bar{t}_{125}, \ c_{1925} = \bar{k}_{1.5}^{23}\bar{t}_{34}\bar{k}_{1}^{41}\bar{t}_{12},$$

$$c_{1,10.4} = \sum_{k=1}^{3} c_{1,10k4}, \ c_{1,1014} = \bar{k}_{1}^{23}\bar{t}_{1-3}\bar{k}_{1.4}^{23},$$

$$c_{1,1024} = \bar{s}^{1}\bar{k}_{1}^{23}\bar{t}_{1-4}, \ c_{1,1034} = \bar{s}^{1}\bar{t}_{1-3}\bar{k}_{1.4}^{23},$$

$$\bar{K}_{43.5} = c_{31.5} + 12c_{36.5} + 12c_{3,10.5} + 4c_{3,11.5} for c_{31.5} = 4\bar{t}_{51} \bar{k}_{3}^{1-4} \bar{t}_{2} \bar{t}_{3} \bar{t}_{4} 
+ \bar{t}_{1} \cdots \bar{t}_{4} \bar{k}_{3.5}^{1-4}, c_{36.5} = 2\bar{t}_{51} \bar{k}_{2}^{1-3} \bar{t}_{2} \bar{u}_{3} + \bar{t}_{1} \bar{t}_{2} \bar{u}_{3} \bar{k}_{2.5}^{1-3} + \bar{y}^{3} \bar{t}_{3-5} \bar{s}^{4} + \bar{Y}_{4} (\bar{k}_{1.5}^{41} \bar{t}_{1} + \bar{k}_{1}^{41} \bar{t}_{15}). 
c_{3,10.3} = \bar{u}_{1} \bar{k}_{1.3}^{12} \bar{u}_{2} + 2\bar{v}^{1} \bar{s}^{2} \bar{t}_{1-3} + 2\bar{t}_{1} \bar{x}_{2} \bar{k}_{1.3}^{12} + 2\bar{x}_{1} \bar{k}_{1}^{12} \bar{t}_{23}, 
c_{3,11.5} = 3\bar{s}^{1} \bar{s}^{2} \bar{t}_{1-3} \bar{s}_{.5}^{3} = 3\bar{s}^{1} \bar{s}^{2} \bar{t}_{1-3} \bar{t}_{4} \bar{k}_{1.5}^{34} + 3\bar{s}^{1} \bar{s}^{2} \bar{t}_{1-3} \bar{k}_{1}^{34} \bar{t}_{45}.$$
(9)

PROOF For example substitute  $\bar{u}_{3.5} = \bar{t}_{3-5}\bar{s}^4 + \bar{t}_{34}\bar{k}_{1.5}^{41}\bar{t}_1 + \bar{t}_{34}\bar{k}_1^{41}\bar{t}_{15}$  into  $c_{36.5} = 2\bar{t}_1\bar{t}_{25}\bar{k}_2^{1-3}\bar{u}_3 + \bar{t}_1\bar{t}_2(\bar{k}_{2.5}^{1-3}\bar{u}_3 + \bar{k}_2^{1-3}\bar{u}_{3.5}).$   $\Box$  So now we can write  $D_{re}$  needed for Corollary 5.3 in molecular form:

**Corollary 4.** Assume that the conditions of Corollary 5.3 hold. Then  $D_{re}$  and  $K_{re,j}$  given there satisfy

$$\begin{split} D_{22} &= c_{12} + 2c_{16} \Rightarrow a_{22} = c_{11} + c_{15} + c_{19}/2 + c_{1,10} + c_{12} + 2c_{16}. \\ For \, D_{12}, \, K_{11,1} &= (\bar{t}_{1-3} \, \bar{k}_{1}^{12} + \bar{t}_{12} \, \bar{k}_{1,3}^{12}) \bar{k}_{1}^{3}/2. \\ D_{33} &= (3\bar{y}^{3}\bar{t}_{34} + \bar{t}_{1}\bar{t}_{2}\bar{t}_{3}\bar{k}_{2,4}^{1-3} + 3\bar{s}^{1}\bar{s}^{2}\bar{t}_{124} + 6\bar{v}^{1}\bar{t}_{14} + 6\bar{u}_{1}\bar{k}_{1,4}^{12}\bar{t}_{2}) \bar{k}_{1}^{4}. \\ For \, D_{23}, \, K_{22,1} &= (2\bar{t}_{31}\bar{S}^{1} + \bar{t}_{1}\bar{t}_{2} \, \bar{k}_{2,3}^{12}) \bar{k}_{1}^{3} + (\bar{t}_{14}\bar{k}_{2}^{1-3}\bar{t}_{23} + \bar{t}_{1}\bar{k}_{2}^{1-3}\bar{t}_{2-4} + \bar{t}_{1}\bar{k}_{2,4}^{1-3}\bar{t}_{23}) \bar{k}_{1}^{4} \\ &+ (\bar{t}_{125}\bar{X}^{12} + \bar{k}_{1.5}^{23}\bar{t}_{34}\bar{k}_{1}^{41}\bar{t}_{12}) \bar{k}_{1}^{5} + [\bar{k}_{1}^{23}\bar{t}_{1-3}(\bar{k}_{1}^{15}\bar{t}_{54} + \bar{t}_{5}\bar{k}_{1.4}^{15}) + \bar{s}^{1}\bar{k}_{1}^{23}\bar{t}_{1-4} + \bar{s}^{1}\bar{t}_{1-3}\bar{k}_{1.4}^{23}] \bar{k}_{1}^{4}; \\ K_{21,2} &= (2\bar{u}_{3} + \bar{t}_{1}\bar{t}_{2}\bar{k}_{1.3}^{12}) \, \bar{k}_{2}^{3}/2 + [4\bar{t}_{1}\bar{k}_{1.3}^{12}\bar{t}_{24} + \bar{s}^{1}\bar{t}_{134} + \bar{t}_{31}\bar{k}_{1}^{12}\bar{t}_{24} \\ &+ \bar{t}_{1}\bar{t}_{2}\bar{k}_{1.34}^{12}] \, \bar{k}_{1}^{3}\bar{k}_{1}^{4}. \\ D_{44} &= [4\bar{t}_{51} \, \bar{k}_{3}^{1-4}\bar{t}_{2}\bar{t}_{3}\bar{t}_{4} + \bar{t}_{1} \cdots \bar{t}_{4} \, \bar{k}_{3.5}^{1-4} + 24\bar{t}_{51}\bar{k}_{2}^{1-3}\bar{t}_{2}\bar{u}_{3} + 12\bar{t}_{1}\bar{t}_{2}\bar{u}_{3}\bar{k}_{2.5}^{1-3} \\ &+ 12\bar{y}^{3}(\bar{t}_{3-5}\bar{s}^{4} + \bar{t}_{34}\bar{k}_{1.5}^{41}\bar{t}_{1} + \bar{t}_{34}\bar{k}_{1}^{41}\bar{t}_{15})] \, \bar{k}_{1}^{5} + 12[2\bar{v}^{2}\bar{t}_{2-4}\bar{s}^{4} + 2\bar{t}_{1}\bar{x}_{2}\bar{k}_{1.3}^{12} \\ &+ 2\bar{x}_{1}\bar{k}_{1}^{12}\bar{t}_{23} + 12\bar{u}_{1}\bar{k}_{12}^{12}\bar{u}_{2}] \, \bar{k}_{1}^{3} + 12(\bar{s}^{1}\bar{s}^{2}\bar{t}_{1-3}\bar{t}_{4}\bar{k}_{1}^{3}\bar{t}_{3} + \bar{s}^{1}\bar{s}^{2}\bar{t}_{1-3}\bar{k}_{1}^{3}\bar{k}_{1}^{3}\bar{t}_{4} + \bar{t}_{1}) \, \bar{k}_{1}^{5}. \end{split}$$

