

Further Results on the Non-Central Multivariate
Beta Distribution and Moments of Traces of Two Matrices

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C. G. Khatri and K.C.S. Pillai

Gujarat University and Purdue University

Department of Statistics

Division of Mathematical Sciences

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1. Introduction and Summary. Let A_1 and A_2 be two symmetric matrices of order p , A_1 positive definite and having a Wishart distribution [2,17] with f_1 degrees of freedom, and A_2 , at least positive semi-definite and having a (pseudo) non-central (linear) Wishart distribution [1,3,4,17,18] with f_2 degrees of freedom. Now let

$$A_2 = C \underline{Y} \underline{Y}' C'$$

where \underline{Y} is $p \times f_2$ and C is a lower triangular matrix such that

$$A_1 + A_2 = C C' .$$

Now Pillai's $V^{(s)}$ criterion [9,10,11,12] is the sum of the non-zero characteristic roots of the matrix $\underline{Y} \underline{Y}'$. Here s is minimum (f_2, p) . Also we may note that $V^{(s)} = \text{trace } \underline{Y} \underline{Y}' = \text{trace } \underline{Y}' \underline{Y}$. It can be shown that the density function of the characteristic roots of $\underline{Y}' \underline{Y}$ for $f_2 \leq p$ can be obtained from that of the characteristic roots of $\underline{Y} \underline{Y}'$ for $f_2 \geq p$ if in the latter case the following changes are made: [5,17]

$$(f_1, f_2, p) \longrightarrow (f_1 + f_2 - p, p, f_2).$$

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Hence we only consider $V^{(s)}$ when $s = p$, i.e. $V^{(p)}$, based on the density function [7] of $\underline{L} = \underline{Y} \underline{Y}'$ for $f_2 \geq p$. The first four moments of $V^{(2)}$ in the linear case were obtained by Pillai [13] and the first two moments of $V^{(p)}$ by the authors in an earlier paper [6]. In the present paper the third and fourth moments of $V^{(p)}$ are derived using a method which differs in certain respects from the previous one.

Further, if $U^{(p)} = \text{tr}(\underline{I}_p - \underline{L})^{-1}$, [9, 10, 11, 12, 16] the first four moments of $U^{(2)}$ were derived by Pillai [13] and the first two moments of $U^{(p)}$ by the authors [6]. These results are extended in the present paper, obtaining the third and fourth moments of $U^{(p)}$ and further, some approximations to the distribution of $U^{(p)}$ are suggested in the linear case.

2. Moments of $V^{(p)}$. In the previous paper [6] the authors have shown that

$$(2.1) \quad \begin{aligned} V^{(p)} &= l_{11} + (1-l_{11}) \underline{u}'(\underline{I}_{p-1} - \underline{L}_{22}) \underline{u} + \text{tr } \underline{L}_{22} \\ &= -(1-l_{11})(1-\underline{u}'\underline{u} + \underline{u}'\underline{L}_{22}\underline{u}) + \text{tr } \underline{L}_{22} + 1 \end{aligned}$$

where l_{11} , \underline{u} : $(p-1) \times 1$ and \underline{L}_{22} : $(p-1) \times (p-1)$ are independently distributed and their respective distributions are given by

$$(2.2) \quad \exp(-\lambda^2) \sum_{j=0}^{\infty} \frac{(\lambda^2)^j}{j!} \frac{l_{11}^{\frac{1}{2}f_2+j-1} (1-l_{11})^{\frac{1}{2}f_1-1}}{\beta \left[\frac{1}{2}f_2+j, \frac{1}{2}f_1 \right]} d l_{11} ,$$

$$(2.3) \quad \pi^{-\frac{1}{2}(p-1)} \left\{ \Gamma\left(\frac{f_1-p+1}{2}\right) \right\}^{-1} \Gamma\left(\frac{1}{2}f_1\right) (1-\underline{u}'\underline{u})^{\frac{1}{2}(f_1-p+1)-1} d\underline{u}$$

and

$$(2.4) \quad \prod_{i=1}^{p-1} \left[\Gamma\left(\frac{f_1+f_2-i}{2}\right) / \left\{ \Gamma\left(\frac{f_1-i+1}{2}\right) \Gamma\left(\frac{f_2-i}{2}\right) \right\} \right] \pi^{-\frac{1}{4}(p-1)(p-2)}$$

$$\left| \underline{L}_{22} \right|^{\frac{1}{2}\{f_2-1-(p-1)-1\}} \left| \underline{I}_{p-1-\underline{L}_{22}} \right|^{\frac{1}{2}\{f_1-(p-1)-1\}} d\underline{L}_{22} .$$

Let λ_i 's ($i=1,2,\dots, p-1$) be the characteristic roots of \underline{L}_{22} . Then, noting that (2.3) is invariant under an orthogonal transformation of \underline{u} we can write without loss of generality the statistic $v^{(p)}$ of (2.1) to be equivalent to

$$(2.5) \quad v^{(p)} - 1 = \sum_{i=1}^{p-1} \lambda_i^{-1} (1-\ell_{11}) \left[(1-\underline{u}'\underline{u}) + \sum_{i=1}^{p-1} \lambda_i u_i^2 \right] .$$

Let $\ell_{11,0}$ be a variate whose distribution is the same as that of ℓ_{11} when $\lambda = 0$ and independently distributed of \underline{u} and \underline{L}_{22} . Let $v_0^{(p)}$ be the $v^{(p)}$ statistic when $\lambda = 0$. We may note that

$$(2.6) \quad x_1 = E(1-\ell_{11,0}) - E(1-\ell_{11}) = f_1 \delta(v) ,$$

$$(2.7) \quad x_2 = E(1-\ell_{11,0})^2 - E(1-\ell_{11})^2 = \frac{f_1(f_1+2)}{2} \Delta_1 ,$$

$$(2.8) \quad x_3 = E(1-\ell_{11,0})^3 - E(1-\ell_{11})^3 = \frac{1}{8} f_1(f_1+2)(f_1+4) \Delta_2$$

and

$$(2.9) \quad x_4 = E(1-\ell_{11,0})^4 - E(1-\ell_{11})^4 = \frac{1}{48} f_1(f_1+2)(f_1+4)(f_1+6) \Delta_3$$

where $v = f_1 + f_2$,

$$\begin{aligned}
 (2.10) \quad \delta(v) &= \frac{\lambda^2}{v} \exp(-\lambda^2) \sum_{i=0}^{\infty} \frac{(\lambda^2)^i}{i!} \frac{1}{\frac{1}{2}v+i+1} = \frac{\lambda^2}{v} \int_0^1 (1-y)^{\frac{1}{2}v} \exp(-\lambda^2 y) dy \\
 &= \frac{2\lambda^2}{v} \sum_{i=0}^{\infty} \frac{(2\lambda^2)^i (-1)^i}{(v+2)(v+4)\dots(v+2i+2)} \quad \text{if } \lambda^2 < \frac{1}{2}v+1. \\
 &= \frac{1}{v} \left[\sum_{i=0}^{\frac{1}{2}v} \binom{\frac{1}{2}v}{i} \frac{(-1)^i (i!)}{(\lambda^2)^i} - \frac{(-1)^{\frac{1}{2}v} (\frac{1}{2}v!) \exp(-\lambda^2)}{(\lambda^2)^{\frac{1}{2}v}} \right] \quad \text{if } \frac{1}{2}v \text{ is an}
 \end{aligned}$$

integer,

$$(2.11) \quad \Delta_1 = \delta(v) - \delta(v+2), \Delta_2 = \delta(v) - 2\delta(v+2) + \delta(v+4) \text{ and}$$

