

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

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1. Introduction and Summary. Let A_1 and A_2 be two positive definite matrices of order p , A_1 having a Wishart distribution [2, 12] with f_1 degrees of freedom and A_2 (pseudo) non-central (linear) Wishart distribution [1,3,4,12,13] with f_2 degrees of freedom. Now transform

$$A_2 = C Y Y' C'$$

where C is a lower triangular matrix such that

$$A_1 + A_2 = C C'$$

and the density function of Y : $p \times f_2$ is given by

$$(1.1) \quad k_1 e^{-\lambda^2} \sum_{j=0}^{\infty} (2\lambda y_{11})^j \Gamma\left[\frac{1}{2}(f_1 + f_2 + j)\right] |I_p - YY'|^{-\frac{1}{2}(f_1 - p - 1)} / j!$$

where I_p is an identity matrix of order p ,

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$$k_1 = \prod_{i=2}^p \Gamma[\frac{1}{2}(f_1+f_2-i+1)] / \prod_{i=1}^p \Gamma[(f_1-i+1)/2],$$

λ is the only non-centrality parameter in the linear case and y_{11} is the element in the top left corner of the \underline{Y} matrix.

Now $V^{(s)}$ criterion suggested by Pillai and $U^{(s)}$ (a constant times Hotelling's T_0^2), [7,8,9,10] are the sums of the non-zero characteristic roots of the matrix $\underline{Y}\underline{Y}'$ and $(\underline{I}_p - \underline{Y}\underline{Y}')^{-1} - \underline{I}_p$ respectively. Here s is minimum (f_2, p) . Also we may note that $V^{(s)} = \text{trace } \underline{Y}\underline{Y}' = \text{trace } \underline{Y}'\underline{Y}$ and $U^{(s)} = \text{tr}(\underline{I}_p - \underline{Y}\underline{Y}')^{-1} - p = \text{tr}(\underline{I}_{f_2} - \underline{Y}'\underline{Y})^{-1} - f_2$. It can be shown that the density function of $\underline{Y}'\underline{Y}$ for $f_2 \leq p$ can be obtained from the density function of $\underline{Y}\underline{Y}'$ for $f_2 \geq p$ if in the latter case the following changes are made: [12,5]

$$(1.2) \quad (f_1, f_2, p) \rightarrow (f_1 - f_2 + p, p, f_2).$$

Hence, for the criterion $V^{(s)}$, (and similarly for $U^{(s)}$), we shall only consider the density function of $\underline{L} = \underline{Y}\underline{Y}'$ for $f_2 \geq p$ which is given by [6]

$$(1.3) \quad f(\underline{L}) = k e^{-\lambda^2} {}_1F_1\left\{\frac{1}{2}(f_1+f_2), \frac{1}{2}f_2, \lambda^2 \ell_{11}\right\} |\underline{L}|^{(f_2-p-1)/2} |\underline{I}_p - \underline{L}|^{(f_1-p-1)/2}$$

where

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

ℓ_{11} is the element in the top left corner of the matrix \underline{L} and ${}_1F_1$ denotes the confluent hypergeometric function. We shall call the distribution

of \underline{L} : $p \times p$ the non-central (linear) multivariate beta distribution with f_2 and f_1 degrees of freedom.

Pillai [11] had noted that the elements of the matrix \underline{L} can be transformed into independent beta variables which he showed for $p = 2, 3, 4$ and 5 . In this paper we give a theorem which proves the general case. In addition, when $\lambda = 0$ the first and second order moments of l_{ij} are obtained and used to derive the first two moments of $V^{(s)}$ in the non-central case when $f_2 \geq p$. The moments of $V^{(s)}$ for $f_2 \leq p$ can be written down with the help of (1.2). Similar results are obtained for $U^{(p)} = t_r(\underline{I}_p - \underline{L})^{-1-p}$.