PROOF  $a_{1e}$  were given for  $i \le 4$  by Theorem 4.1. Corollaries 5.3, 5.4 agree with  $a_{11}$ ,  $a_{32}$ ,  $a_{22}$ ,  $a_{43}$  given for  $P_r(x)$ ,  $r \le 2$  on p59 of Withers (1982) except that  $D_{22}$  in  $a_{22}$  was overlooked.  $\Box$ 

#### 6. An Extension to Theorem 2.1

Here we remove the condition in Theorem 2.1 that  $E \hat{w} = w$  and give the extra terms referred to but not given on p49 of James and Mayne (1962). We use  $_{\ell}\bar{K}^{1-r}$  of Theorem 2.1, and the shorthand  $\bar{f}_{.m} = \partial_{i_m}\bar{f}$  where  $\partial_i = \partial/\partial w_i$ . Suppose that for  $r \geq 1$ , the rth order cumulants of (1) can be expanded as

$$\bar{k}^{1-r} = \kappa(\hat{w}^{i_1}, \dots, \hat{w}^{i_r}) = \sum_{e=r-1}^{\infty} {}_{e}\bar{k}^{1-r} \text{ for } 1 \le i_1, \dots, i_r \le p,$$
where  ${}_{e}\bar{k}^{1-r} = O(n^{-e})$ , and that  $\hat{w} \to w$  as  $n \to \infty$ , so that  ${}_{0}\bar{k}^1 = \bar{w}^1$ .

There is a key difference with Theorem 2.1: there,  $e^{\bar{k}^{1-r}}$  was treated as an algebraic expression. But now we must view each of them as a function of w. So we assume that the distribution of  $\hat{w}$  is determined by w.

The derivatives of  ${}_{\ell}\bar{K}^{1-r}$  of Theorem 2.1 are given by Leibniz's rule for the derivatives of a product:

$${}_{1}\bar{K}_{,3}^{12} = (\bar{t}_{1}^{1}\bar{t}_{2}^{2}\,\bar{k}^{12})_{,3} = (\bar{t}_{13}^{1}\bar{t}_{2}^{2} + \bar{t}_{1}^{1}\bar{t}_{23}^{2})\,\bar{k}^{12} + \bar{t}_{1}^{1}\bar{t}_{2}^{2}\,\bar{k}_{,3}^{12}, \\ {}_{1}\bar{K}_{,3}^{1} = (\bar{t}_{12}^{1}\bar{k}^{12})_{,3}/2 = \bar{t}_{1-3}^{1}\bar{k}^{12}/2 + \bar{t}_{12}^{1}\bar{k}_{,3}^{12}/2, \\ (\bar{t}_{1}^{1}\bar{t}_{2}^{2}\bar{t}_{3}^{3}\,\bar{k}^{1-3})_{,4} = (\bar{t}_{1}^{1}\bar{t}_{2}^{2}\bar{t}_{3}^{3})_{,4}\bar{k}^{1-3} + \bar{t}_{1}^{1}\bar{t}_{2}^{2}\bar{t}_{3}^{3}\,\bar{k}_{,4}^{1-3}, (\bar{t}_{1}^{1}\bar{t}_{2}^{2}\bar{t}_{3}^{3})_{,4} = \bar{t}_{1}^{1}\bar{t}_{2}^{2}\bar{t}_{34}^{3} + \bar{t}_{1}^{1}\bar{t}_{24}^{2}\bar{t}_{3}^{3} + \bar{t}_{14}^{1}\bar{t}_{2}^{2}\bar{t}_{3}^{3}, \\ (T_{1-4}^{1-3}\,\bar{k}^{12}\bar{k}^{34})_{,5} = (T_{1-4}^{1-3})_{,5}\,\bar{k}^{12}\bar{k}_{3}^{34} + T_{1-4}^{1-3}\,(\bar{k}^{12}\bar{k}^{34})_{,5}, \\ (T_{1-4}^{1-3})_{,5} = \sum_{i=1}^{3}(\bar{t}_{135}^{1}\bar{t}_{2}^{2}\bar{t}_{4}^{3} + \bar{t}_{13}^{1}\bar{t}_{25}^{2}\bar{t}_{4}^{3} + \bar{t}_{13}^{1}\bar{t}_{25}^{2}\bar{t}_{4}^{3} + \bar{t}_{13}^{1}\bar{t}_{25}^{2}\bar{t}_{3}^{3}), (\bar{k}^{12}\bar{k}^{34})_{,5} = \bar{k}_{.5}^{12}\bar{k}^{34} + \bar{k}^{12}\bar{k}_{.5}^{34}.$$

