

On the Non-Central Multivariate Beta Distribution and the Moments of Traces of Some Matrices

By K. C. Sreedhārān Pillai

Purdue University

1. Introduction and summary. Let A1 and A2 be two positive definite symmetric matrices of order p, A1 having a Wishart distribution [3, 15] with f1 degrees of freedom and A2 having an independent non-central Wishart distribution with f2 degrees of freedom, corresponding to the linear case [1,2]. Now let

A1 = C L C'

where C is a lower triangular matrix such that

A1 + A2 = C C'

It has been shown [6] that the density function of L is given by

(1.1) f(L) = Ke^{-lambda^2/2} 1F1{1/2(f1+f2), 1/2f2, 1/2 lambda^2 (1-l11)} |L|^{(f1-p-1)/2} |I-L|^{(f2-p-1)/2}

where

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

$\lambda^2$  is the single non-centrality parameter in the linear case,  $l_{11}$  is the element in the top left corner of the  $L$  matrix, and  ${}_1F_1$  denotes the confluent hypergeometric function.

In this paper, the density function of  $L$  given by (1.1) has been observed to be a product of density functions of  $p(p+1)/2$  independent beta variables, explicit expressions for these variables being given for  $p = 2, 3, 4$  and  $5$ . In view of the independence of the beta variables, it has been shown how the moments of the trace of  $L$  (say  $W^{(p)}$ ) and of  $I-L$  (say  $V^{(p)}$ , which is actually Pillai's  $V^{(s)}$  criterion with  $s = p$  [8]) can be computed from those of the beta variables. Again, if we denote the characteristic roots of  $I-L$  by  $\theta_i$  ( $i = 1, 2, \dots, p$ ), a method has been given for computing the moments of  $U^{(2)} = \sum_{i=1}^2 [\theta_i/(1-\theta_i)] = \sum_{i=1}^2 \lambda_i$  (a constant times Hotelling's  $T_0^2$ ,  $s = 2$ ), [8], also from those of the independent beta variables. The case of  $p = 2$  has been considered in detail, deriving the first four moments of  $W^{(2)}$ ,  $V^{(2)}$  and  $U^{(2)}$  and suggesting approximate distributions for them.

In addition, for tests of the hypothesis:  $H_0: \lambda = 0$  against  $H_1: \lambda > 0$  based on the three criteria,  $V^{(2)}$ ,  $U^{(2)}$  and Wilks' criterion,  $\Lambda = \prod_{i=1}^2 (1-\theta_i)$ , [16] comparison of power functions has been carried out for different values of  $f_1$  and  $f_2$  using the moments of these criteria. Further, such comparison has been extended to include also Roy's largest root criterion in testing

the hypothesis  $H_0: \rho = 0$  against  $H_1: \rho > 0$  where  $\rho$  is the single non-null population canonical correlation coefficient.

2. Independent beta variables. Let

$$L = T T'$$

where  $T$  is a lower triangular matrix  $[t_{ij}]$ . It has been shown [6] that then the diagonal elements  $t_{ii}$  are independently distributed and that  $t_{ii}^2$  ( $i = 2, 3, \dots, p$ ) follows the distribution

$$(2.1) \quad f_i(t_{ii}^2) = (t_{ii}^2)^{\frac{1}{2}(f_1+1-i)-1} (1-t_{ii}^2)^{\frac{1}{2}f_2-1} / \beta\{\frac{1}{2}(f_1+1-i), \frac{1}{2}f_2\}$$

$$(0 \leq t_{ii}^2 \leq 1),$$

while  $t_{11}^2$  is distributed as

$$(2.2) \quad f_1(t_{11}^2) = \frac{e^{-\lambda^2/2} (t_{11}^2)^{\frac{f_1}{2}-1} (1-t_{11}^2)^{\frac{f_2}{2}-1} {}_1F_1\{\frac{1}{2}(f_1+f_2), \frac{1}{2}f_2, \frac{1}{2}\lambda^2(1-t_{11}^2)\}}{\beta(\frac{1}{2}f_1, \frac{1}{2}f_2)}$$

$$(0 \leq t_{11}^2 \leq 1) .$$

(i)  $p = 2$ . Now, if  $p = 2$ , it can be shown that

$$(2.3) \quad f(\ell_{11}, \ell_{22}, \ell_{21}) = f_1(u_{11})f_2(u_{22})f_{21}(u_{21})$$

where

$$(2.4) \quad u_{11} = t_{11}^2, \quad u_{22} = t_{22}^2 \quad \text{and} \quad u_{21} = t_{21}^2 / \{(1-t_{11}^2)(1-t_{22}^2)\} ,$$