2. Independent Beta Variables. Let

$$\underline{L} = \begin{pmatrix} l_{11} & \underline{l}' \\ \underline{l} & \underline{L}_{11} \end{pmatrix} \begin{matrix} 1 \\ p-1, \underline{L}_{22} = \underline{L}_{11}^{-1} \underline{l} \underline{l}' \underline{l}_{11} \\ 1 \quad p-1 \end{matrix}$$

and we note that

$$|\underline{L}| = l_{11} |\underline{L}_{22}|$$

and

$$|\underline{I}_p - \underline{L}| = (1 - l_{11}) |\underline{I}_{p-1} - \underline{L}_{22} - \underline{l} \underline{l}' / [l_{11}(1 - l_{11})]|.$$

Then it is easy to show that

$$l_{11} \text{ and } (\underline{L}_{22}, \underline{v} = \underline{l} / \sqrt{l_{11}(1 - l_{11})})$$

are independently distributed and their respective distributions are

$$(2.1) \quad f_1(\underline{l}_{11}) = [\beta(\frac{1}{2}f_2, \frac{1}{2}f_1)]^{-1} \exp(-\lambda^2) \ell_{11}^{\frac{1}{2}f_2-1} \ell_{11}^{\frac{1}{2}f_1-1} \Gamma[\frac{1}{2}(f_1+f_2), \frac{1}{2}f_2, \lambda^2 \ell_{11}]$$

and

$$(2.2) \quad f_2(\underline{L}_{22}, \underline{v}) = k_2 |\underline{L}_{22}|^{\frac{1}{2}[(f_2-1)-(p-1)-1]} |\underline{I}_{p-1} - \underline{L}_{22}^{-1} \underline{v} \underline{v}'|^{\frac{1}{2}(f_1-p-1)}$$

where

$$k_2 = k \beta(\frac{1}{2}f_2, \frac{1}{2}f_1).$$

For further independence, we can use two types of transformations given by

$$(2.3) \quad \underline{u} = (\underline{I}_{p-1} - \underline{L}_{22})^{\frac{1}{2}} \underline{v} \quad \text{or} \quad \underline{w} = \underline{T}^{-1} \underline{v}$$

where $\underline{I}_{p-1} - \underline{L}_{22} = \underline{T} \underline{T}'$ and $\underline{T} : (p-1) \times (p-1)$ is a lower triangular matrix. It is easy to show that \underline{u} (or \underline{w}) and \underline{L}_{22} are independently distributed and their respective distributions are

$$(2.4) \quad f_3(\underline{u}) = \pi^{-\frac{1}{2}(p-1)} \frac{\Gamma(\frac{1}{2}f_1)}{\Gamma(\frac{1}{2}(f_1-p+1))} (1 - \underline{u}' \underline{u})^{\frac{1}{2}(f_1-p-1)} \quad [\text{or } f_3(\underline{w})]$$

and

$$(2.5) \quad f_4(\underline{L}_{22}) = k_3 |\underline{L}_{22}|^{\frac{1}{2}[(f_2-1)-(p-1)-1]} |I-\underline{L}_{22}|^{\frac{1}{2}[f_1-(p-1)-1]}$$

where $k_3 = \Pi^{\frac{1}{2}(p-1)} k_2$. We may note that the distribution of $\underline{L}_{22}:(p-1) \times (p-1)$ is central multivariate beta distribution with (f_2-1) and f_1 degrees of freedom, and the similar reduction from \underline{L}_{22} can be carried successively. We may also note that the transformation

$$(2.6) \quad x_i = u_i^2 / (1 - u_1^2 - \dots - u_{i-1}^2), \quad i = 1, 2, \dots, p-1$$

in (2.4) gives us the independent beta-variates and their density functions are given by

$$(2.7) \quad g_i(x_i) = \{\beta[\frac{1}{2}, \frac{1}{2}(f_1-i)]\}^{-1} x_i^{\frac{1}{2}-1} (1-x_i)^{\frac{1}{2}(f_1-i)-1}$$

From the foregone, we have the following theorem:

Theorem I: If the distribution of $\underline{L} = \begin{pmatrix} \ell_{11} & \ell' \\ \ell & \underline{L}_{11} \end{pmatrix}$ is given by (1.3), then

$$\ell_{11}, \underline{L}_{22} = \underline{L}_{11} - \frac{\ell \ell'}{\ell_{11}} \quad \text{and} \quad \underline{u} = (\underline{I}_{p-1} - \underline{L}_{22})^{\frac{1}{2}} \frac{\ell}{\sqrt{\ell_{11}(1-\ell_{11})}}$$

[or $\underline{w} = \underline{T}^{-1} \frac{\ell}{\sqrt{\ell_{11}(1-\ell_{11})}}$ where $\underline{T} \underline{T}' = \underline{I}_{p-1} - \underline{L}_{22}$ and \underline{T} is a lower triangular matrix] are independently distributed and their respective distributions are defined in (2.1), (2.5) and (2.4).

It can be verified for $p=3$ that from the variates ℓ_{11}, \underline{w} and \underline{L}_{22} , we can obtain the independent

β -variates exactly the same as given by Pillai [11], but the use of l_{11} , u and L_{22} will give independent β -variates different from those of Pillai [11] in spite of the identical β -distributions.

3. The first and second order moments of l_{ij} when $\lambda=0$. Let the density function of L be given by

$$(3.1) \quad k |L|^{-\frac{1}{2}(f_2-p-1)} |L_p - L|^{-\frac{1}{2}(f_1-p-1)},$$

where k is the same as in (1.3). It is easy to see that

$$(3.2) \quad \begin{aligned} E(l_{ij}) &= E(l_{11}) \quad \text{when } i = j \\ &= E(l_{12}) \quad \text{when } i \neq j \end{aligned}$$

and

$$(3.3) \quad \begin{aligned} E(l_{ij} l_{i'j'}) &= E(l_{11}^2) && \text{when } i=j=i'=j' \\ &= E(l_{11} l_{12}) && \text{when } i=j=i', i \neq j' \\ &= E(l_{11} l_{22}) && \text{when } i=j, i' \neq i, i'=j' \\ &= E(l_{12}^2) && \text{when } i=i', j=j', i \neq i' \\ &= E(l_{11} l_{23}) && \text{when } i=j, i' \neq j' \neq i \\ &= E(l_{12} l_{13}) && \text{when } i=i', j \neq j' \neq i' \\ &= E(l_{12} l_{34}) && \text{when } i \neq j \neq i' \neq j' \end{aligned}$$

It is easy to see that if $\nu = f_1 + f_2$,

$$E(l_{11}) = f_2/\nu, \quad E(l_{12}) = 0,$$

and

$$E(l_{11}^2) = : f_2(f_2+2)/\nu(\nu+2).$$

For $E(l_{12}^2)$, we integrate over other variates except l_{11}, l_{12} and l_{22} . Then as in theorem I, $u_1 = l_{12}/\sqrt{(1-l_{11})(1-z)l_{11}}$, l_{11} and $(l_{22}-l_{12}^2/l_{11}) = z$ are independently distributed. Hence

$$\begin{aligned} E(l_{12}^2) &= E[(1-l_{11})l_{11}], \quad E(1-z) \quad E(u_1^2 = x_1) \\ &= f_1 f_2 / \{\nu(\nu-1)(\nu+2)\}, \end{aligned}$$

$$E(l_{11}l_{12}) = E[l_{11} \sqrt{l_{11}(1-z)(1-l_{11})}] \quad E(u_1) = 0,$$

and

$$\begin{aligned} E(l_{11}l_{22}) &= E(l_{11} z) + E\{l_{11}(1-l_{11})(1-z)x_1\} \\ &= \{f_2(f_2-1) + f_1 f_2 / (\nu+2)\} / \nu(\nu-1). \end{aligned}$$