$$\Delta_3 = \delta(v) - 3\delta(v+2) + 3\delta(v+4) - \delta(v+6).$$

The results (2.6) ---- (2.9) are obtained by using the partial fractions

for $\frac{1}{v(v+2)(v+4)\dots}$

Let E_1 stand for the expectation over u_i 's. Let

$$\beta_1 = \sum_{i=1}^{p-1} \lambda_i u_i^2. \quad \text{It is easy to see from (2.3) that for } i \neq j \neq k \neq \ell,$$

$$\begin{aligned}
 (2.12) \quad & E(u_i^2) = \frac{1}{f_1}, \quad E(u_i^4) = \frac{3}{f_1(f_1+2)}, \quad E(u_i^6) = \frac{15}{f_1(f_1+2)(f_1+4)}, \\
 & E(u_i^3) = \frac{105}{f_1(f_1+2)(f_1+4)(f_1+6)}, \\
 & E(u_i^2 u_j^2) = \frac{1}{f_1(f_1+2)}, \quad E(u_i^4 u_j^2) = \frac{3}{f_1(f_1+2)(f_1+4)}, \\
 & E(u_i^6 u_j^2) = \frac{15}{f_1(f_1+2)(f_1+4)(f_1+6)}, \\
 & E(u_i^4 u_j^4) = \frac{9}{f_1(f_1+2)(f_1+4)(f_1+6)}, \quad E(u_i^2 u_j^2 u_k^2) = \frac{1}{f_1(f_1+2)(f_1+4)}, \\
 & E(u_i^4 u_j^2 u_k^2) = \frac{3}{f_1(f_1+2)(f_1+4)(f_1+6)} \\
 & \text{and } E(u_i^2 u_j^2 u_k^2 u_\ell^2) = \frac{1}{f_1(f_1+2)(f_1+4)(f_1+6)}.
 \end{aligned}$$

Then putting $\sum_{i=1}^{p-1} \lambda_i = \text{tr } L_{22}$, $\sum_{i>j} \lambda_i \lambda_j = \text{tr}_2 L_{22}$, $\sum_{i>j>k} \lambda_i \lambda_j \lambda_k = \text{tr}_3 L_{22}$

and $\sum_{i>j>k>\ell} \lambda_i \lambda_j \lambda_k \lambda_\ell = \text{tr}_4 L_{22}$ and using (2.12), we get

$$(2.13) \quad E_1(\beta_1) = (\text{tr } L_{22}) / f_1,$$

$$\begin{aligned}
 (2.14) \quad E_1(\beta_1^2) &= \frac{1}{f_1(f_1+2)} \left[3 \sum_i \lambda_i^2 + 2 \sum_{i>j} \lambda_i \lambda_j \right] \\
 &= \frac{1}{f_1(f_1+2)} \left[3(\text{tr } L_{22})^2 - 4(\text{tr}_2 L_{22}) \right],
 \end{aligned}$$

$$\begin{aligned}
 (2.15) \quad E_1(\beta_1^3) &= \frac{1}{f_1(f_1+2)(f_1+4)} \left[15 \sum_i \lambda_i^3 + 9 \sum_{i \neq j} \lambda_i^2 \lambda_j + 6 \sum_{i > j > k} \lambda_i \lambda_j \lambda_k \right] \\
 &= \frac{1}{f_1(f_1+2)(f_1+4)} \left[15(\text{tr} L_{22})^3 - 36(\text{tr} L_{22})(\text{tr}_2 L_{22}) + 24(\text{tr}_3 L_{22}) \right],
 \end{aligned}$$

and

$$\begin{aligned}
 (2.16) \quad E_1(\beta_1^4) &= \frac{1}{f_1(f_1+2)(f_1+4)(f_1+6)} \left[105 \sum_i \lambda_i^4 + 60 \sum_{i \neq j} \lambda_i^3 \lambda_j \right. \\
 &\quad \left. + 27 \sum_{i \neq j} \lambda_i^2 \lambda_j^2 + 18 \sum_{i \neq j \neq k} \lambda_i^2 \lambda_j \lambda_k + \sum_{i \neq j \neq k \neq \ell} \lambda_i \lambda_j \lambda_k \lambda_\ell \right], \\
 &= \frac{1}{f_1(f_1+2)(f_1+4)(f_1+6)} \left[105(\text{tr} L_{22})^4 - 360(\text{tr} L_{22})^2 (\text{tr}_2 L_{22})^2 \right. \\
 &\quad \left. + 288(\text{tr} L_{22})(\text{tr}_3 L_{22}) - 192(\text{tr}_4 L_{22}) + 144(\text{tr}_2 L_{22})^2 \right],
 \end{aligned}$$

because

$$\sum_i \lambda_i^3 = (\text{tr} L_{22})^3 - 3 \text{tr} L_{22} \text{tr}_2 L_{22} + 3 \text{tr}_3 L_{22}$$

$$\sum_{i \neq j} \lambda_i^2 \lambda_j = (\text{tr} L_{22})(\text{tr}_2 L_{22}) - 3(\text{tr}_3 L_{22}),$$

$$\sum_{i \neq j} \lambda_i \lambda_j^3 = (\sum_i \lambda_i^3)(\sum_j \lambda_j) - \sum_i \lambda_i^4$$

$$= (\text{tr} L_{22})^4 - 3(\text{tr} L_{22})^2 (\text{tr}_2 L_{22}) + 3(\text{tr} L_{22})(\text{tr}_3 L_{22}) - \sum_i \lambda_i^4,$$

$$\sum_{i \neq j} \lambda_i^2 \lambda_j^2 = [(\text{tr} L_{22}) - 2(\text{tr}_2 L_{22})]^2 - \sum \lambda_i^4,$$

$$\sum_{i \neq j \neq k} \lambda_i^2 \lambda_j \lambda_k = 2(\text{tr} L_{22})^2 (\text{tr}_2 L_{22}) - 4(\text{tr}_2 L_{22})^2 - 2 \sum_{i \neq j} \lambda_i^3 \lambda_j,$$

and therefore

$$\sum_i \lambda_i^4 = (\text{tr} L_{22})^4 - 4(\text{tr} L_{22})^2 (\text{tr}_2 L_{22}) + 4(\text{tr} L_{22}) (\text{tr}_3 L_{22}) - 4(\text{tr}_4 L_{22}) + 2(\text{tr}_2 L_{22})^2,$$

$$\sum_{i \neq j} \lambda_i^3 \lambda_j = (\text{tr} L_{22})^2 (\text{tr}_2 L_{22}) - (\text{tr} L_{22}) (\text{tr}_3 L_{22}) + 4(\text{tr}_4 L_{22}) - 2(\text{tr}_2 L_{22})^2,$$

$$\sum_{i \neq j} \lambda_i^2 \lambda_j^2 = 2(\text{tr}_2 L_{22})^2 - 4(\text{tr} L_{22}) (\text{tr}_3 L_{22}) + 4 \text{tr}_4 L_{22}, \text{ and}$$

$$\sum_{i \neq j \neq k} \lambda_i^2 \lambda_j \lambda_k = 2(\text{tr} L_{22}) (\text{tr}_3 L_{22}) - 8(\text{tr}_4 L_{22}).$$

With the help of above results, we can write

$$(2.17) \quad E[V^{(p)} - 1] = E(V_0^{(p)} - 1) + x_1 E(\beta)$$

$$(2.18) \quad E[V^{(p)} - 1]^2 = E(V_0^{(p)} - 1)^2 - x_2 E(\beta^2) + 2x_1 E(\alpha\beta)$$

$$(2.19) \quad E(V^{(p)} - 1)^3 = E(V_0^{(p)} - 1)^3 + x_3 E(\beta^3) - 3x_2 E(\alpha\beta^2) + 3x_1 E(\alpha^2\beta)$$

and

$$(2.20) \quad E(V^{(p)} - 1)^4 = E(V_0^{(p)} - 1)^4 - x_4 E(\beta^4) + 4x_3 E(\beta^3 \alpha) - 6x_2 E(\beta^2 \alpha^2) \\ + 4x_1 E(\beta \alpha^3)$$

where $\alpha = (\text{tr} L_{22})$ and $\beta = 1 - \underline{u}'\underline{u} + \beta_1$.