$f_1(u_{11})$  is given by (2.2),  $f_2(u_{22})$  by (2.1) with  $i = 2$ ,

and

$$(2.5) \quad f_{21}(u_{21}) = u_{21}^{\frac{1}{2}-1} (1-u_{21})^{\frac{1}{2}(f_2-1)-1} / \beta\{\frac{1}{2}, \frac{1}{2}(f_2-1)\}, \quad (0 \leq u_{21} \leq 1) .$$

Thus, from (2.3) it may be seen that  $u_{11}$ ,  $u_{22}$  and  $u_{21}$  are independently distributed.

(ii)  $p = 3$ . When  $p = 3$ , it can be shown that

$$(2.6) \quad f(l_{11}, l_{22}, l_{33}, u_{21}, l_{32}, l_{31}) = f_1(u_{11}) f_2(u_{22}) f_3(u_{33}) f_{21}(u_{21}) f_{21}(u_{32}) f_{31}(v_{31})$$

where  $u$ 's are defined in a similar manner as in (2.4),  $v_{31}$  is defined by

$$(2.7) \quad v_{31} = (\sqrt{u_{31}} + \sqrt{u_{21} u_{32} u_{22}})^2 / [(1-u_{21})(1-u_{32})] ,$$

$f_1(u_{11})$  follows (2.2),  $f_i(u_{ii})$  ( $i = 2, 3$ ) is given in (2.1),  $f_{21}(u_{21})$  and  $f_{21}(u_{32})$  both follow the form as in (2.5) and  $f_{31}(v_{31})$  is given by

$$(2.8) \quad f_{31}(v_{31}) = v_{31}^{\frac{1}{2}-1} (1-v_{31})^{\frac{1}{2}(f_2-2)-1} / \beta\{\frac{1}{2}, \frac{1}{2}(f_2-2)\}, \quad (0 \leq v_{31} \leq 1) .$$

(iii)  $p = 4$ . Now, if  $p = 4$ ,

$$f(L) = \left[ \prod_{i=1}^4 f_i(u_{ii}) \right] f_{21}(u_{21}) f_{21}(u_{32}) f_{21}(u_{43}) f_{31}(v_{31}) f_{31}(v_{42}) f_{41}(w_{41}),$$

where  $u$ 's are similarly defined as before,  $v_{31}$  is given in (2.7),  $v_{42}$  is given by

$$(2.9) \quad v_{42} = (\sqrt{u_{42}} + \sqrt{u_{43}u_{32}u_{33}})^2 / [(1-u_{32})(1-u_{43})] ,$$

$$(2.10) \quad w_{41} = (\sqrt{v_{41}} + \sqrt{v_{31}z_{42}})^2 / [(1-v_{31})(1-v_{42})] ,$$

where

$$(2.11) \quad v_{41} = (\sqrt{u_{41}} + \sqrt{u_{42}u_{21}u_{22}})^2 / [(1-u_{21})(1-u_{43})] ,$$

$$(2.12) \quad z_{42} = (\sqrt{u_{42}u_{32}} + \sqrt{u_{43}u_{33}})^2 / [(1-u_{32})(1-u_{43})] ,$$

and where  $f_1(u_{11})$  as before is given by (2.2),  $f_i(u_{ii})(i = 2, 3, 4)$  by (2.1),  $f_{21}(u_{21})$ ,  $f_{21}(u_{32})$  and  $f_{21}(u_{43})$  follow the form (2.5),  $f_{31}(v_{31})$  and  $f_{31}(v_{42})$  follow the form (2.8) and

$$(2.13) \quad f_{41}(w_{41}) = w_{41}^{\frac{1}{2}-1} (1-w_{41})^{\frac{1}{2}(f_2-3)-1} / \beta\left(\frac{1}{2}, \frac{1}{2}(f_2-3)\right), \quad (0 \leq w_{41} \leq 1) .$$

(iv)  $p = 5$ . When  $p = 5$ ,

$$(2.14) \quad f(L) = \left[ \prod_{i=1}^5 f_i(u_{ii}) \right] \left[ \prod_{i=1}^4 f_{21}(u_{i+1,i}) \right] \left[ \prod_{i=1}^3 f_{31}(v_{i+2,i}) \right] \times$$

$$\left[ \prod_{i=1}^2 f_{41}(w_{i+3,i}) \right] f_{51}(x_{51})$$

where  $u$ 's are defined as before,  $v_{31}$  and  $v_{42}$  are given by (2.7) and (2.9) respectively  $v_{53}$  is given by

$$(2.15) \quad v_{53} = (\sqrt{u_{53}} + \sqrt{u_{43}u_{54}u_{44}})^2 / [(1-u_{43})(1-u_{54})] ,$$