**Theorem 4.** Let  $\hat{w} \in R^p$  be a biased standard estimate of w satisfying (1). Then  $\hat{t} = t(\hat{w}) \in R^p$  is a standard estimate of t = t(w):

$$\kappa(\hat{t}^{j_1}, \dots, \hat{t}^{j_r}) = \sum_{e=r-1}^{\infty} {}_{e}\bar{a}^{1-r} \text{ for } r \ge 1, \ 1 \le j_1, \ \dots, \ j_r \le q, \tag{2}$$

where 
$$_{e}\bar{a}^{1-r} = _{e}\bar{K}^{1-r} + _{e}\bar{D}^{1-r}, _{r-1}\bar{D}^{1-r} = 0,$$
 (3)

and the other  $_e\bar{D}^{1-r}=_eD^{j_1...j_r}$  needed for  $P_{rn}(x)$  of (6) for the distribution of  $Y_{nt}$  of (5) are as follows.

$$\begin{split} &For \, P_0(x): \ _1\bar{D}^{12} = 0 \Rightarrow \ _1\bar{a}^{12} = \ _1\bar{K}^{12} = \ _1K^{j_1j_2} = \bar{t}_1^1\bar{t}_2^2 \ _1\bar{k}^{12}], \\ &For \, P_1(x): \ _1\bar{D}^1 = \ _1\bar{t}^1\bar{k}^1 \Rightarrow \ _1\bar{a}^1 = \bar{t}_1^1 \ _1\bar{k}^1 + \bar{t}_{12}^1 \ _1\bar{k}^{12}/2, \\ &For \, P_2(x): \ _2\bar{D}^{12} = \ _1\bar{K}_{.3}^{12} \ _1\bar{k}^3 \Rightarrow \bar{a}_2^{12} = \bar{t}_1^1\bar{t}_2^2 \ _2\bar{k}^{12} + T_{1-3}^{12} \ _2\bar{k}^{1-3}/2 \\ &+ T_{1-4}^{12} \ _1\bar{k}^{12} \ _1\bar{k}^{34}/2 + (\bar{t}_{13}^1\bar{t}_2^2 + \bar{t}_1^1\bar{t}_{23}^2) \ _1\bar{k}^{12} + \bar{t}_1^1\bar{t}_2^2 \ _1\bar{k}_{.3}^{12} \ _1\bar{k}^3. \\ &For \, P_3(x): \ _2\bar{D}^1 = \ _1\bar{K}_1^1 + \ _0\bar{K}_2^1, \ _1\bar{K}_1^1 = (\ _1\bar{K}^1)_{.3} \ _1\bar{k}^3, \\ &0\bar{K}_2^1 = \bar{t}_1^1 \ _2\bar{k}^1 + \bar{t}_{12}^1 \ _1\bar{k}^1 \ _1\bar{k}^2/2 \Rightarrow \\ &2\bar{a}^1 = \bar{t}_1^1 \ _2\bar{k}^1 + \bar{t}_{12}^1 (\ _2\bar{k}^{12} + \ _1\bar{k}^1 \ _1\bar{k}^2 + \ _1\bar{K}_{.3}^{12} \ _1\bar{k}^3)/2 \\ &+ \bar{t}_{1-3}^1 (\ _2\bar{k}^{1-3}/6 + \ _1\bar{k}^1 \ _1\bar{k}^{23}/2) + \bar{t}_{1-4}^1 \ _1\bar{k}^{12} \ _1\bar{k}^{34}/8, \end{split}$$

$$\begin{split} &_3\bar{D}^{1-3} = \ _2\bar{K}_{.4}^{1-3} \ \bar{k}_1^4 = (\bar{t}_1^1\bar{t}_2^2\bar{t}_3^3 \ _2\bar{k}^{1-3})_{.4} \ _1\bar{k}^4 + (T_{1-4}^{1-3} \ _1\bar{k}^{12} \ _1\bar{k}^{34})_{.5} \ _1\bar{k}^5. \\ &For \ P_4(x): \ _3\bar{D}^{12} = \ _2\bar{K}_1^{12} + \ _1\bar{K}_2^{12}, \ _2\bar{K}_1^{12} = \ _2\bar{K}_{.3}^{12} \ _1\bar{k}^3, \\ &_1\bar{K}_2^{12} = \ _1\bar{K}_{.3}^{12} \ _2\bar{k}^3/2 + \ _1\bar{K}_{.34}^{12} \ _1\bar{k}^3 \ _1\bar{k}^4, \ _4\bar{D}^{1-4} = \ _3\bar{K}_{.5}^{1-4} \ _1\bar{k}^5, \ _5\bar{D}^{1-5} = 0. \end{split}$$

PROOF  $\bar{K}^{1-r}(w) = \bar{K}^{1-r}$  and  ${}_{e}\bar{K}^{1-r}(w) = {}_{e}\bar{K}^{1-r}$  are functions of w. By (3)

$$\bar{K}^{1-r}(w_n) = \sum_{e=r-1}^{\infty} {}_k \bar{K}^{1-r}(w_n) \text{ for } w_n = E \hat{w} = w + d_n,$$
(4)

where by (1),  $d_n$  has  $i_1$ th component  $\bar{d}_n^1 = d_n^{i_1} = \sum_{e=1}^{\infty} e^{\bar{k}^1}$ . Consider the Taylor series expansion

$$_{k}\bar{K}^{1-r}(w+d_{n}) = _{k}\bar{K}^{1-r} + _{k}\bar{K}^{1-r}_{.1}\bar{d}_{n}^{1} + _{k}\bar{K}^{1-r}_{.12}\bar{d}_{n}^{1}\bar{d}_{n}^{2}/2! + \cdots = \sum_{e=0}^{\infty} _{k}\bar{K}^{1-r}_{e}$$

say. Substituting into (3) gives (2) with

$$_{c}\bar{a}^{1-r} = \sum_{k+e=c} {}_{k}\bar{K}_{e}^{1-r} = \sum_{e=0}^{c-r+1} {}_{c-e}\bar{K}_{e}^{1-r}.$$

Also  $_k \bar{K}_0^{1-r} = _k \bar{K}^{1-r}$  so that (3) holds with

$$_{c}\bar{D}^{1-r} = \sum_{e=1}^{c-r+1} \ _{c-e}\bar{K}_{e}^{1-r}: \ _{r}\bar{D}^{1-r} = \ _{r-1}\bar{K}_{1}^{1-r}, \ _{r+1}\bar{D}^{1-r} = \sum_{e=1}^{2} \ _{r+1-e}\bar{K}_{e}^{1-r}, \ \cdots \qquad \Box$$

The Edgeworth expansion (6) holds if  $\{e^{\bar{K}^{1-r}}\}$  are replaced by  $\{e^{\bar{a}^{1-r}}\}$ .