Similarly for obtaining $E(l_{11}l_{23})$ and $E(l_{12}l_{13})$, we consider (3.1) with $p=3$ only. Using the successive reduction of theorem I, it can be shown that

$$E(l_{11}l_{23}) = E(l_{12}l_{13}) = 0.$$

The same type of reduction gives us after some algebra

$$E(l_{12}l_{34}) = 0.$$

Hence, we have the following theorem:

Theorem II: Let the distribution of \underline{L} : $p \times p$ be given by (3.1). Then

$$(3.4) \quad E(l_{ij}) = \begin{cases} f_2/v & \text{if } i = j \\ 0 & \text{otherwise,} \end{cases}$$

and

$$(3.5) \quad E(l_{ij}l_{i'j'}) = \begin{cases} f_2(f_2+2)/\{v(v+2)\} & \text{if } i=j=i'=j' \\ f_1f_2/\{v(v-1)(v+2)\} & \text{if } i=i', j=j', i \neq j \\ \text{and } i=j', i'=j, i \neq j \\ f_2\{(f_2-1)+f_1/(v+2)\}/\{v(v-1)\} & \text{if } i=j, i \neq j', i \neq i' \\ 0 & \text{otherwise.} \end{cases}$$

4. First two moments of $V^{(s)}$ criterion. We note that

$$(4.1) \quad V^{(s)} = t_r L = l_{11} + t_r L_{22} + (1-l_{11})\underline{u}'(I_{p-1} - L_{22})\underline{u},$$

where l_{11} , \underline{u} and L_{22} are independently distributed and their respective distributions are given by (2.1), (2.4) and (2.5). With the help of Theorem II, we find that

$$(4.2) \quad E(\underline{I}_{p-1} - \underline{L}_{22}) = \underline{I}_{p-1} \{f_1 / (v-1)\},$$

$$(4.3) \quad E[(t_r \underline{L}_{22})(\underline{I}_{p-1} - \underline{L}_{22})] = \delta_1 \underline{I}_{p-1}$$

and

$$(4.4) \quad E(t_r \underline{L}_{22})^2 = \frac{(p-1)(f_2-1)}{v-1} \left\{ \frac{f_2+1}{v+1} + \frac{(p-2)(f_2-2)}{v-2} + \frac{f_1(p-2)}{(v+1)(v-2)} \right\},$$

where

$$(4.5) \quad \delta_1 = \frac{(f_2-1)}{(v-1)} \left\{ (p-1) - \frac{f_2+1}{v+1} \frac{(f_2-2)(p-2)}{v-2} - \frac{f_1(p-2)}{(v+1)(v-2)} \right\}.$$

Moreover,

$$(4.6) \quad E[\underline{u}'(\underline{I}_{p-1} - \underline{L}_{22})\underline{u}] = \{f_1 / (v-1)\} E(\underline{u}' \underline{u}) = (p-1) / (v-1),$$

$$(4.7) \quad E[(t_r \underline{L}_{22}) \underline{u}'(\underline{I}_{p-1} - \underline{L}_{22})\underline{u}] = \delta_1 E(\underline{u}' \underline{u}) = \delta_1 (p-1) / f_1,$$

and

$$(4.8) \quad E\{\underline{u}'(\underline{I}_{p-1} - \underline{L}_{22})\underline{u}\}^2 = E\{\underline{u}' \underline{S} \underline{u}\}^2 \quad \text{if } \underline{S} = \underline{I}_{p-1} - \underline{L}_{22}$$

$$= E(s_{11}^2) \sum_{i=1}^{p-1} E(v_i^4) + \{E(s_{11}s_{22}) + 2E(s_{12}^2)\} \sum_{i \neq j=1}^{p-1} E(v_i v_j)$$

$$= \frac{3(p-1)}{(v-1)(v+1)} + \frac{(p-1)(p-2)}{(v-1)(v-2)(f_1+2)} \left\{ (f_1-1) + \frac{3(f_2-1)}{v+1} \right\}.$$