$w_{41}$  is defined in (2.10),  $w_{52}$  is given by

$$(2.16) \quad w_{52} = (\sqrt{v_{52}} + \sqrt{v_{42}z_{53}})^2 / (1-v_{42})(1-v_{53}) ,$$

where

$$(2.17) \quad v_{52} = (\sqrt{u_{52}} + \sqrt{u_{32}u_{53}u_{33}})^2 / [(1-u_{32})(1-u_{54})]$$

$$(2.18) \quad z_{53} = (\sqrt{u_{53}u_{43}} + \sqrt{u_{54}u_{44}})^2 / [(1-u_{43})(1-u_{54})] ,$$

and where  $x_{51}$  is given by

$$(2.19) \quad x_{51} = \frac{[(\sqrt{v_{51}} + \sqrt{v_{31}z_{52}})(1-v_{42}) + (\sqrt{v_{41}} + \sqrt{v_{31}z_{42}})(\sqrt{v_{42}v_{52}} + \sqrt{z_{53}})]^2}{(1-v_{31})(1-v_{42})^2(1-v_{53})(1-w_{41})(1-w_{52})}$$

and where

$$(2.20) \quad v_{51} = (\sqrt{u_{51}} + \sqrt{u_{21}u_{52}u_{22}})^2 / [(1-u_{21})(1-u_{54})] ,$$

$$(2.21) \quad \text{and} \quad z_{52} = (\sqrt{u_{52}u_{32}} + \sqrt{u_{53}u_{33}})^2 / [(1-u_{32})(1-u_{54})] .$$

Here again  $f_1(u_{11})$  is given by (2.2),  $f_i(u_{ii})$  ( $i = 2, 3, 4, 5$ ) is given by (2.1),  $f_{21}(u_{21})$ ,  $f_{21}(u_{32})$ ,  $f_{21}(u_{43})$  and  $f_{21}(u_{54})$  follow the form (2.5),  $f_{31}(v_{31})$ ,  $f_{31}(v_{42})$  and  $f_{31}(v_{53})$  follow the form (2.8),  $f_{41}(w_{41})$  and  $f_{41}(w_{52})$  follow the form (2.13) and  $f_{51}(x_{51})$  is given by

$$(2.22) \quad f_{51}(x_{51}) = x_{51}^{\frac{1}{2}-1} (1-x_{51})^{\frac{1}{2}(f_2-4)-1} / \beta(\frac{1}{2}, \frac{1}{2}(f_2-4)), \quad 0 \leq x_{51} \leq 1.$$

(v) General case (p). In this subsection, for convenience, let us relabel  $u_{i+1,i}$  as  $u_{i+1,i}^{(2)}$ ,  $i = 1, 2, \dots, p-1$ ;  $v_{i+2,i}$  as  $u_{i+2,i}^{(3)}$ ,  $i = 1, \dots, p-2$ ;  $w_{i+3,i}$  as  $u_{i+3,i}^{(4)}$ ,  $i = 1, \dots, p-3$ ;  $x_{i+4,i}$  as  $u_{i+4,i}^{(5)}$ ,  $i = 1, \dots, p-4$ ; etc. Now from (2.4),

$$(2.23) \quad 1 - u_{21}^{(2)} = |I - L|_{(p=2)} / [(1-u_{11})(1-u_{22})],$$

where  $L = T T'$ . Further,  $u_{32}^{(2)}$  is obtained from  $u_{21}^{(2)}$  by adding simultaneously unity to both suffixes of each of the  $t$ 's involved in  $u_{21}^{(2)}$ , which is reflected in the notation 32 which replaces 21. Similarly  $u_{43}^{(2)}$  is obtained from  $u_{32}^{(2)}$ ,  $u_{54}^{(2)}$  from  $u_{43}^{(2)}$  etc. Again,

$$(2.24) \quad 1 - u_{31}^{(3)} = \frac{|I - L|_{(p=3)}}{(1-u_{11})(1-u_{22})(1-u_{33})(1-u_{21}^{(2)})(1-u_{32}^{(2)})}$$