# 7. Discussion

Approximations to the distributions of estimates is of vital importance in statistical inference. Asymptotic normality uses just the 1st term of the Edgeworth expansion. That approximation can be greatly improved with further terms. When the estimate is a sample mean, basic results were given by Chebyshev, Charlier and Edgeworth in the 19th century with major advances in the 20th century by Cramer, Rao and many others. See Stuart and Ord (1987) for some historical references. For a derivation of the Edgeworth expansion for a sample mean from the Gram-Charlier expansion, see Withers and Nadarajah (2009, 2014a) for the univariate and vector cases. These showed for the 1st time that the coefficients in these expansions were Bell polynomials in the cumulants.

The first extension from a sample mean for univariate estimates was by Cornish and Fisher (1937) and Fisher and Cornish (1960). They assumed that the rth cumulant of the estimate was  $\kappa_r(\hat{w}) = n^{1-r}k_r$  where  $k_r$  is a constant. However in applications they assumed that  $\hat{w}$  was a Type A estimate, and collected terms. It was not until Withers (1984) that explicit results were given a univariate Type A estimate. Major advances were made in Withers and Nadarajah (2010b). This gave explicit results for the terms in the Edgeworth expansion of a Type A or B estimate using Bell polynomials, as outlined in §1. It also allowed for expansions about *asymptotically* normal random variables. The advantage of this approach in greatly reducing the number of terms in each  $P_r(x)$  was illustrated in Withers and Nadarajah (2012d, 2014c).

For univariate estimates, Cornish and Fisher (1937) also showed how to invert the Edgeworth expansion to obtain an expansion for the distribution quantiles. This was extended to Type A estimates in Withers (1984). For extensions to transformations of multivariate estimates, like  $t(\hat{w}) = (\hat{w} - w))'V^{-1}(\hat{w} - w)$ , see Hill and Davis (1968) and Withers and Nadarajah (2012a, 2012e). An application to the amplitude and phase of the mean of a complex sample is given in Withers and Nadarajah (2013b).

Turning now to smooth functions of a Type A estimate, the 1st univariate results were given by Withers (1982, 1983). These built on a deep result of James and Mayne (1962). This is why if  $\hat{w}$  is a Type A (or B) estimate of w, then a smooth function of  $\hat{w}$ , say  $t(\hat{w})$ , is a Type A (or B) estimate of t(w).

The extension from a vector to a matrix estimate is just a matter of relabelling: a single sum becomes a double sum. The first examples of this we know of are in Withers and Nadarajah (2011a,

2011b, 2011c, 2012b, 2020). The extension to a *complex* scalar or vector or matrix w was given in these same papers. The 1st of these 3 papers applied it to the multi-tone problem in electrical engineering, and the other 4 papers to channel capacity problems where  $\hat{w}$  is a weighted mean of complex matrix random variables, and n is no longer a sample size, but the number of transmitters or receivers.

A different type of extension can be obtained by identifying a sample mean  $\hat{w} = \bar{X}$  from a distribution F(x) with its empirical distribution  $F_n(x)$ , and t(w) with T(F), a smooth functional of F(x), such as the bivariate correlation.  $T(F_n)$  is a Type A estimate of T(F), and its cumulant coefficient can be read off those of  $t(\hat{w})$ . In this way one obtains the Edgeworth expansion for  $n^{1/2}(T(F_n) - T(F))$ . See Withers (1983) Withers and Nadarajah (2008, 2010c, 2012c, 2013a)

A caveat on the use of an Edgeworth expansion is that including more terms makes it more inaccurate in the tails. This is where the *tilted* expansions, also known as *saddlepoint*, or *small sample* expansions, become essential. Results for the density of  $Y_{nw}$  for a sample mean, were given in §5 of Daniels (1983) and §6 of Daniels (1987). Withers and Nadarajah (2010b) shows how the cumulant coefficients given in this paper can be used to obtain the tilted expansion for the distribution and density of *any* Type A estimate.

#### 8. Conclusion

Let  $\hat{w}$  be a Type A estimate of an unknown parameter  $w \in R^p$ . Its cumulant coefficients are defined by (1). They are the building blocks of the Edgeworth expansion (2) in powers of  $n^{-1/2}$  for its distribution of  $\hat{w}$ . n is typically the sample size. Those coefficients needed for the rth term,  $P_r(x)$ , are given in (6) and (7). Let  $t(\hat{w}) \in R^q$  be a smooth function of  $\hat{w}$ . Then it is a Type A estimate of  $t(w) \in R^q$ . This paper gives its cumulant coefficients in terms of those of  $\hat{w}$  and the derivatives of  $\hat{w}$ . Replacing the coefficients in (2) by these coefficients provides the Edgeworth expansion of  $n^{1/2}(t(\hat{w}) - t(w))$  to  $n^{-5/2}$ .

The tilted Edgeworth expansion for  $\hat{w}$  needed for accuracy in the tails, was given in Withers and Nadarajah (2010b) in terms of its cumulant coefficients. Replacing these by those of  $t(\hat{w})$  given here gives the tilted Edgeworth expansion for  $t(\hat{w})$ .

In many practical statistical estimation problems, simulations are a popular way to approximate distributions. However these generally have the severe limitation that the parameters chosen cannot represent the whole parametric landscape.