We note that

$$(2.21) \quad E(\alpha^i \beta) = \frac{1}{f_1} E[(\text{tr} L_{22})^i f^{(1)} + (\text{tr} L_{22})^{i+1}] \\ = \frac{1}{f_1} r_i^{(1)}; \quad \text{for } i=0,1,2,3$$

$$(2.22) \quad E(\alpha^i \beta^2) = \frac{1}{f_1(f_1+2)} E[(\text{tr} L_{22})^i \{f^{(2)} + 2f^{(1)}(\text{tr} L_{22}) + 3(\text{tr} L_{22})^2 \\ - 4(\text{tr}_2 L_{22})\}] = \frac{1}{f_1(f_1+2)} r_i^{(2)} \quad \text{for } i=0,1,2;$$

$$(2.23) \quad E(\alpha^i \beta^3) = \frac{1}{f_1(f_1+2)(f_1+4)} E[(\text{tr} L_{22})^i \{f^{(3)} + 3f^{(2)}(\text{tr} L_{22}) \\ + 3f^{(1)}\{3(\text{tr} L_{22})^2 - 4(\text{tr}_2 L_{22})\} + 15(\text{tr} L_{22})^3 \\ - 36(\text{tr} L_{22})(\text{tr}_2 L_{22}) + 24 \text{tr}_3 L_{22}\}] = \frac{1}{f_1(f_1+2)(f_1+4)} r_i^{(3)} \\ \text{for } i = 0, 1,$$

and

$$\begin{aligned}
(2.24) \quad E(\beta^4) &= \frac{1}{f_1(f_1+2)(f_1+4)(f_1+6)} E[f^{(4)} + 4f^{(3)}(\text{tr}L_{22}) \\
&\quad + 6f^{(2)}\{3(\text{tr}L_{22})^2 - 4(\text{tr}_2L_{22})\} + 4f^{(1)}\{15(\text{tr}L_{22})^3 \\
&\quad - 36(\text{tr}L_{22})(\text{tr}_2L_{22}) + 24(\text{tr}_3L_{22})\} + 105(\text{tr}L_{22})^4 \\
&\quad - 360(\text{tr}L_{22})^2(\text{tr}_2L_{22}) + 288(\text{tr}L_{22})(\text{tr}_3L_{22}) - 192(\text{tr}_4L_{22}) \\
&\quad + 144(\text{tr}_2L_{22})^2] = \frac{1}{f_1(f_1+2)(f_1+4)(f_1+6)} r_0^{(4)},
\end{aligned}$$

where

$$(2.25) \quad f^{(i)} = \prod_{j=1}^i (f_1 - p - 1 + 2j).$$

Hence, we have

$$(2.26) \quad E(V^{(p)}_{-1}) = E(V_0^{(p)}_{-1}) + \delta(v) r_0^{(1)},$$

$$(2.27) \quad E(V^{(p)}_{-1})^2 = E(V_0^{(p)}_{-1})^2 + \delta(v)[2r_1^{(1)} - \frac{1}{2}r_0^{(2)}] + (\frac{1}{2}r_0^{(2)})\delta(v+2),$$

$$\begin{aligned}
(2.28) \quad E(V^{(p)}_{-1})^3 &= E(V_0^{(p)}_{-1})^3 + \delta(v)[3r_2^{(1)} - \frac{3}{2}r_1^{(2)} + \frac{1}{8}r_0^{(3)}] \\
&\quad + \delta(v+2)[\frac{3}{2}r_1^{(2)} - \frac{1}{4}r_0^{(3)}] + \delta(v+4)[\frac{1}{8}r_0^{(3)}],
\end{aligned}$$

and

$$\begin{aligned}
(2.29) \quad E(V^{(p)}_{-1})^4 &= E(V_0^{(p)}_{-1})^4 + \delta(v) [4r_3^{(1)} - 3r_2^{(2)} + \frac{1}{2}r_1^{(3)} - \frac{1}{48}r_0^{(4)}] \\
&\quad + \delta(v+2) [9r_2^{(2)} - r_1^{(3)} + \frac{1}{16}r_0^{(4)}] + \delta(v+4) \\
&\quad [\frac{1}{2}r_1^{(3)} - \frac{1}{16}r_0^{(4)}] + \delta(v+6) [\frac{1}{48}r_0^{(4)}] .
\end{aligned}$$

Expressions for the first two moments of $V^{(p)}$ are presented in [6]. For the third moment

$$\begin{aligned}
(2.30) \quad E(V^{(p)}_{-1})^3 &= E(V_0^{(p)}_{-1})^3 + \Delta_2 f^{(3)}/8 \\
&\quad + (3/8) [\{-3\delta(v) + 2\delta(v+2) + \delta(v+4)\} f^{(2)} E(\text{tr}_{\tilde{L}_{22}}^{(2)}) \\
&\quad + \{3\delta(v) + 2\delta(v+2) + 3\delta(v+4)\} f^{(1)} E(\text{tr}_{\tilde{L}_{22}}^{(1)})^2 \\
&\quad + \{\delta(v) + 2\delta(v+2) + 5\delta(v+4)\} E(\text{tr}_{\tilde{L}_{22}})^3 \\
&\quad + 4\{\delta(v) + 2\delta(v+2) - 3\delta(v+4)\} E(\text{tr}_{\tilde{L}_{22}} \text{tr}_{\tilde{L}_{22}}) \\
&\quad + 4\Delta_2 \{-f^{(1)} E(\text{tr}_{\tilde{L}_{22}}) + 2E(\text{tr}_{\tilde{L}_{22}})\}] ,
\end{aligned}$$

where

$$\begin{aligned}
(2.31) \quad E(\text{tr}_{\tilde{L}_{22}}^{(i)}) &= \binom{p-1}{i} \prod_{j=1}^i [(f_2 - j)/(f_1 + f_2 - j)] \\
&\quad (i=1, 2, \dots, p-1; \text{tr}_{\tilde{L}_{22}} = \text{tr}_{\tilde{L}_{22}}) ,
\end{aligned}$$

$$(2.32) \quad E(\text{tr}L_{22}\text{tr}_iL_{22}) = [E(\text{tr}_iL_{22})/\{(f_1+f_2+1)(f_1+f_2-i-1)\}] .$$

$$\begin{aligned} & [(f_1-p)\{(f_2-1)(p-1)+2i\}+(p-1)(f_2-p-1)(f_2+2p-i+1) \\ & \quad +2p^3-ip^2+(i-2)p+2i] \end{aligned}$$

and

$$(2.33) \quad E(\text{tr}L_{22})^3 = E(\text{tr}L_{22}) \left[\frac{2^3(f_1-f_2+1)f_1(f_1+f_2-2p+1)(f_1+f_2-p)}{(f_1+f_2-3)(f_1+f_2-2)(f_1+f_2-1)^2(f_1+f_2+1)(f_1+f_2+3)} \right. \\ \left. +3 E(\text{tr}L_{22})^2 -2\{E(\text{tr}L_{22})\}^2 \right] .$$

