

# APPROXIMATION OF THE DISTRIBUTION OF THE PRODUCT OF BETA VARIABLES BY A SINGLE BETA VARIABLE

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**1. Introduction.** In an article published elsewhere in the present issue of the *Annals of Mathematical Statistics* [1] the  $g$ -th moments of two statistical test criteria  $L_{mve}$  and  $L_{vc}$  were found to have the following expressions, respectively:

$$(1) \quad (k-1)^{\sigma(k-1)} \prod_{i=1}^{k-1} \left[ \frac{\Gamma(\frac{1}{2}(n-1-i) + g)}{\Gamma(\frac{1}{2}(n-1-i))} \right] \cdot \frac{\Gamma(\frac{1}{2}n(k-1))}{\Gamma(\frac{1}{2}n(k-1) + g(k-1))}$$

and

$$(2) \quad (k-1)^{\sigma(k-1)} \prod_{i=1}^{k-1} \left[ \frac{\Gamma(\frac{1}{2}(n-1-i) + g)}{\Gamma(\frac{1}{2}(n-1-i))} \right] \cdot \frac{\Gamma(\frac{1}{2}(n-1)(k-1))}{\Gamma(\frac{1}{2}(n-1)(k-1) + g(k-1))}.$$

If we denote by  $(a)_\sigma$  the expression  $a(a+1)(a+2) \cdots (a+g-1)$  and make use of the fact that

$$(3) \quad \Gamma(a+g) = \Gamma(a) \cdot (a)_\sigma$$

and

$$(4) \quad \Gamma(a+rg) = \Gamma(a) \cdot (a)_{r\sigma} = \Gamma(a) \cdot r^{r\sigma} \prod_{i=1}^r \left( \frac{a+i-1}{r} \right)_\sigma$$

where  $r$  is a positive integer, the two moments (1) and (2) reduce to

$$(5) \quad \frac{\prod_{i=1}^{k-1} \left( \frac{n}{2} + \frac{i-k-1}{2} \right)_\sigma}{\prod_{i=1}^{k-1} \left( \frac{n}{2} + \frac{i-1}{k-1} \right)_\sigma} \quad \text{and} \quad \frac{\prod_{i=1}^{k-1} \left( \frac{n-1}{2} + \frac{i-k}{2} \right)_\sigma}{\prod_{i=1}^{k-1} \left( \frac{n-1}{2} + \frac{i-1}{k-1} \right)_\sigma}$$

respectively.

For any given value of  $i$  ( $i = 1, 2, \dots, k-1$ ) the ratio

$$\frac{\left( \frac{n}{2} + \frac{i-k-1}{2} \right)_\sigma}{\left( \frac{n}{2} + \frac{i-1}{k-1} \right)_\sigma} \quad \text{or} \quad \frac{\left( \frac{n-1}{2} + \frac{i-k}{2} \right)_\sigma}{\left( \frac{n-1}{2} + \frac{i-1}{k-1} \right)_\sigma}$$

may be expressed in the form

$$\frac{\Gamma(p_i + g)}{\Gamma(p_i + q_i + g)}$$

which is the  $g$ -th moment of a beta variable  $u_i$  distributed according to

$$\frac{\Gamma(p_i + q_i)}{\Gamma(p_i)\Gamma(q_i)} u_i^{p_i-1} (1-u_i)^{q_i-1} du_i.$$



Each of the moments in (5) is therefore of the form

$$\prod_{i=1}^{k-1} \frac{\Gamma(p_i + g)}{\Gamma(p_i + q_i + g)}.$$

Thus,  $L_{m_{vc}}$  and  $L_{vc}$  are each distributed like the product of  $k - 1$  independent beta variables.

Each of the moments in (5) can be expressed in the general form

$$(6) \quad M_g = \frac{\prod_{i=1}^{r'} \left(\frac{1}{x} - A_i + 1\right)_g}{\prod_{i=1}^{r'} \left(\frac{1}{x} - B_i + 1\right)_g}$$

where  $x = \frac{2}{n}$  (or  $\frac{2}{n-1}$ ),  $A_i$  and  $B_i$  are real numbers.