Hence, we get

$$(4.9) \quad E(V^{(s)}) = 1 + \frac{(p-1)(f_2-1)}{v-1} + f_1 \left\{ \frac{p-1}{v-1} - 1 \right\} a_1$$

and

$$(4.10) \quad E(V^{(s)})^2 = 1 + \frac{(p-1)(f_2-1)}{v-1} \left\{ 2 + \frac{f_2+1}{v+1} + \frac{(p-2)(f_2-2)}{v+2} + \frac{f_1(p-2)}{(v+1)(v-2)} \right\}$$

$$- 2 \left[f_1 \left\{ 1 - \frac{p-1}{v-1} + \frac{(p-1)(f_2-1)}{v-1} \right\} \right.$$

$$+ \frac{(p-1)(f_2-1)}{v-1} \left\{ 1-p + \frac{f_2+1}{v+1} + \frac{(f_2-2)(p-2)}{v-2} + \frac{f_1(p-2)}{(v+1)(v-2)} \right\} \left. \right] a_1$$

$$+ f_1(f_1+2) \left[1 - \frac{2(p-1)}{v-1} + \frac{3(p-1)}{(v-1)(v+1)} + \right.$$

$$\left. + \frac{(p-1)(p-2)}{(v-1)(v-2)(f_1+2)} \left\{ f_1-1 + \frac{3(f_2-1)}{v+1} \right\} \right] a_2,$$

where

$$(4.11) \quad a_1 = \left\{ \sum_{i=0}^{\infty} (\lambda^2)^i / [i!(v+2i)] \right\} \exp(-\lambda^2),$$

and

$$(4.12) \quad a_2 = \left\{ \sum_{i=0}^{\infty} (\lambda^2)^i / [i!(v+2i)(v+2i+2)] \right\} \exp(-\lambda^2).$$

The expressions for the moments of $V^{(s)}$ given by (4.9) and (4.10) reduce to the results for $s=2$ given by Pillai [11] when $p=2$. However, Pillai has provided the first four moments of $V^{(2)}$ in that paper [11]. For obtaining the moments of $V^{(s)}$ when $f_2 \leq p$ replace in the expression of the moments in (4.9) and (4.10) f_1 by $f_1 - f_2 + p$, f_2 by p and p by f_2 as in (1.2).

5. The First three moments of $U^{(p)}$. We prove first the following theorem for obtaining the moments of $U^{(p)}$ [7,8,9,10].

Theorem III. Let $\underline{M}: p \times p = (m_{ij})$ be distributed as

$$(5.1) \quad \prod_{i=1}^p \prod_{j=i+1}^p \left\{ \frac{\Gamma\left(\frac{f_1+f_2-i+1}{2}\right)}{\Gamma\left(\frac{f_1-i+1}{2}\right)\Gamma\left(\frac{f_2-i+1}{2}\right)} \right\} |\underline{M}|^{\frac{1}{2}(f_2-p-1)} \\ \prod_{i=1}^p \prod_{j=i+1}^p \frac{\Gamma\left(\frac{f_1+f_2}{2}\right)}{\Gamma\left(\frac{f_1}{2}\right)\Gamma\left(\frac{f_2}{2}\right)} d\underline{M}.$$

Then for $f_1 > (p+1)$,

$$(5.2) \quad E(m_{ij}) = f_2 / (f_1 - p - 1) \quad \text{if } i=j \\ = 0 \quad \text{otherwise}$$

and for $f_1 > (p+3)$,

$$(5.3) \quad E(m_{ij}m_{i'j'}) = \begin{cases} f_2(f_2+2)/\{(f_1-p-1)(f_1-p-3)\} & \text{if } i=j=i'=j' \\ f_2(f_2+f_1-p-1)/\{(f_1-p)(f_1-p-1)(f_1-p-3)\} & \\ & \text{if } i=i', j=j', i \neq j \\ f_2\{(f_1-p)(f_1-p-1)\}^{-1}[(f_2-1)+(f_2+f_1-p-1)(f_1-p-3)^{-1}] & \\ & \text{if } i=j, i'=j', i \neq i' \\ 0 & \text{otherwise.} \end{cases}$$

Proof: \underline{M} is symmetric and positive definite and for evaluating $E(m_{ij})$ and $E(m_{ij}m_{i'j'})$ it is easy to see from (3.2) and (3.3) the various cases which should be considered separately.