Now, for the fourth moment

$$(2.34) \quad E\{V^{(p)}_{-1}\}^4 = E(V_0^{(p)}_{-1})^4 -\Delta_3 f^{(4)}/48 \\ + (1/i2)\{5\delta(v)-9\delta(v+2)+3\delta(v+4)+\delta(v+6)\}f^{(3)}E(\text{tr}L_{22}) \\ + (3/8)\{-5\delta(v)+19\delta(v+2)+\delta(v+4)+\delta(v+6)\}f^{(2)}E(\text{tr}L_{22})^2 \\ + (1/4)\{5\delta(v)+51\delta(v+2)+3\delta(v+4)+5\delta(v+6)\}f^{(1)}E(\text{tr}L_{22})^3 \\ + (1/16)\{5\delta(v)+297\delta(v+2)+15\delta(v+4)+35\delta(v+6)\}E(\text{tr}L_{22})^4 \\ + (3/2)\{\delta(v)-15\delta(v+2)+3\delta(v+4)-5\delta(v+6)\}E[(\text{tr}L_{22})^2\text{tr}_2L_{22}] \\ + 3\Delta\{-f^{(1)}E(\text{tr}L_{22}\text{tr}_2L_{22})+2E(\text{tr}L_{22}\text{tr}_3L_{22})\} \\ + \Delta_3\{(f^{(2)}/2)E(\text{tr}_2L_{22})-2f^{(1)}E(\text{tr}_3L_{22})+4E(\text{tr}_4L_{22}) \\ -3E(\text{tr}_2L_{22})^2\} ,$$

where $\Delta = \delta(v) - \delta(v+2) - \delta(v+4) + \delta(v+6)$,

$$\begin{aligned}
 (2.35) \quad E[(\text{tr}L_{22})^2 \text{tr}_2 L_{22}] &= \frac{(p-1)(p-2)(f_2-2)(f_2-1)}{12(f_1+f_2-3)(f_1+f_2-2)(f_1+f_2-1)(f_1+f_2)} \times \\
 &\left[\frac{pf_2}{2(f_1+f_2+3)} \{ (f_1-p)[(3p-5)(f_2-p-1)+3p^2-7p+14] \right. \\
 &\quad + (3p-5)(f_2-p-1)^2 + (9p^2-17p+6)(f_2-p-1) \\
 &\quad \left. + 2(3p^3-7p^2+6p+4) \right] \\
 &+ \frac{1}{(f_1+f_2-4)(f_1+f_2+1)} \{ (f_1-p)^2 [p(p-1)(f_2-p-1)^2 + (p-1)(2p^2+p+6)(f_2-p-1) \\
 &\quad + p^4+5p^2-6p+12] \\
 &\quad + (f_1-p)[2p(p-1)(f_2-p-1)^3 + (p-1)(8p^2+p+6)(f_2-p-1)^2 \\
 &\quad + (p-1)(10p^3+15p-6)(f_2-p-1) \\
 &\quad + 4p^5-3p^4+8p^3-15p^2+6p+12] \\
 &\quad \left. + p(p-1)(f_2-1)f_2(f_2+p-3)(f_2+p-2) \right\} \\
 &+ \frac{(f_2+p+1)}{(f_1+f_2+3)} [E(\text{tr}L_{22} \text{tr}_2 L_{22}) - E(\text{tr}_3 L_{22})] \\
 &+ 2E(\text{tr}L_{22} \text{tr}_3 L_{22}) - E(\text{tr}_4 L_{22}) ,
 \end{aligned}$$

$$\begin{aligned}
 (2.36) \quad E(\text{tr}_2 L_{22})^2 &= [(p-1)(p-2)(f_2-2)(f_2-1) / \{3! \prod_{j=1}^6 (f_1+f_2-j+2)\}] \times \\
 &[\{3(f_1-p)^2/2\} \{ (p-1)(p-2)(f_2-p-1)^2 + (p-2)(2p^2-3p+9)(f_2-p-1) \\
 &\quad + p^4-4p^3+13p^2-26p+28 \}]
 \end{aligned}$$

$$\begin{aligned}
& +\{3(f_1-p)/2\}\{2(p-1)(p-2)(f_2-p-1)^3+(p-2)(8p^2-11p+11)(f_2-p-1)^2 \\
& \quad + (p-2)(10p^3-18p^2+33p-1)(f_2-p-1) \\
& \quad + 4p^5-17p^4+40p^3-53p^2+10p+52\} \\
& +\{p(p-1)(f_2-1)f_2(f_2+p-3)(f_2+p-2)/2\} \\
& + (p-3)(f_2+p)(f_2-3)\{(p-1)(f_2-p-1)^2+(p-1)(3p-1)(f_2-p-1) \\
& \quad + 2p^3-3p^2+p+6\} ,
\end{aligned}$$

and

$$\begin{aligned}
(2.37) \quad E(\text{tr}L_{22})^4 &= \frac{3 \cdot 2^8 (p-1)(f_2-1)f_1 \left\{ \left(\sum_{i=0}^6 f_{i,p-1} d^{6-i} \right) + \left\{ (f_1-p)(f_2-p-1)/4 \right\} \left(\sum_{i=0}^5 g_{i,p-1} d^{5-i} \right) \right\}}{(f_1+f_2-4)(f_1+f_2-3)(f_1+f_2-2)(f_1+f_2-1)^4(f_1+f_2)(f_1+f_2+1)(f_1+f_2+3)(f_1+f_2+5)} \\
& + 4E(\text{tr}L_{22})^3 E(\text{tr}L_{22}) - 6E(\text{tr}L_{22})^2 [E(\text{tr}L_{22})]^2 + 3[E(\text{tr}L_{22})]^4 ,
\end{aligned}$$

where $d = (f_1+f_2-2p-1)/2$,

$f_{i,p-1}$ ($i=0, \dots, 6$) and $g_{i,p-1}$ ($i=0, \dots, 5$) can be obtained from f_i 's and g_i 's (which are polynomials in $s = p$) presented in [12, pp.7-8] by substituting in the latter $p-1$ for s .

3. Moments of $U^{(p)}$. In the previous paper by the authors [6] it has been shown that

$$(3.1) \quad 1+U^{(p)} = \{(1-\ell_{11})(1-\underline{u}'\underline{u})\}^{-1} + (1-\underline{u}'\underline{u})^{-1}(\underline{u}'\underline{M}\underline{u}) + \text{tr}\underline{M}$$

where $\underline{M} = (\underline{I}_{p-1} - \underline{I}_{22})^{-1} \underline{I}_{p-1}$. The distribution of \underline{M} is given by

$$(3.2) \quad \prod_{i=1}^{p-1} \left[\frac{\Gamma\left(\frac{f_1+f_2-i}{2}\right)}{\Gamma\left(\frac{f_1-i+1}{2}\right)\Gamma\left(\frac{f_2-i}{2}\right)} \right]^{\prod} \frac{|\underline{M}|^{-\frac{1}{4}(p-1)(p-2) \frac{1}{2}\{f_2-1-(p-1)-1\}}}{|\underline{I}_{p-1} + \underline{M}|^{\frac{1}{2}(f_1+f_2-1)}} d\underline{M}.$$