We have given a number of applications to electrical engineering. For example numerical comparisons of the 1st 3 approximations to channel capacity for multiple arrays were given in Withers and Nadarajah (2020). In that case p = 1, so that an expansion for the percentile was possible. Thre are a host of other practical applications possible to electrical engineering and other fields.

Finally we mention some possible future research directions. Chain rules for  $t(\hat{w})$  can be applied to obtain the cumulant coefficients of its Studentized form. This can be followed up with expansions for the coverage probability of confidence regions, and corrections making them more accurate. Cumulant coefficients can be applied to bias reduction, Bayesian inference, confidence regions and power. The Edgeworth expansion can give a negative value in the tails of a distribution. Tilted expansions avoid this. Another way to get around this, is to choose  $y \in R^p$  such that  $\nabla_{nx} = Prob.(Y_{nw} \le x + n^{-1/2}y) - \Phi_V(x)$  is  $O(n^{-1})$ . For p > 1 such y can be chosen in an infinite number of ways, so that it may be possible to choose it so that  $\nabla_{nx} = O(n^{-3/2})$  or smaller. One could also replace  $n^{-1/2}y$  by  $y(n) = n^{-1/2}y_0 + n^{-1}y_1$  say, giving more choices.

# Abbreviations

The following abbreviations are used in this manuscript:

MDPI Multidisciplinary Digital Publishing Institute

DOAJ Directory of open access journals

TLA Three letter acronym

LD linear dichroism

# Appendix A: Some comments on the references

Here we give some comments and corrections to some of our papers.

**Withers (1982):** To the expression for  $a_{22}$  on p.59 add  $c_{12} + 2c_{16}$  where

$$c_{12} = I_2 \binom{12}{01} = t_i t_j k_{1,k}^{ij} k_1^k, \ c_{16} = I_{11} \binom{12}{00} = t_i k_1^{ij} t_{jk} k_1^k.$$

This correction does not effect applications in which  $\hat{\omega}$  is unbiased, as in Withers (1982, 1988).

In the expression on p.60 for  $(a_{22})_2$ ,  $I_{31}\binom{23}{22}$  should be  $c_{36} = I_{31}\binom{23}{00}$ .

On p61, 4 lines before Table 1, replace n/2)<sup>r-1</sup> by n/2)<sup>1-r</sup>.

On p67 add to  $K_2^{ab}$ ,  $t_i^a t_{ib} k_{1k}^{ij} k_1^k$ . For r = 3, 4 see §4.

On p68 in (A3) replace  $\binom{r+2}{r}$  by  $\binom{r+2}{2}$ . Changing to the simpler notation of Withers and Nadarajah (2008), denote the expressions for  $I_2\binom{2}{0}$ , ...,  $I_{301}\binom{222}{000}$  and  $I_3\binom{22}{01}$ , ...,  $I_{31}\binom{222}{001}$  given on p.58–59 by  $\sigma^2 = V = V(w) = t_i k_1^{ij} t_j$ ,  $c_{01}$ ,  $c_{02}$ ,  $c_{21}$ , and  $c_{23}$ ,  $c_{11}$ ,  $c_{15}$ ,  $c_{19}$ ,  $c_{1,10}$ ,  $c_{31}$ ,  $c_{36}$ ,  $c_{3,10}$ ,  $c_{3,11}$ ,  $c_{22}$ ,  $c_{12}$ ,  $c_{14}$ ,  $c_{17}$ ,  $c_{32}$ ,  $c_{34}$ ,  $c_{35}$ ,  $c_{38}$ ,  $c_{39}$ . So the expressions on pp59-60 become

$$a_{11} = c_{01} + c_{02}/2, \ a_{32} = c_{21} + 3c_{23},$$

$$a_{22} = c_{11} + c_{15} + c_{19}/2 + c_{1,10} + c_{12} + 2c_{16} \text{ by Corollary 5.3,}$$

$$a_{43} = c_{31} + 12c_{36} + 12c_{3,10} + 4c_{3,11},$$

$$(a_{11})_1 = \sigma^{-1}(c_{01} + c_{02}/2) - \sigma^{-3}(c_{22}/2 + c_{23}), \ (a_{32})_2 = \sigma^{-3}(c_{21} - 3c_{22} - 3c_{23}),$$

$$(a_{22})_2 = \sum_{i=1}^3 V^{-i}A_i, \ (a_{43})_4 = \sum_{i=2}^3 V^{-i}B_i \text{ for}$$

$$A_1 = c_{11} - c_{14}/2 + c_{15} - 2c_{17} - c_{19}/2, \ A_2 = -(c_{01} + c_{02}/2)(c_{22} + 2c_{23}) - c_{32} - c_{34} + c_{35} - 2c_{36} - 4c_{38} + 2c_{39} - 2c_{3,10} - 2c_{3,11},$$

$$A_3 = 7(c_{22} + 2c_{23})^2/4, \ B_2 = c_{31} - 6c_{32} - 6c_{34} + 3c_{35} - 24c_{38} - 12c_{3,10} - 8c_{3,11},$$

$$B_3 = 6(c_{22} + 2c_{23})(-c_{21} + 3c_{22} + 3c_{23}).$$

We now illustrate how the results on p.60 were obtained. Let  $c'_{rs}$  denote  $c_{rs}$  when  $t(\hat{w})$  is replaced by its Studentized form  $t_{(0)}(\hat{w})$ . Then

$$(a_{22})_2 = c'_{11} + c'_{15} + c'_{19}/2 + c'_{1,10} + c'_{12} + 2c'_{16},$$
(A1)