Other likelihood ratio statistical test criteria which have been discussed in the literature have moments which can be expressed in the general form (6). For example, the likelihood ratio criterion  $L_1$  for testing the homogeneity of sample variances [2, Neyman and Pearson 1931] has moments of this type. The generalized  $L_1$  criterion for samples from a normal multivariate population [3, Wilks 1933] has such moments. The criterion for testing sphericity [4, Mauchly 1940] of a normal multivariate distribution has moments of this kind. All test criteria having this type of moment lie on the interval (0, 1). The exact distribution functions of the criteria, except possibly for  $r = 1$  or 2 in some cases, are very complicated.

The purpose of this note is to consider a method of finding a fractional power of the test criterion which is approximately distributed (in a sense to be described later) according to an incomplete beta (Pearson Type I) distribution function,

$$(7) \quad dF(u) = \frac{\Gamma(p + q)}{\Gamma(p)\Gamma(q)} u^{p-1} (1 - u)^{q-1} du$$

and to find the appropriate values of  $p$ ,  $q$ , and the exponent of the criterion.

**2. Generalized hypergeometric series as moment generating functions.**

Suppose  $L$  is a statistical test criterion, or more generally a random variable having as its  $g$ -th moment the expression (6). The moment generating function  $\varphi(t)$  of  $L$  can be expressed as

$$(8) \quad \varphi(t) = \sum_{g=0}^{\infty} M_g t^g = \frac{\prod_{i=1}^{r'} \left(\frac{1}{x} - A_i + 1\right)_g}{\prod_{i=1}^{r'} \left(\frac{1}{x} - B_i + 1\right)_g} \cdot t^g.$$

This can be written as

$$(9) \quad \varphi(t) = {}_{r'+1}F_{r'} \left[ \begin{matrix} 1, \frac{1}{x} - A_1, \dots, \frac{1}{x} - A_{r'}; t \\ \frac{1}{x} - B_1, \dots, \frac{1}{x} - B_{r'} \end{matrix} \right]$$

where  ${}_{r'+1}F_{r'}$  [ ] is a generalized hypergeometric series [5, Bailey 1935]. We shall not make explicit use of this fact; instead, we shall work with the coefficient of  $t^g$  in the series, i.e.,  $M_g$ .

Let us consider

$$(10) \quad \ln M_g = \sum_{i=1}^{r'} \ln \left( \frac{1}{x} - A_i + 1 \right)_g - \sum_{i=1}^{r'} \ln \left( \frac{1}{x} - B_i + 1 \right)_g.$$

To expand this in a power series in  $x$  consider a single term

$$(11) \quad \begin{aligned} \ln \left( \frac{1}{x} - A + 1 \right)_g &= \sum_{j=1}^g \ln \left( \frac{1}{x} - A + j \right) \\ &= -g \ln x + g \ln (1 - Ax) + \sum_{j=1}^g \ln \left( 1 + \frac{jx}{1 - Ax} \right). \end{aligned}$$

Now

$$1 + \frac{jx}{1 - Ax} = 1 + jx + jAx^2 + jA^2x^3 + \dots$$

Writing

$$S_m(g) = \sum_{i=1}^g j^m,$$

and using the usual expansion for  $\ln(1+x)$ , we find

$$\begin{aligned} \ln \left( \frac{1}{x} - A + 1 \right) &= -g \ln x + [S_1(g) - Ag]x + \left[ \frac{1}{2}A^2 + AS_1(g) - \frac{1}{2}S_2(g) \right]x^2 \\ &\quad + \left[ -\frac{1}{6}A^3 + A^2S_1(g) - AS_2(g) + \frac{1}{6}S_3(g) \right]x^3 + \dots \end{aligned}$$

Applying this expansion to the separate terms in (10) and writing

$$(12) \quad C_m = \sum_{i=1}^{r'} A_i^m - \sum_{i=1}^{r'} B_i^m$$

the terms not involving  $A_i$  or  $B_i$  cancel out leaving

$$(13) \quad \begin{aligned} \ln M_g &= (-C_1g)x + \left[ \frac{1}{2}C_2 + C_1S_1(g) \right]x^2 \\ &\quad + \left[ -\frac{1}{6}C_3 + C_2S_1(g) - C_1S_2(g) \right]x^3 + \dots \end{aligned}$$

We shall return to this expression later.