Now we have

$$(3.3) \quad \begin{aligned} E(1-l_{11})^{-1} - E(1-l_{11,0})^{-1} &= 2\lambda^2 / (f_1-2), E(1-l_{11})^{-2} - E(1-l_{11,0})^{-2} \\ &= 1 / [(f_1-2)(f_1-4)] [(2\lambda^2)^2 + 2(v-2)(2\lambda^2)], \\ E(1-l_{11})^{-3} - E(1-l_{11,0})^{-3} &= \\ &= \left[\frac{(2\lambda^2)^3 + 3(v-2)(2\lambda^2)^2 + 3(v-2)(v-4)(2\lambda^2)}{(f_1-2)(f_1-4)(f_1-6)} \right] \end{aligned}$$

and

$$\begin{aligned} E(1-l_{11})^{-4} - E(1-l_{11,0})^{-4} &= \\ &= \left[\frac{(2\lambda^2)^4 + 4(v-2)(2\lambda^2)^3 + 6(v-2)(v-4)(2\lambda^2)^2 + 4(v-2)(v-4)(v-6)(2\lambda^2)}{(f_1-2)(f_1-4)(f_1-6)(f_1-8)} \right]. \end{aligned}$$

Let $\alpha_1 = 1/(\underline{1} - \underline{u}'\underline{u})$ and $\beta_2 = \text{tr}\underline{M} + (\underline{u}'\underline{M}\underline{u}) / (\underline{1} - \underline{u}'\underline{u})$;

$$(3.4) \quad \begin{aligned} E(\alpha_1^i) &= [(f_1-2)(f_1-4) \dots (f_1-2i)] / [(f_1-p-1)(f_1-p-3) \dots \\ &\quad (f_1-p-2i+1)] \text{ for } i=1,2,3,4 ; \\ &= \eta_i^{(0)}(f_1-2)(f_1-4) \dots (f_1-2i) . \end{aligned}$$

$$(3.5) \quad E(\alpha_1 \beta_2^i) = \frac{(f_1-2)(f_1-4)\dots(f_1-2i)(f_1-p-2i)}{(f_1-p-1)(f_1-p-3)\dots(f_1-p-2i-1)} E(\text{tr}M) \\ = (f_1-2)(f_1-4)\dots(f_1-2i) \eta_i^{(1)} \quad \text{for } i=1,2,3.$$

$$(3.6) \quad E(\alpha_1 \beta_2^{i,2}) \\ = \frac{(f_1-2)(f_1-4)\dots(f_1-2i)}{(f_1-p-1)(f_1-p-3)\dots(f_1-p-2i+1)} E \left[\frac{(f_1-p-2i)(f_1-p-2i-2)}{(f_1-p-2i-1)(f_1-p-2i-3)} (\text{tr}M)^2 \right. \\ \left. - \frac{4(\text{tr}_2 M)}{(f_1-p-2i-1)(f_1-p-2i-3)} \right] \\ = (f_1-2)(f_1-4)\dots(f_1-2i) \eta_i^{(2)} \quad \text{for } i = 1,2.$$

$$(3.7) \quad E(\alpha_1 \beta_2^3) \\ = \frac{(f_1-2)}{(f_1-p-1)(f_1-p-3)(f_1-p-5)(f_1-p-7)} E \left[(f_1-p-2)(f_1-p-4)(f_1-p-6)(\text{tr}M)^3 \right. \\ \left. - 12(f_1-p-4)(\text{tr}M)(\text{tr}_2 M) + 24(\text{tr}_3 M) \right] \\ = (f_1-2) \eta_1^{(3)}.$$

Hence, we get

$$(3.8) \quad E(1+U^{(p)}) = E(1+U_0^{(p)}) + (2\lambda^2) \eta_1^{(0)}$$

$$(3.9) \quad E(1+U^{(p)})^2 = E(1+U_0^{(p)})^2 + (2\lambda^2)^2 \eta_2^{(0)} + 2(2\lambda^2) [(v-2)\eta_2^{(0)} + \eta_1^{(1)}]$$

$$(3.10) \quad E(1+U^{(p)})^3 = E(1+U_0^{(p)})^3 + (2\lambda^2)^3 \eta_3^{(0)} + 3(2\lambda^2)^2 [(v-2)\eta_3^{(0)} + \eta_2^{(1)}] \\ + 3(2\lambda^2) [(v-2)(v-4)\eta_3^{(0)} + 2(v-2)\eta_2^{(1)} + \eta_1^{(2)}],$$

and

$$(3.11) \quad E(1+U^{(p)})^4 = E(1+U_0^{(p)})^4 + (2\lambda^2)^4 \eta_4^{(0)} + 4(2\lambda^2)^3 [(v-2)\eta_4^{(0)} + \eta_3^{(1)}] \\ + 6(2\lambda^2)^2 [(v-2)(v-4)\eta_4^{(0)} + 2(v-2)\eta_3^{(1)} + \eta_2^{(2)}] \\ + 4(2\lambda^2) [(v-2)(v-4)(v-6)\eta_4^{(0)} + 3(v-2)(v-4)\eta_3^{(1)} + 3(v-2)\eta_2^{(2)} + \eta_1^{(3)}].$$

Now

$$(3.12) \quad E(\text{tr}_{i\tilde{M}}) = \binom{p-1}{i} \prod_{j=1}^i [(f_2 - j)/(f_1 - p + j - 1)], \quad (i=1, 2, \dots, p-1; \text{tr}\tilde{M} = \text{tr}_1\tilde{M})$$

$$(3.13) \quad E(\text{tr}\tilde{M})^2 = E(\text{tr}\tilde{M}) \left\{ \frac{2(f_1 + f_2 - p - 1)(f_1 - 1)}{(f_1 - p)(f_1 - p - 2)(f_1 - p + 1)} + E(\text{tr}\tilde{M}) \right\},$$

$$(3.14) \quad E(\text{tr}\tilde{M})^3 = E(\text{tr}\tilde{M}) \left\{ \frac{2^3(f_1 + f_2 - p - 1)(f_1 + 2f_2 - p - 2)(f_1 + p - 2)(f_1 - 1)}{(f_1 - p)^2(f_1 - p - 4)(f_1 - p - 2)(f_1 - p + 1)(f_1 - p + 2)} \right. \\ \left. + 3E(\text{tr}\tilde{M})^2 - 2[E(\text{tr}\tilde{M})]^2 \right\},$$

and

$$(3.15) \quad E(\text{trMtr}_2 M) = E(\text{tr}_2 M) [\{ (f_1 - p)(p-1) + 4 \} (f_2 - p - 1) \\ + 2(p+1)(p+2)] / [(f_1 - p)^2 - 4] .$$

In obtaining the expected values (2.31) - (2.33), (2.35) - (2.37) and (3.12) - (3.15), use has been made of lemmas 1 and 2 in [4] and the values of the special type of Vandermonde determinants given in [5].

Expressions for the first two moments of $U^{(p)}$ have been presented in the previous paper by the authors [6]. Now using the results (3.12) - (3.15) in (3.10) and (3.11) we get

$$(3.16) \quad E(1+U^{(p)})^3 = E(1+U_0^{(p)})^3 + A_1(2\lambda^2)^3 + 3A_2(2\lambda^2)^2 + 3A_3(2\lambda^2)$$

where

$$(3.17) \quad A_1 = \eta_3^{(0)} = [(f_1 - p - 1)(f_1 - p - 3)(f_1 - p - 5)]^{-1} ,$$

$$(3.18) \quad A_2 = (v-2)\eta_3^{(0)} + \eta_2^{(1)} ,$$

where

$$(3.19) \quad \eta_2^{(1)} = (p-1)(f_2-1)(f_1-p-4) A_1 / (f_1-p) ,$$

$$(3.20) \quad A_3 = (v-2)(v-4)\eta_3^{(0)} + 2(v-2)\eta_2^{(1)} + \eta_1^{(2)} ,$$

where

$$(3.21) \quad \eta_1^{(2)} = \frac{(p-1)(f_2-1)}{(f_1-p-3)(f_1-p+1)(f_1-p)} \left\{ (p-2)(f_2-1) + \frac{(f_2+1)(f_1-1)}{(f_1-p-2)} \right. \\ \left. + \frac{(p+1)(f_2+1)(f_1-p+1)}{(f_1-p-1)(f_1-p-2)(f_1-p-5)} \right\} .$$