The first few derivatives of  $t_{(0)}(\hat{\omega})$  at w, and of V(w), are

$$\begin{split} t_{0,i} &= V^{-1/2}t_{i}, \ t_{0,ij} = V^{-1/2}t_{ij} - V^{-3/2}(t_{i}V_{j} + t_{j}V_{i})/2, \\ t_{0,ijk} &= V^{-1/2}t_{ijk} - V^{-3/2}\sum_{ijk}^{3}(t_{ij}V_{k} + V_{ij}t_{k})/2 + 3V^{-5/2}\sum_{ijk}^{3}t_{i}V_{j}V_{k}/4, \\ V_{i} &= 2t_{ia}k_{1}^{ab}t_{b} + k_{1,i}^{ab}t_{a}t_{b}, \ \text{ and} \\ V_{ij} &= 2t_{ija}k_{1}^{ab}t_{b} + 2t_{ia}k_{1}^{ab}t_{jb} + 2\sum_{ij}^{2}t_{ia}k_{1,j}^{ab}t_{b} + k_{1,ij}^{ab}t_{a}t_{b}, \ \text{ where} \\ \\ \sum_{ij}^{2}a_{i}b_{j} &= a_{i}b_{j} + a_{j}b_{i}, \sum_{ijk}^{3}a_{ij}b_{k} = a_{ij}b_{k} + a_{ik}b_{j} + a_{jk}b_{i} \ \text{for} \ a_{ij} = a_{ji}. \\ \text{So} \ c_{11}' &= V^{-1}c_{11}, \ c_{15}' &= V^{-1}c_{15} - V^{-2}M_{1}, \\ c_{19}' &= V^{-1}c_{19} - 2V^{-2}M_{4} + V^{-3}(VM_{2} + M_{3}^{2})/2, \\ c_{1,10}' &= V^{-1}c_{1,10} - V^{-2}(M_{3}c_{02} + 2M_{4} + VM_{5} + 2M_{6})/2 + 3V^{-3}(VM_{2} + 2M_{3}^{2})/4, \\ c_{12}' &= V^{-1}c_{12}, \ c_{16}' &= V^{-1}c_{16} - V^{-2}(VM_{7} + c_{01}M_{3})/2, \ \text{ where} \\ \\ M_{1} &= t_{i}t_{j}k_{2}^{ijk}V_{k} = c_{32} + 2c_{36}, \ M_{2} &= V_{i}k_{1}^{ij}V_{j} = c_{35} + 4c_{39} + 4c_{3,10}, \\ M_{3} &= t_{i}k_{1}^{ij}V_{j} = c_{22} + 2c_{23} = \tilde{c}_{22} \ \text{say}, \ M_{4} &= t_{i}k_{1}^{ij}t_{jk}k^{kl}V_{l} = c_{39} + 2c_{3,10}, \\ M_{5} &= k_{1}^{ij}V_{ij} = 2c_{1,10} + 2c_{19} + c_{14} + 4c_{17}, \\ M_{6} &= t_{i}k_{1}^{ij}V_{jk}k_{1}^{kl}t_{l} = 2c_{3,10} + 2c_{3,11} + c_{34} + 4c_{38}, \\ M_{7} &= k_{1}^{i}V_{i} &= c_{12} + 2c_{16} \Rightarrow \\ c_{19}' &= V^{-1}c_{19} + 2V^{-2}(4c_{35} - c_{3,10}) + V^{-3}\tilde{c}_{22}^{2}/2, \\ c_{1,10}' &= \sum_{1}^{3}V^{-i}b_{i} \ \text{for} \ b_{1} &= -c_{14}/2 - 2c_{17} - c_{19}, \ b_{3} = 3\tilde{c}_{22}^{2}/2, \\ \end{cases}$$

Substitution into (A1) yields  $(a_{22})_2$ . The other  $a'_{ri} = (a_{ri})_r$  given on p.60 are obtained similarly.

 $b_2 = -\tilde{c}_{22}c_{02}/2 - c_{34} + 3c_{35}/4 - 4c_{38} + 2c_{39} - c_{3,10} - 2c_{3,11}$ 

### Withers (1984):

p393 In the 5 line expression for  $f_4(x, L)$ , replace  $(x^3 - x)/30$  by  $(x^3 - 3x)/30$ , and +2473)/7776 by +2473x)/7776.

p394 In (3.4) replace  $(x^2 - 1)/2$ , by  $(x^2 - 1)/6$ ,

p394 In line 2 of Section 4, 'of Section 2' should be 'of Section 3'.

The following corrigendum for a printer's error appeared in

Withers, C.S. (1986) Inl. Royal Statist. Soc. B, 48 p258:

The expression  $-l_3^4(252x^5 - 1688x^3 + 1511x)/7776$  should be added to the last line on p393.

That also gives  $g_5(y)$  and  $g_6(y)$  for the last line on p393.

# Withers (1987):

p2371 (2.4):  $\sum_{i=1}^{\infty} n^{-i}C_i$  need not converge. We only require an asymptotic expansion. The same is true for (3.2) p2375.

p2371, 3rd to last paragraph. Replace 'Appendix C, which also' by 'Appendix D. Appendix C' p2372, Example 2.2, line 2. Replace  $t^{2j}(\theta)$  by  $t^{(2j)}(\theta)$ , the 2j derivative of  $t(\theta)$ . In line 3 and in Example 3.1,  $(y)_j = y(y-1)\cdots(y-j+1)$ .

p2377 line 2: replace (1.2) by (2.4)

p2377 line 3: replace /N by /N

p2377 line 9: replace 'Section 2' by 'Section 3' Since  $E_1 = 0$ ,  $C_1$  of Example 3.2 p2376 is unaffected.

p2378: these expression for  $C_j$  are correct if  $\hat{\theta}$  is *unbiased*. In that case the terms on p2378 with a 1 in the top line are zero so that  $C_j$  has only  $m_j$  terms where  $m_1 = 1$ ,  $m_2 = 3$ ,  $m_3 = 6$ ,  $m_4 = 12$ . However if  $\hat{\theta}$  is *biased*, then these expression for  $C_j$  did not allow for contributions from replacing  $\theta$  by E  $\hat{\theta}$  in the cumulant coefficients  $k_j^{a_1 \cdots a_r}$  of (3.2). These are corrected in Withers, C.S. and Nadarajah, S. (Submitted), Bias-reduced estimates for parametric problems.

p2379 Appendix D. Add at start: For p = 1 see (3.4) of

Withers, C.S. and Nadarajah, S. (2013), Delta and jackknife estimates of low bias for functions of binomial and multinomial parameters. *Journal of Multivariate Analysis*, **118**, 138–147. DOI:10.1016/j.jmva.2013.02.006