Moreover, we may note that

$$m_{11}, W: (p-1) \times 1 = \{m_{11}(1+m_{11})\}^{\frac{1}{2}} T_1^{-1} m \text{ and } M_{22.1} = M_{11}^{-m} m' / m_{11}$$

are independently distributed and their respective density functions are

$$(5.4) \quad \{B[\frac{1}{2} f_2, \frac{1}{2}(f_1-p+1)]\}^{-1} m_{11}^{\frac{1}{2} f_2 - 1} (1+m_{11})^{-\frac{1}{2} (f_1+f_2-p+1)},$$

$$(5.5) \quad \Pi^{\frac{1}{2}(p-1)} \{\Gamma(\frac{f_1+f_2-p+1}{2})\}^{-1} \{\Gamma(\frac{f_1+f_2}{2})\} (1+W' W)^{\frac{1}{2} (f_1+f_2)},$$

and

$$(5.6) \quad \Pi^{\frac{1}{4}(p-1)(p-2)} \prod_{i=1}^{p-1} \left\{ \frac{\Gamma(\frac{f_1+f_2-i}{2})}{\Gamma(\frac{f_1-i+1}{2}) \Gamma(\frac{f_2-i}{2})} \right\} \times \\ |M_{22.1}|^{\frac{1}{2}(f_2-p-1)} |I_{p-1} + M_{22.1}|^{-\frac{1}{2}(f_1+f_2-1)}$$

where

$$M_{22.1} = (m_{ij.1}, i, j=2, 3, \dots, p), I_{p-1} + M_{22.1} = T_1 T_1', \quad T_1: (p-1) \times (p-1)$$

is a lower-triangular matrix and M_{11} is obtained from M by deleting the first row and column.

From the above results, it is easy to verify the following,

$$E(m_{11}) = f_2 / (f_1 - p + 1); \quad E(m_{12}) = (E w_1) E[m_{11}(1+m_{11})(1+m_{22.1})]^{\frac{1}{2}} = 0,$$

$$\begin{aligned} E(m_{12}^2) &= E(w_1^2) [E m_{11}(1+m_{11})] [E(1+m_{22.1})] \\ &= f_2(f_1+f_2^{-p-1})/\{(f_1^{-p})(f_1^{-p-1})(f_1^{-p-3})\}; \end{aligned}$$

$$E(m_{11}m_{22}) = E(m_{11}m_{22.1}) + E(m_{12}^2) = f_2(f_2^{-1}) \{(f_1^{-p})(f_1^{-p-1})\}^{-1} + E(m_{12}^2);$$

$$E(m_{11}m_{12}) = E(w_1) [E m_{11}^{3/2}(1+m_{11})^{1/2} m_{22.1}^{1/2}] = 0;$$

$$\begin{aligned} E(m_{12}m_{13}) &= E\{m_{11}(1+m_{11})w_1^2 m_{23.1}\} + E\{m_{11}(1+m_{11}) \left((1+m_{33.1})^{-\frac{m_{23.1}^2}{1+m_{22.1}}} \right)^{1/2} w_1 w_2\} \\ &= 0, \end{aligned}$$

where w_1 and w_2 are the first two elements in \underline{W} . Again

$$E(m_{12}m_{34}) = 0.$$

This proves the theorem III.