Similarly

$$(3.22) \quad E(1+U^{(p)})^4 = E(1+U_0^{(p)})^4 + B_1(2\lambda^2)^4 + 4B_2(2\lambda^2)^3 + 6B_3(2\lambda^2)^2 + 4B_4(2\lambda^2) ,$$

where

$$(3.23) \quad B_1 = \eta_4^{(0)} = A_1 / (f_1 - p - 7)$$

$$(3.24) \quad B_2 = (v-2) \eta_4^{(0)} + \eta_3^{(1)}$$

where

$$\eta_3^{(1)} = (p-1)(f_2-1)(f_1-p-6) B_1 / (f_1-p)$$

$$(3.25) \quad B_3 = (v-2)(v-4)\eta_4^{(0)} + 2(v-2)\eta_3^{(1)} + \eta_2^{(2)}$$

where

$$(3.26) \quad \eta_2^{(2)} = \left\{ \frac{(f_1-p-4)(f_1-p-6)(p-1)(f_2-1)}{(f_1-p)^2} \left[\frac{2(f_1-1)(f_1+f_2-p-1)}{(f_1-p+1)(f_1-p-2)} \right. \right. \\ \left. \left. + (p-1)(f_2-1) \right] - 2(p-1)(p-2)(f_2-1)(f_2-2) / \{(f_1-p)(f_1-p+1)\} \right\} B_1$$

$$(3.27) \quad B_4 = (v-2)(v-4)(v-6)\eta_4^{(0)} + 3(v-2)(v-4)\eta_3^{(1)} + 3(v-2)\eta_2^{(2)} + \eta_1^{(3)}$$

where

$$(3.28) \quad \eta_1^{(3)} = \left\{ \frac{(f_1-p-2)(f_1-p-4)(f_1-p-6)(p-1)(f_2-1)}{(f_1-p)^3} \right. \\ \left. \left[\frac{2^3(f_1-1)(f_1+f_2-p-1)(f_1+2f_2-p-2)(f_1+p-2)}{(f_1-p-2)(f_1-p-4)(f_1-p+1)(f_1-p+2)} \right. \right. \\ \left. \left. + \frac{6(f_2-1)(f_1+f_2-p-1)(p-1)(f_1-1)}{(f_1-p-2)(f_1-p+1)} + (p-1)^2(f_2-1)^2 \right] \right. \\ \left. - \frac{6(f_1-p-4)(p-1)(p-2)(f_2-1)(f_2-2)}{(f_1-p-2)(f_1-p)(f_1-p+1)(f_1-p+2)} \left[\{(f_1-p)(p-1)+4\}(f_2-p-1)+2(p+1)(p+2) \right] \right. \\ \left. + 4(p-1)(p-2)(p-3)(f_2-3)(f_2-2)(f_2-1) / \{(f_1-p)(f_1-p+1)(f_1-p+2)\} \right\} \cdot B_1$$

4. Approximations to the Distribution of $U^{(p)}$. Pillai [13] has given an approximation to the distribution of $U^{(2)}$ for $f_1 > f_2$ and which is good even for very small values of f_2 . The following approximation to the distribution of $U^{(p)}$ for $f_1 > (p-1)f_2$, based on its moments discussed in the preceding section and [6], generalizes Pillai's results for $U^{(2)}$ [13].

$$(4.1) \quad g(U^{(p)}) = (U^{(p)})^{p_1-1} / (1+U^{(p)}/k)^{p_1+q_1+1} k^{p_1} \beta(p_1, q_1+1), \quad 0 < U^{(p)} < \infty,$$

where

$$p_1 = 2q_1 / \{q_1(h-1) - 2h\},$$

$$q_1 = 2\{c^2(f_1-p-3)h - (c+d)^2(f_1-p-1)\} / \{c^2(f_1-p-3)(h+1) - 2(c+d)^2(f_1-p-1)\},$$

$$k = c\{q_1(h-1) - 2h\} / 2(f_1-p-1),$$

$$h = (c+1.99d)^3(f_1-p-1) / \{(c+d)^2(f_1-p-5)c\},$$

$$c = pf_2 + 2\lambda^2 \quad \text{and} \quad d = (f_1 + (1-p)f_2 - 1) / (f_1 - p)$$

A comparison of the moments from (4.1) with the **respective** exact ones may be made from Table 1. However, the moments for the case $p = 2$ are **not** presented in Table 1 since comparisons of the approximate and exact moments for $U^{(2)}$ have already been made [13]. It may further be pointed out that the case $p = 1$ is that of the non-central F [8]. Hence the accuracy of the approximation may be compared in this case with the approximation to the distribution of non-central F obtained by Patnaik and the exact distribution using Table 7 of [8]. However, it should be

Table 1

Moments (central) of $U^{(p)}$ from the exact and approximate λ distributions for different values of $f_1 > (p-1)f_2$, λ^2 and p [Eq. (4.1)]

Moments	$f_1=10$ $f_2=3$ $\lambda^2=0.5$ $p=1$			$f_1=10$ $f_2=15$ $\lambda^2=2$ $p=1$		
	Exact	Approximate	Ratio(A/E)	Exact	Approximate	Ratio (A/E)
μ_1^1	0.5000	0.5000	1.0000	2.3750	2.3750	1.0000
μ_2	0.2917	0.2708	0.9286	2.8335	2.6927	0.9486
μ_3	0.6458	0.5881	0.9107	16.8841	15.5835	0.9230
μ_4	5.0312	4.7122	0.9366	373.7200	336.9354	0.9016
$\sqrt{\mu_2}$	0.5401	0.5204	0.9636	1.6848	1.6409	0.9740
β_1	16.8105	17.4119	1.0358	12.4644	12.4383	0.9979
β_2	59.1429	64.2422	1.0862	46.3827	46.4695	1.0019
Moments	$f_1=10$ $f_2=15$ $\lambda^2=32$ $p=1$			$f_1=10$ $f_2=50$ $\lambda^2=2$ $p=1$		
	Exact	Approximate	Ratio(A/E)	Exact	Approximate	Ratio (A/E)
μ_1^1	9.8750	9.8750	1.0000	6.7500	6.7500	1.0000
μ_2	38.4635	35.8177	0.9312	17.6042	17.4583	0.9917
μ_3	827.1185	742.8304	0.8980	256.5521	253.0986	0.9865
μ_4	66672.3710	57685.2030	0.8652	14005.6130	13732.1110	0.9805
$\sqrt{\mu_2}$	6.2019	5.9848	0.9650	4.1957	4.1783	0.9958
β_1	12.0223	12.0084	0.9988	12.0644	12.0385	0.9979
β_2	45.0658	44.9644	0.9977	45.1930	45.0538	0.9969
Moments	$f_1=10$ $f_2=3$ $\lambda^2=32$ $p=1$			$f_1=10$ $f_2=100$ $\lambda^2=2$ $p=1$		
	Exact	Approximate	Ratio(A/E)	Exact	Approximate	Ratio (A/E)
μ_1^1	8.3750	8.3750	1.0000	13.0000	13.0000	1.0000
μ_2	28.8385	26.1927	0.9083	60.8333	60.6875	0.9976
μ_3	536.8841	464.7210	0.8656	1644.8333	1637.6403	0.9956
μ_4	37469.2820	30867.1300	0.8238	166726.9900	165521.0700	0.9928
$\sqrt{\mu_2}$	5.3702	5.1179	0.9530	7.7996	7.7902	0.9988
β_1	12.0182	12.0183	1.0000	12.0176	11.9988	0.9984
β_2	45.0535	44.9920	0.9986	45.0529	44.9422	0.9975