#### Withers (1988):

p729: in the 10th line from the bottom, replace "their range  $1 \cdots p$ " by "their range  $1 \cdots k$ " p732 line 9:  $T^{a_1 a_1 \dots}_{x_1 x_2 \dots}$  should be  $T^{a_1 a_2 \dots}_{x_1 x_2 \dots}$ .

p734: in the expression for  $h_1$  in the 5th to last line, replace  $He_2$  by  $He_2/6$ .

p737: in line 11, "Sections 1 and 2 of Withers (1983a)" should read "Sections 1 and 2 of Withers (1983b)".

p741: in the 4th equation from the bottom, at the end of the line, replace  $[1^j]T_2^j$ , by  $[1^j]_1T_2^j$ , Withers and Nadarajah (2008):

p743 para 2, line 4. Replace 'about zero.' by 'about zero when G puts mass 1 at x.'

p754, 756. Replace  $w_{in_a}$  by  $w_{ina}$ . Different samples can have different weights.

p754 2nd to last line. The first term on RHS,  $c_{11}/2$ , should be  $c_{11}$ .

p755, line 6. There is a typesetting error in the first of the 2 lines for  $a_{220}$ . Replace the first line with

$$a_{220} = \sigma^{-2}(c_{11} - c_{12} - c_{14}/2 + c_{15} - 2c_{17} - c_{19}/2) - \sigma^{-4}[(c_{01} + c_{02}/2)(c_{22} + 2c_{23}) + c_{32}]$$

p756. The 3rd and 4th lines after (8), should be

$$[1^r]_a = \int T_F(^a_x)^r dF_a(x),$$

$$[1,12,2^r]_{a_1a_2} = \int \int T_F(^{a_1}_{x_1}) T_F(^{a_1a_2}_{x_1x_2}) T_F(^{a_2}_{x_2})^r dF_{a_1}(x_1) dF_{a_2}(x_2),$$

# Withers and Nadarajah (2009):

p 272. Line 3: Convergence of S(t) is not needed, since  $B_{rk}$  is a finite sum.

 $\kappa_r$  on LHS(1.1) should be  $k_r$ .

p 273 last paragraph: also  $f(x)/\phi(x)$  is only meaningful if X is dimension-free.

p 275. (2.8) is correct but since  $B_1 = B_2 = 0$ , (2.8) can also be written

$$\int f^2/\phi = 1 + \sum_{k=3}^{\infty} B_k^2/k!.$$

In the 5th line of Section 3 insert after  $B_{rk}(\alpha)$ , 'at  $\alpha_1 = \alpha_2 = 0$ '.

The first line of (3.1) should read

$$B_r = \sum (B_{rk}\epsilon^{r-2k}: 1 \le k \le r/3)$$

(3.2) can be written  $K_s = s - 2[s/3]$  where [x] is the integral part of x.

p276. In the expression for  $B_{10}$ ,  $B_{10,18}$  should be  $B_{10,1}$ .

p 277. In the 2nd to last line,  $b_{64}$  should be  $B_{61}$ .

p 278. In the expession for  $b_4$ , the first term should be doubled. In the expession for  $b_5$ ,  $b_{82}$  should be  $B_{82}$ 

#### Withers and Nadarajah (2010a):

p3. In 5th and 6th to last lines, replace =  $n^{-2}K_{ab} + O(n^{-2})$  by =  $n^{-1}K_{ab} + O(n^{-2})$ 

p5. 2 lines above Theorem 2.2, replace "third moments" by "third central moments"

p7, lines 2-3: delete "and its Studentised version"

p7, lines 3-4: delete "or  $n^{-1/2}(\beta_a - \beta_a)\hat{J}_{aa}^{-1/2}$ "

p7, line 7-10: delete from "So, a one-sided" to "by  $O(n^{-1/2})$ ;

p9. Move "Set  $\phi' = \cdots m = p + 1$ . on the last 2 lines of p9 and 1st line of p 10 to just before "Set" on p9 line 9.

p10, lines 14-15: replace " $g_{ij}$  where" by " $g_{ij}$ ." and move the rest of the sentence, " $g_{ij} = \cdots g_{N.ij}$ ..." to the line after (6.1) p9, preceded by the word "Set"

# Withers and Nadarajah (2010b):

p1129 line 7: replace  $b_k^{i_1...i_r}(Y_n)$  by  $b_k^{i_1...i_r}(Y)$ 

To the 9th to last line we can add

$$\tilde{P}_3(t,B) = e_3(t) + e_1(t)e_2(t) + e_1(t)^3/6,$$

From p1130 line 6 to the end of §5: replace s by p, the dimension of  $\theta$ .

p1130 line 7 is clearer we replace line 8 by

for 
$$p_{(n)}^{(k)}(y) = \partial_{k_1} \cdots \partial_{k_p} p_{(n)}(y)$$
,  $\partial_k = \partial/\partial y_k$ ,

p1130 line 9. replace  $H_{\nu+k}(y)$  by  $H_{\nu+k}(y,V)$ 

p1130 The 5th and 6th to last lines:

for example  $K_i^{i_1...i_r} = k_i^{(i_1...i_r)}(t) = \partial_{i_1} \cdots \partial_{i_r} k_j(t)$  where  $\partial_i = \partial/\partial t_i$ .

p1132. A note on Corollary 3.2. For the duality of I(x) and  $k_0(t)$  see p176 of McCullagh, P., (1987) *Tensor methods in statistics.* Chapman and Hall, London.

p1133. In line 14 replace  $H_{\nu+\lambda}(\theta, V_t)$  by  $H_{\nu+\lambda}(0, V_t)$ 

# Withers and Nadarajah (2014a):

p81. In (2.14), replace  $J_r(x)$  and  $J'_r(x)$  by  $J_r(x)/r!$  and  $J'_r(x)/r!$ .

p81. The 2nd line after (2.15) should read

$$J'_r(x) = \int_{V\bar{y} < V\bar{x}} \bar{H}_r(\bar{y}, \bar{V}) \phi_{\bar{V}}(\bar{y}) d\bar{y}$$

The next line is correct:

$$J'_r(x) = \int_{V\bar{y} < V\bar{x}} (-\partial_{\bar{y}})^r \phi_{\bar{V}}(\bar{y}) d\bar{y}.$$

p82. In (2.20), replace  $J_r(y_1, y_2)$  by  $J_r(y_1, y_2)/r!$ .

p 85. In Withers, C.S. and Nadarajah, S. (2009), replace 'via' by 'in terms of'.