**3. Powers of a beta variable.** If  $u$  has (7) as its distribution function, then

$$(14) \quad E(u^h) = \frac{(p)_h}{(p+q)_h}.$$

If  $v = u^r$ ,  $r$  integral, then its  $g$ -th moment is given by setting  $h = rg$  in (14).

We have

$$E(v^g) = \frac{(p)_{rg}}{(p+q)_{rg}}$$

But

$$(p)_{rg} = r^{rg} \left(\frac{p}{r}\right)_g \cdot \left(\frac{p+1}{r}\right)_g \dots \left(\frac{p+r-1}{r}\right)_g$$

so that

$$(15) \quad E(v^g) = \frac{\prod_{i=1}^r \left(\frac{p+i-1}{r}\right)_g}{\prod_{i=1}^r \left(\frac{p+q+i-1}{r}\right)_g}$$

which is a special case of (6) when  $p$  is of order  $n$ .

Putting  $\frac{1}{x} = \frac{p+q}{r}$ ,  $A_i = 1 + (q-i+1)/r$ , and  $B_i = 1 - (i-1)/r$

we have

$$C_1 = q,$$

$$C_2 = \frac{q^2}{r} + q \left(1 + \frac{1}{r}\right).$$

For any given moment of the form (6), from which  $x$ ,  $C_1$ , and  $C_2$  can be computed, we determine  $p$ ,  $q$ , and  $r$  so as to satisfy

$$(16) \quad \frac{p+q}{r} = \frac{1}{x}$$

$$q = C_1$$

and to satisfy, as nearly as possible, (with  $r$  integral)

$$(17) \quad \frac{q^2}{r} + q \left(1 + \frac{1}{r}\right) = C_2,$$

i.e.,

$$r = \frac{q(q+1)}{C_2 - q}.$$

The use of fractional  $r$  is obviously suggested, but its value and validity are not discussed here. Using the values of  $p$ ,  $q$  and  $r$  thus obtained, the distribution of the criterion  $L$  (having moments (7)), is given approximately by

$$(18) \quad \frac{\Gamma(p+q)}{\Gamma(p)\Gamma(q)} (\sqrt[r]{L})^{p-1} (1 - \sqrt[r]{L})^{q-1} d(\sqrt[r]{L})$$

where the approximation is such that all moments are correct through terms of order  $\left(\frac{r}{p+q}\right)$  (when moments are expanded in series of  $\frac{r}{p+q}$ ) and nearly (exactly if there is an integral value of  $r$  satisfying (17)) correct through terms of order  $\left(\frac{r}{p+q}\right)^2$ .

**4. Examples.** Returning to the  $g$ -th moment of  $L_{mve}$  given by the first expression in (5) we have

$$x = \frac{2}{n}, \quad r' = k - 1$$

$$A_i = \frac{k+3-i}{2}, \quad B_i = \frac{k-i}{k-1},$$

$$C_1 = \sum_{i=1}^{k-1} A_i - \sum_{i=1}^{k-1} B_i = \frac{1}{2}(k^2 + 3k - 6)$$

$$C_2 = \sum_{i=1}^{k-1} A_i^2 - \sum_{i=1}^{k-1} B_i^2 = \frac{1}{24}[(k+2)(k+3)(2k+5) - 84] - \frac{k(2k-1)}{6(k-1)}.$$

To determine  $p$ ,  $q$  and  $r$  for the fitted distribution of  $L_{mve}$  we set

$$\frac{p+q}{r} = \frac{n}{2}$$

$$q = \frac{1}{2}(k^2 + 3k - 6)$$

$$r = \frac{q(q+1)}{C_2 - q}$$

and solve for  $p$ ,  $q$  and  $r$ . We have the following table of values,  $p$ ,  $q$  and  $r$  for various values of  $k$  ( $p$  being calculated by using the rounded values of  $r$ ):

$k$	$r$	$r$ (rounded)	$p$	$q$
3	2	2	$n - 3$	3
4	2.93	3	$1.5n - 5.5$	5.5
5	3.82	4	$2n - 8.5$	8.5
6	4.68	5	$2.5n - 12$	12
7	5.53	6	$3n - 16$	16
8	6.35	6	$3n - 20.5$	20.5
9	7.17	7	$3.5n - 25.5$	25.5
10	7.96	8	$4n - 31$	31
20	15.76	16	$8n - 113.5$	113.5

Thus, by rounding  $r$  off to the nearest integer and using this rounded value of  $r$  in determining  $p$ , we have values of  $p$ ,  $q$  and  $r$  for each value of  $k$ , which, when substituted in (18) give us the desired fitted beta distribution for  $L_{mve}$ . For  $k = 3$ , the fitted distribution is the exact distribution.