Lemma I: If \underline{L} : $p \times p$ is a symmetric and positive definite matrix and

$$U^{(p)} = t_r(\underline{I}_p - \underline{L})^{-1} - p, \text{ then}$$

$$(5.7) \quad 1+U^{(p)} = \{(1 - \underline{l}_{11})(1 - \underline{u}' \underline{u})\}^{-1} + (1 - \underline{u} \underline{u}')^{-1} (\underline{u}' \underline{M} \underline{u}) + t_r \underline{M}$$

where

$$\tilde{L} = \begin{pmatrix} l_{11} & \underline{l}' \\ \underline{l} & \underline{L}_{11} \end{pmatrix}, \quad \underline{l} : (p-1) \times 1 = \{l_{11}(1-l_{11})\}^{\frac{1}{2}} (\underline{I}_{p-1} - \underline{L}_{22})^{\frac{1}{2}} \underline{u}$$

$$\underline{L}_{22} : (p-1) \times (p-1) = \underline{L}_{11} - \underline{l} \underline{l}' / l_{11} \quad \text{and} \quad \underline{M} : (p-1) \times (p-1) = (\underline{I}_{p-1} - \underline{L}_{22})^{-1} - \underline{I}_{p-1}$$

Proof: We may note that

$$(\underline{I}_p - \underline{L})^{-1} = \begin{pmatrix} (1-l_{11})^{-\frac{1}{2}} & \underline{0} \\ \underline{0} & (\underline{I}_{p-1} - \underline{L}_{22})^{-\frac{1}{2}} \end{pmatrix} \begin{pmatrix} 1 & \sqrt{l_{11}} \underline{u}' \\ -\sqrt{l_{11}} \underline{u} & \underline{I}_{p-1} - (1-l_{11}) \underline{u} \underline{u}' \end{pmatrix}^{-1} \times$$

$$\begin{pmatrix} (1-l_{11})^{-\frac{1}{2}} & \underline{0} \\ \underline{0} & (\underline{I}_{p-1} - \underline{L}_{22})^{-\frac{1}{2}} \end{pmatrix}$$

and

$$\begin{pmatrix} 1 & -\sqrt{l_{11}} \underline{u}' \\ -\sqrt{l_{11}} \underline{u} & \underline{I}_{p-1} - (1-l_{11}) \underline{u} \underline{u}' \end{pmatrix}^{-1} = \begin{pmatrix} 1 + l_{11} \underline{u}' \underline{u} / (1 - \underline{u}' \underline{u}) & \sqrt{l_{11}} \underline{u}' / (1 - \underline{u}' \underline{u}) \\ \sqrt{l_{11}} \underline{u} / (1 - \underline{u}' \underline{u}) & \underline{I}_{p-1} + \underline{u} \underline{u}' / (1 - \underline{u}' \underline{u}) \end{pmatrix}.$$

Hence

$$t_r(\underline{I}_{\underline{p}} - \underline{L})^{-1} = 1 - (1 - \underline{u}' \underline{y})^{-1} + \{(1 - \ell_{11})(1 - \underline{u}' \underline{y})\}^{-1} + t_r(\underline{I}_{\underline{p}-1} - \underline{L}_{22})^{-1} \\ + \underline{y}'(\underline{I}_{\underline{p}-1} - \underline{L}_{22})^{-1} \underline{y} / (1 - \underline{u}' \underline{y}).$$

From this, the lemma is obvious.

Theorem IV: If the distribution of \underline{L} is non-central (linear) multivariate beta distribution and $U^{(p)} = t_r(\underline{I}_{\underline{p}} - \underline{L})^{-1} - p$, then for $f_1 > (p+1)$,

$$(5.8) \quad E(U^{(p)}) = (pf_2 + 2\lambda^2) / (f_1 - p - 1)$$

and for $f_1 > (p+3)$,

$$(5.9) \quad \text{Var}(U^{(p)}) = 2[4\lambda^4(f_1 - p) + (4\lambda^2 + pf_2)(f_1 - 1)(f_1 + f_2 - p - 1)] / \{(f_1 - p)(f_1 - p - 1)^2 \\ (f_1 - p - 3)\}.$$