# Withers and Nadarajah (2014c):

p 676. Multiply RHS of (1.13) by  $n^{1/2}$ . That is, replace it by

$$Y_{IK\theta} = (n/s_{2K\theta})^{1/2}(\hat{\theta} - s_{1I\theta}), J \ge 0, K \ge 1.$$

p 699. In the editing of the original paper of 64 pages down to 21 pages, some details had to be removed. Here are some more details for Theorem 1.2 after (1.24).

$$\nabla_{1} = 0 \text{ if } J \geq 1, \ \nabla_{2} = \bar{A}_{43}H_{2} \text{ if } K \geq 2,$$

$$\nabla_{3} = \bar{A}_{33}H_{2} + \bar{A}_{54}H_{4} \text{ if } J \geq 2, \ \nabla_{4} = \bar{A}_{44}H_{3} + \bar{A}_{65}H_{5} \text{ if } K \geq 3,$$

$$\nabla_{5} = \bar{A}_{34}H_{2} + \bar{A}_{55}H_{4} + \bar{A}_{76}H_{6} \text{ if } J \geq 3,$$

$$\nabla_{6} = \bar{A}_{45}H_{3} + \bar{A}_{66}H_{5} + \bar{A}_{87}H_{7} \text{ if } K \geq 4.$$

$$\nabla_{re} = 0 \text{ if } r \leq 3 \text{ for } e = h, f, g, \ \nabla_{4e} = [4^{2}]_{0} e(4^{2}),$$

$$\nabla_{5e} = [45]_{0} e(45) + [34]_{1} e(34),$$

$$\nabla_{6e} = \sum_{i} \{ [\pi]_{0} e(\pi) : \pi = 5^{2}, 46, 4^{3} \} + \sum_{i} \{ [\pi]_{1} e(\pi) : \pi = 4^{2}, 35 \} + [3^{2}]_{2} e(3^{2}),$$

where the  $e(\pi)$ ,  $[\pi]_i$  needed for  $e_r(x)$ ,  $1 \le r \le 6$ , are as follows.

$$h(ij\cdots)=H_{i+j+\cdots-1}$$
 so that  $h(1^{i_1}2^{i_2}\cdots)=H_{1i_1+2i_2+\cdots-1}$ .  
For  $r=4$ :  $[4^2]_0=\bar{A}_{43}^2/2!$ ,  $h(4^2)=H_7$ ,  $f(4^2)=H_7-H_1H_3^2$ ,  $g(4^2)=H_7-2H_3H_4+H_1H_3^2$ .

For 
$$r = 5$$
:  $[45]_0 = \bar{A}_{43}\bar{A}_{54}$ ,  $h(45) = H_8$ ,  
 $f(45) = H_8 - H_1H_3H_4$ ,  $g(45) = H_8 - H_3H_5 - H_4^2 + H_1H_3H_4$ ,  
 $[34]_1 = \bar{A}_{33}\bar{A}_{43}$ ,  $h(34) = H_6$ ,  $f(34) = H_6 - H_1H_2H_3$ ,  
 $g(34) = H_6 - H_2H_4 - H_3^2 + H_1H_2H_3$ .  
For  $r = 6$ :  $[5^2]_0 = \bar{A}_{54}^2/2!$ ,  $h(5^2) = H_9$ ,  
 $f(5^2) = H_9 - H_1H_4^2$ ,  $g(5^2) = H_9 - 2H_4H_5 + H_1H_4^2$ ,  
 $[46]_0 = \bar{A}_{43}\bar{A}_{65}$ ,  $h(46) = H_9$ ,  
 $f(46) = H_9 - H_1H_3H_5$ ,  $g(46) = H_9 - H_3H_6 - H_4H_5 + H_1H_3H_5$ ,  
 $[4^3]_0 = \bar{A}_{43}^3/3!$ ,  $h(4^3) = H_{11}$ ,  $f(4^3) = H_{11} - 3H_1H_3H_7 - H_2H_3^3 + 3H_1^2H_3^3$ ,  
 $g(4^3) = H_{11} - 3H_3H_8 - 3H_4H_7 + 3H_1H_3H_7 + 3H_3^2H_5 + 6H_3H_4^2$   
 $-9H_1H_3^2H_4 - H_2H_3^3 + 3H_1^2H_3^3$ ,  
 $[4^2]_1 = \bar{A}_{43}\bar{A}_{44}$ ,  
 $[35]_1 = \bar{A}_{33}\bar{A}_{54}$ ,  $f(35) = H_7 - H_1H_2H_4$ ,  $g(35) = H_7 - H_3H_4 - H_2H_5$ ,  
 $[3^2]_2 = \bar{A}_{33}^2/2!$ ,  $f(3^2) = H_5 - H_1H_2^2$ ,  $g(3^2) = H_5 - 2H_2H_3 + H_1H_2^2$ .

p 702 §2. In the 3rd equation of Theorem 2.1,  $\ln \Gamma(\mu)$  should be  $\ln \Gamma(m)$ . p 704. Disregard Table 3. Withers and Nadarajah (2015):

In (22) and the formulas for  $d_2, \dots, d_6$  that follow, replace  $d_r$  by

$$\bar{d}_r = rd_r = c_r(x)/(r-1)!.$$

As stated this gives  $c_r = c_r(x) = r!d_r$ . For example

$$c_2 = a_1$$
,  $c_3 = 2a_1^2 + a_2$ ,  $c_4 = 3!a_1^3 + 7a_1a_2 + a_3$ ,  
 $c_5 = 4!a_1^4 + 46a_1^2a_2 + 11a_1a_3 + 7a_2^2 + a_4$ ,  
 $c_6 = 5!a_1^5 + 326a_1^3a_2 + 101a_1^2a_3 + 127a_1a_2^2 + 16a_1a_4 + 25a_2a_3 + a_5$ .

In the first reference, [1], replace J. J. Alfredo by J. A. Jimenez.

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