For the  $g$ -th moment of  $L_{ve}$  which is given by the second expression in (5), it is convenient to expand in powers of  $\frac{2}{n-1}$ . Hence we have

$$\begin{aligned}
 x &= \frac{2}{n-1}, & r' &= k-1 \\
 A_i &= \frac{k+2-i}{2}, & B_i &= \frac{k-i}{k-1} \\
 C_1 &= \frac{1}{4}(k^2 + k - 4) \\
 C_2 &= \frac{1}{24}[(k+1)(k+2)(2k+3) - 30] - \frac{k(2k-1)}{6(k-1)}.
 \end{aligned}$$

To determine  $p$ ,  $q$  and  $r$  for fitting the distribution function of  $L_{ve}$  we put

$$\begin{aligned}
 \frac{p+q}{r} &= \frac{n-1}{2} \\
 q &= \frac{1}{4}(k^2 + k - 4) \\
 r &= \frac{q(q+1)}{C_2 - q}.
 \end{aligned}$$

We have the following table of values of  $p$ ,  $q$  and  $r$  for several values of  $k$ :

$k$	$r$	$r$ (rounded)	$p$	$q$
3	2	2	$n-3$	2
4	2.88	3	$1.5n-5.5$	4
5	3.71	4	$2n-8.5$	6.5
6	4.52	5	$2.5n-12$	9.5
7	5.32	5	$2.5n-15.5$	13
8	6.14	6	$3n-20$	17
9	6.88	7	$3.5n-25$	21.5
10	7.82	8	$4n-30.5$	26.5
20	15.26	15	$7.5n-111.5$	104

By rounding  $r$  off to the nearest integer, and using the rounded value of  $r$  in determining  $p$ , we have values of  $p$ ,  $q$  and  $r$  for each value of  $k$  which, when substituted in (18), give us the desired fitted beta distribution for  $L_{ve}$ . For  $k = 3$ , the fitted distribution is the exact distribution.

For a given value of  $k$ , approximate 5% and 1% points of  $\sqrt[k]{L_{mve}}$  and  $\sqrt[k]{L_{ve}}$  can therefore be obtained from Thompson's [6] tables of the Incomplete Beta Function by entering the tables with  $\nu_1 = 2q$ , and  $\nu_2 = 2p$ . For example, for  $k = 6$  the 5% and 1% points of  $\sqrt[6]{L_{mve}}$  are obtained by entering Thompson's tables with  $\nu_1 = 24$ , and  $\nu_2 = 5n - 24$ .

## REFERENCES

- [1] S. S. WILKS, "Sample criteria for testing equality of means, equality of variances and equality of covariances in a normal multivariate population," *Annals of Math. Stat.* Vol. 17 (1946) pp. 257-281.
- [2] J. NEYMAN and E. S. PEARSON, "On the problem of  $k$  samples," *Bulletin de l'Académie Polonaise des Sciences et des Lettres, Série A, Sciences Mathématiques, 1930 and 1931*, pp. 460-481.
- [3] S. S. WILKS, "Certain generalizations in the analysis of variance," *Biometrika*, Vol. 24 (1932), pp. 471-494.
- [4] JOHN W. MAUCHLY, "Significance test for sphericity of a normal  $n$ -variate distribution," *Annals of Math. Stat.*, Vol. 11 (1940), pp. 204-209.
- [5] W. N. BAILEY, *Generalized Hypergeometric Series*, Cambridge Tracts in Mathematics and Mathematical Physics, No. 32, Cambridge University Press, 1935.
- [6] CATHERINE M. THOMPSON, "Tables of percentage points of the Incomplete Beta Function," *Biometrika*, Vol. 32, Part II (1941), pp. 187-191.