Table 1 (Cont'd)

Moments	$f_1=12$	$f_2=5$	$\lambda^2=0.5$	$p=3$	$f_1=12$	$f_2=5$	$\lambda^2=32$	$p=3$
	Exact	Approximate	Ratio (A/E)		Exact	Approximate	Ratio (A/E)	
μ_1	2.0000	2.0000	1.0000		9.8750	9.8750	1.0000	
μ_2	1.4120	1.4077	0.9969		33.1672	32.8712	0.9911	
μ_3	4.7556	5.7841	1.2163		649.5157	652.7574	1.0050	
μ_4	63.1979	89.0452	1.4090		48253.3920	48599.6620	1.0072	
$\sqrt{\mu_2}$	1.1883	1.1865	0.9985		5.7591	5.7333	0.9955	
β_1	8.0328	11.9941	1.4932		11.5625	11.9966	1.0375	
β_2	31.6964	44.9379	1.4178		43.8641	44.9782	1.0254	
Moments	$f_1 = 15$	$f_2 = 4$	$\lambda^2 = 4.5$	$p=4$	$f_1 = 15$	$f_2 = 4$	$\lambda^2 = 32$	$p=4$
	Exact	Approximate	Ratio (A/E)		Exact	Approximate	Ratio (A/E)	
μ_1	2.5000	2.5000	1.0000		8.0000	8.0000	1.0000	
μ_2	1.7170	1.6765	0.9764		16.6545	16.3640	0.9826	
μ_3	5.1341	5.7869	1.1271		177.0424	176.4678	0.9968	
μ_4	51.6730	61.7517	1.1950		5924.6806	5886.3824	0.9935	
$\sqrt{\mu_2}$	1.3104	1.2948	0.9881		4.0810	4.0453	0.9912	
β_1	5.2069	7.1062	1.3648		6.7851	7.1065	1.0474	
β_2	17.5267	21.9693	1.2534		21.3599	21.9820	1.0291	
Moments	$f_1=25$	$f_2=5$	$\lambda^2=12.5$	$p=5$	$f_1=25$	$f_2=5$	$\lambda^2=32$	$p=5$
	Exact	Approximate	Ratio (A/E)		Exact	Approximate	Ratio (A/E)	
μ_1	2.6316	2.6316	1.0000		4.6842	4.6842	1.0000	
μ_2	0.9076	0.8768	0.9660		2.7709	2.6917	0.9714	
μ_3	1.1645	1.2747	1.0947		6.7802	6.8592	1.0116	
μ_4	5.8355	6.2180	1.0656		59.1995	58.6460	0.9907	
$\sqrt{\mu_2}$	0.9527	0.9364	0.9829		1.6646	1.6406	0.9856	
β_1	1.8137	2.4107	1.3292		2.1609	2.4123	1.1164	
β_2	7.0840	8.0887	1.1418		7.7106	8.0942	1.0498	

pointed out that the approximation to the distribution of $U^{(p)}_{\wedge}$ in (4.1) has been suggested in this paper using the first three moments and with considerations of accuracy for $p > 1$. This may be noted from Table 3 where, in general, the respective exact and approximate moments seem to be closer as p increases (from 1 to 4). Table 2 gives some idea of the accuracy of the approximation when $p = 1$.

Table 2

Values of $\int_0^{U_0} g(U^{(1)}) dU^{(1)}$ from approximate and exact distributions

f_1	f_2	λ^2	U_0	Probability		
				Approximate Eqn. (4.1)	Patnaik	Exact
10	3	2	1.1124	.765	.752	.745
10	3	8	1.1124	.154	.203	.206
10	3	8	1.9656	.503	.520	.517
10	5	3	1.663	.738	.731	.731
10	5	3	2.818	.920	.913	.914
20	3	2	0.4647	.708	.706	.700
20	5	3	0.67775	.671	.665	.664
20	5	12	1.02575	.196	.244	.245

It may be observed that the approximation suggested for $U^{(1)}$ is more accurate at the upper tail end than the lower. Now, in view of the fact that the moments are closer for larger values of p in Table 3, it is reasonable to assume that the approximation to the distribution of $U^{(p)}$

Table 3

Moments (central) of $U^{(p)}$ from the exact and approximate distributions [EQ. (4.1)]
for different values of $f_1 > (p-1) f_2$, λ^2 and p .

Moments	$f_1=10$	$f_2=3$	$\lambda^2=2$	$p=1$	$f_1=10$	$f_2=3$	$\lambda^2=2$	$p=2$
	Exact	Approximate	Ratio (A/E)		Exact	Approximate	Ratio (A/E)	
μ_1^2	0.8750	0.8750	1.0000		1.4286	1.4286	1.0000	
μ_2	0.7135	0.5677	0.7956		1.4163	1.2610	0.8903	
μ_3	2.2747	1.6241	0.7140		6.8812	6.1366	0.8918	
μ_4	26.1770	17.3107	0.6613		179.7042	160.3333	0.8922	
$\sqrt{\mu_2}$	0.8447	0.7535	0.8920		1.1901	1.1229	0.9436	
β_1	14.2431	14.4157	1.0121		16.6666	18.7820	1.1269	
β_2	51.4140	53.7110	1.0447		89.5842	100.8362	1.1256	
Moments	$f_1=50$	$f_2=10$	$\lambda^2=4.5$	$p=1$	$f_1=50$	$f_2=10$	$\lambda^2=4.5$	$p=2$
	Exact	Approximate	Ratio (A/E)		Exact	Approximate	Ratio (A/E)	
μ_1^2	0.3958	0.3958	1.0000		0.6170	0.6170	1.0000	
μ_2	0.0322	0.0245	0.7607		0.0461	0.0395	0.8568	
μ_3	0.0 ² 628	0.0 ² 410	0.6539		0.0 ² 878	0.0 ² 752	0.8570	
μ_4	0.0 ² 529	0.0 ² 306	0.5787		0.0 ² 947	0.0 ² 738	0.7802	
$\sqrt{\mu_2}$	0.1794	0.1564	0.8722		0.2147	0.1988	0.9256	
β_1	1.1822	1.1484	0.9714		0.7857	0.9174	1.1676	
β_2	5.1121	5.1126	1.0000		4.4507	4.7297	1.0627	
Moments	$f_1=50$	$f_2=10$	$\lambda^2=4.5$	$p=3$	$f_1=50$	$f_2=10$	$\lambda^2=4.5$	$p=4$
	Exact	Approximate	Ratio (A/E)		Exact	Approximate	Ratio (A/E)	
μ_1^2	0.8478	0.8478	1.0000		1.0889	1.0889	1.0000	
μ_2	0.0619	0.0566	0.9144		0.0799	0.0762	0.9530	
μ_3	0.0119	0.0123	1.0348		0.0158	0.0190	1.1999	
μ_4	0.0159	0.0148	0.9313		0.0253	0.0266	1.0491	
$\sqrt{\mu_2}$	0.2489	0.2380	0.9563		0.2827	0.2760	0.9762	
β_1	0.5969	0.8358	1.4003		0.4904	0.8156	1.6633	
β_2	4.1326	4.6027	1.1138		3.9660	4.5811	1.1551	

provides at least a two decimal accuracy at the upper tail end. However, it should be remembered that the approximation is valid only for $f_1 > (p-1) f_2$ which for $p = 1$ reduces to $f_1 > 0$. In the latter case, as is evident from Table 1, in many cases, the moments are closer when $f_2 > f_1$.