Proof: By theorem I, we may note that ℓ_{11} , \underline{u} and $M = (\underline{I}_{\underline{p}-1} - \underline{L}_{22})^{-1} - \underline{I}_{\underline{p}-1}$ are independently distributed and their respective density functions are given by (2.1), (2.4) and

$$\prod_{i=1}^{\frac{1}{4}(p-1)(p-2)p-1} \left\{ \frac{\Gamma(\frac{f_1 + f_2 - i}{2})}{\Gamma(\frac{f_1 - i + 1}{2})\Gamma(\frac{f_2 - i}{2})} \right\} |M|^{\frac{1}{2}(f_2 - p - 1)} \times \\ | \underline{I}_{\underline{p}-1} + M |^{-\frac{1}{2}(f_1 + f_2 - 1)}$$

Let $\ell_{11,0}$ be the variate whose distribution is the same as that of ℓ_{11} when $\lambda^2 = 0$. Then

$$E(1-\ell_{11})^{-1} = E(1-\ell_{11,0})^{-1} + 2\lambda^2/(f_1-2),$$

$$E(1-\ell_{11})^{-2} = E(1-\ell_{11,0})^{-2} + 4\lambda^2\{(f_1+f_2-2)+\lambda^2\}/\{(f_1-2)(f_1-4)\}.$$

If $U_0^{(p)}$ be the $U^{(p)}$ statistic when ℓ_{11} is replaced by $\ell_{11,0}$, then

$$(5.10) \quad E(U^{(p)}) = E(U_0^{(p)}) + [2\lambda^2/(f_1-2)] E(1-\underline{u}' \underline{u})^{-1}$$

and

$$(5.11) \quad E[1+U^{(p)}]^2 = E[1+U_0^{(p)}]^2 + \{4\lambda^2/(f_1-2)\} E\{(1-\underline{u}' \underline{u})^{-1} [t_{rM} + (1-\underline{u}' \underline{u})^{-1} \times \\ (u' M u)]\} + [4\lambda^2(f_1+f_2-2+\lambda^2)/\{(f_1-2)(f_1-4)\}] E(1-\underline{u}' \underline{u})^{-2}.$$

That is,

$$(5.11a) \quad \text{Var}(U^{(p)}) = \text{Var}(U_0^{(p)}) + \alpha,$$

where

$$\alpha = \{4\lambda^2/(f_1-2)\} E\{(1-\underline{u}' \underline{u})^{-1} [t_{rM} + (1-\underline{u}' \underline{u})^{-1} (u' M u)]\} + \\ [4\lambda^2(f_1+f_2-2+\lambda^2)/\{(f_1-2)(f_1-4)\}] E(1-\underline{u}' \underline{u})^{-2} - [4\lambda^4/(f_1-2)^2] \times \\ [E(1-\underline{u}' \underline{u})^{-1}]^2 - 2[2\lambda^2/(f_1-2)] E(1+U_0^{(p)}) E(1-\underline{u}' \underline{u})^{-1}.$$

We note that

$$E(U_0^{(p)}) = pf_2/(f_1-p-1), \quad E(\underline{M}) = (f_2-1) \frac{I_{p-1}}{f_1-p-1}/(f_1-p),$$

$$E(t_{rM}) = (p-1)(f_2-1)/(f_1-p), \quad E(1-\underline{u}'\underline{u})^{-1} = (f_1-2)/(f_1-p-1)$$

and

$$E(1-\underline{u}'\underline{u})^{-2} = (f_1-4)(f_1-2)/\{(f_1-p-1)(f_1-p-3)\}.$$

Putting these values in α , we get

$$(5.12) \quad \alpha = \frac{8\lambda^4}{(f_1-p-1)^2(f_1-p-3)} + \frac{8\lambda^2(f_1-1)(f_1+f_2-p-1)}{(f_1-p)(f_1-p-1)^2(f_1-p-3)}.$$

From theorem III, it is easy to find $\text{Var}(U_0^{(p)})$. However the first four (central) moments of $U_0^{(p)}$ are available in [7, 9, 10] and substituting the value of $\text{Var}(U_0^{(p)})$ in (5.11a), we get theorem IV.

The expressions for moments of $U^{(p)}$ given above check with those obtained by Pillai [11] for $p=2$.

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