Again a comparison of the probabilities in Table 2 arouses the natural curiosity to attempt a generalization of Patnaik's approximation [8]. The following is such a generalization equating the first two respective moments of the exact and approximate distributions:

$$(4.2) \quad \varepsilon_1(U^{(p)}) = (U^{(p)})^{\frac{1}{2}v_1-1} / (1+U^{(p)}/k_1)^{\frac{1}{2}(v_1+v_2)} k_1^{\frac{1}{2}v_1} \beta(\frac{1}{2}v_1, \frac{1}{2}v_2) ,$$

$$0 < U^{(p)} < \infty$$

where

$$k_1 = (pf_2 + 2\lambda^2) / v_1 ,$$

$$v_1 = (pf_2 + 2\lambda^2)^2 (f_1 - p) / (4\lambda^2 + pf_2) \{f_1 + f_2(1-p) - 1\} ,$$

and $v_2 = f_1 - p + 1$.

The two approximations (4.1) and (4.2) may be compared for the same set of values of f_1 , f_2 , λ^2 and p as in Tables 1 and 3 using the ratios of approximate to exact moments given in Table 4 in the two cases. It is obvious from Table 4 that while for $p=1$ the ratios from Eq. (4.2) are closer to unity than the respective ratios from Eq. (4.1), as p increases the latter become closer to unity than the former (except of course for μ_2).

Table 4

Ratios of moments (Central) of $U^{(p)}$ from the exact and approximate distributions [(4.1) and (4.2)] for sets of values of f_1, f_2, λ^2 and p as in

Tables 1 and 3

Moments	$f_1=10$	$f_2=3$	$f_1=10$	$f_2=15$	$f_1=10$	$f_2=15$
	$\lambda^2=0.5$	$p=1$	$\lambda^2=2$	$p=1$	$\lambda^2=32$	$p=1$
	Ratio (A/E)		Ratio (A/E)		Ratio (A/E)	
	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)
μ_1^1	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
μ_2	0.9286	1.0000	0.9486	1.0000	0.9312	1.0000
μ_3	0.9107	1.0161	0.9230	1.0021	0.8980	1.0026
μ_4	0.9366	1.0326	0.9016	1.0036	0.8652	1.0042
$\sqrt{\mu_2}$	0.9636	1.0000	0.9740	1.0000	0.9650	1.0000
β_1	1.0358	1.0325	0.9979	1.0042	0.9988	1.0052
β_2	1.0862	1.0326	1.0019	1.0036	0.9977	1.0042
Moments	$f_1=10$	$f_2=50$	$f_1=10$	$f_2=3$	$f_1=10$	$f_2=100$
	$\lambda^2=2$	$p=1$	$\lambda^2=32$	$p=1$	$\lambda^2=2$	$p=1$
	Ratio (A/E)		Ratio (A/E)		Ratio (A/E)	
	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)
μ_1^1	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
μ_2	0.9917	1.0000	0.9083	1.0000	0.9976	1.0000
μ_3	0.9865	1.0000	0.8656	1.0047	0.9956	1.0000
μ_4	0.9805	1.0001	0.8238	1.0076	0.9928	1.0000
$\sqrt{\mu_2}$	0.9958	1.0000	0.9530	1.0000	0.9988	1.0000
β_1	0.9979	1.0001	1.0000	1.0095	0.9984	1.0000
β_2	0.9969	1.0001	0.9986	1.0076	0.9975	1.0000
Moments	$f_1=12$	$f_2=5$	$f_1=12$	$f_2=5$	$f_1=15$	$f_2=4$
	$\lambda^2=0.5$	$p=3$	$\lambda^2=32$	$p=3$	$\lambda^2=4.5$	$p=4$
	Ratio (A/E)		Ratio (A/E)		Ratio (A/E)	
	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)
μ_1^1	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
μ_2	0.9969	1.0000	0.9911	1.0000	0.9764	1.0000
μ_3	1.2163	1.2227	1.0050	1.0188	1.1271	1.1699
μ_4	1.4090	1.4207	1.0072	1.0260	1.1950	1.2573
$\sqrt{\mu_2}$	0.9985	1.0000	0.9955	1.0000	0.9881	1.0000
β_1	1.4932	1.4951	1.0375	1.0380	1.3648	1.3687
β_2	1.4178	1.4207	1.0254	1.0260	1.2534	1.2573

Table 4 (Cont'd.)

Moments	$f_1=15$	$f_2=4$	$f_1=25$	$f_2=5$	$f_1=25$	$f_2=5$
	$\lambda^2=32$	$p=4$	$\lambda^2=12.5$	$p=5$	$\lambda^2=32$	$p=5$
	Ratio (A/E)		Ratio (A/E)		Ratio (A/E)	
	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)
μ_1^2	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
μ_2	0.9826	1.0000	0.9660	1.0000	0.9714	1.0000
μ_3	0.9968	1.0239	1.0947	1.1563	1.0116	1.0584
μ_4	0.9935	1.0303	1.0656	1.1462	0.9907	1.0522
$\sqrt{\mu_2}$	0.9912	1.0000	0.9829	1.0000	0.9856	1.0000
β_1	1.0474	1.0485	1.3292	1.3370	1.1164	1.1203
β_2	1.0291	1.0303	1.1418	1.1462	1.0498	1.0522
Moments	$f_1=10$	$f_2=3$	$f_1=10$	$f_2=3$	$f_1=50$	$f_2=10$
	$\lambda^2=2$	$p=1$	$\lambda^2=2$	$p=2$	$\lambda^2=4.5$	$p=1$
	Ratio (A/E)		Ratio (A/E)		Ratio (A/E)	
	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)
μ_1^2	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
μ_2	0.7956	1.0000	0.8903	1.0000	0.7607	1.0000
μ_3	0.7140	1.0419	0.8918	1.0722	0.6539	1.0559
μ_4	0.6613	1.0772	0.8922	1.1300	0.5787	1.0479
$\sqrt{\mu_2}$	0.8920	1.0000	0.9436	1.0000	0.8722	1.0000
β_1	1.0121	1.0855	1.1269	1.1497	0.9714	1.1150
β_2	1.0447	1.0772	1.1256	1.1300	1.0000	1.0479
Moments	$f_1=50$	$f_2=10$	$f_1=50$	$f_2=10$	$f_1=50$	$f_2=10$
	$\lambda^2=4.5$	$p=2$	$\lambda^2=4.5$	$p=3$	$\lambda^2=4.5$	$p=4$
	Ratio (A/E)		Ratio (A/E)		Ratio (A/E)	
	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)	Eq.(4.1)	Eq.(4.2)
μ_1^2	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
μ_2	0.8568	1.0000	0.9144	1.0000	0.9530	1.0000
μ_3	0.8570	1.1233	1.0348	1.2161	1.1999	1.3143
μ_4	0.7802	1.0903	0.9313	1.1349	1.0491	1.1708
$\sqrt{\mu_2}$	0.9256	1.0000	0.9563	1.0000	0.9762	1.0000
β_1	1.1676	1.2619	1.4003	1.4788	1.6633	1.7274
β_2	1.0627	1.0902	1.1138	1.1349	1.1551	1.1708

However β_1 and β_2 are generally better in the case of Eq. (4.1) even for $p=1$. From the foregone analysis it may be concluded that approximation (4.2) is useful for smaller values of p while approximation (4.1) might be more accurate for comparatively larger values of p . Further it should be pointed out that the condition $f_1 > (p-1)f_2$ applies for both approximations.

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