Further,  $u_{42}^{(3)}$  is obtained from  $u_{31}^{(3)}$  by increasing as before both suffixes in each of the  $t$ 's in  $u_{31}^{(3)}$  by unity. Similarly  $u_{53}^{(3)}$  is obtained from  $u_{42}^{(3)}$  etc. Following this pattern, it is easy to see that

$$(2.25) \quad 1 - u_{pl}^{(p)} = \frac{|I - L|}{\prod_{i=1}^p (1 - u_{ii}^{(1)}) \prod_{i=1}^{p-1} (1 - u_{i+1,i}^{(2)}) \prod_{i=1}^{p-2} (1 - u_{i+2,i}^{(3)}) \dots \prod_{i=1}^2 (1 - u_{i+p-2,i}^{(p-1)})}$$

Hence it may be seen that in the case of  $p$  variables \*

$$(2.26) \quad f(L) = \left[ \prod_{i=1}^p f_{i,i}(u_{ii}) \right] \left[ \prod_{i=1}^{p-1} f_{21}(u_{i+1,i}^{(2)}) \right] \times \\ \left[ \prod_{i=1}^{p-2} f_{31}(u_{i+2,i}^{(3)}) \right] \dots \left[ \prod_{i=1}^2 f_{p-1,1}(u_{i+p-2,i}^{(p-1)}) \right] f_{pl}(u_{pl}^{(p)}),$$

where

$$(2.27) \quad f_{j1}(u_{i+j-1,i}^{(j)}) = (u_{i+j-1,i}^{(j)})^{\frac{1}{2}-1} (1 - u_{i+j-1,i}^{(j)})^{\frac{1}{2}(f_2 - j + 1) - 1} / \beta(\frac{1}{2}, \frac{1}{2}(f_2 - j + 1)), \\ 0 \leq u_{i+j-1,i}^{(j)} \leq 1, \quad j = 2, 3, \dots, p.$$

Further, it may be noted that  $K$  in (1.1) equals

$$(2.28) \quad \prod_{i=1}^p \beta(\frac{1}{2}(f_1 + 1 - i), \frac{1}{2}(f_2)) [\beta(\frac{1}{2}, \frac{1}{2}(f_2 - i))]^{p-i}.$$

---

\*Since this paper was written, a theorem was proved to establish this. (See Khatri, C.G. and Pillai, K.C.S. (1965) "Some Results on the Non-Central Multivariate Beta Distribution and Moments of Traces of Two Matrices", Ann. Math. Statist., 36, October Issue.)



### 3. Traces of some matrices as functions of independent beta variables.

First, consider the trace of  $L$  when  $p = 2$ . Noting that

$$(3.1) \quad l_{11} + l_{22} = t_{11}^2 + t_{22}^2 + t_{21}^2$$

and using (2.4) we get

$$(3.2) \quad W^{(2)} = l_{11} + l_{22} = u_{11} + u_{22} + u_{21}(1-u_{11})(1-u_{22}) .$$

Similarly

$$(3.3) \quad V^{(2)} = 2 - W^{(2)} = (1-u_{11}) + (1-u_{22}) - u_{21}(1-u_{11})(1-u_{22}) .$$

When  $p = 3$ ,

$$(3.4) \quad W^{(3)} = u_{11} + u_{22} + u_{33} + u_{21}(1-u_{11})(1-u_{22}) + u_{32}(1-u_{22})(1-u_{33}) \\ + (1-u_{11})(1-u_{33}) \left[ v_{31}(1-u_{21})(1-u_{32}) + u_{21}u_{22}u_{32} \right. \\ \left. - 2\sqrt{v_{31}(1-u_{21})(1-u_{32})u_{21}u_{22}u_{32}} \right]$$

$$\text{and } V^{(3)} = 3 - W^{(3)} .$$

Similarly,  $W^{(4)}$ ,  $V^{(4)}$ , and  $W^{(5)}$  and  $V^{(5)}$  can be expressed explicitly as functions of independent beta variables.

Now consider  $U^{(2)} = \sum_{i=1}^2 \lambda_i$ . It may be seen that

$$(3.5) \quad U^{(2)} = \sum_{i=1}^2 [\theta_i / (1 - \theta_i)] = \{[(1 - \theta_1) + (1 - \theta_2)] / [(1 - \theta_1)(1 - \theta_2)]\} - 2.$$

Noting that  $(1 - \theta_1) + (1 - \theta_2) = W^{(2)}$  and  $(1 - \theta_1)(1 - \theta_2) = u_{11}u_{22}$  we get

$$(3.6) \quad U^{(2)} = \frac{1 - u_{11}}{u_{11}} + \frac{1 - u_{22}}{u_{22}} + \frac{u_{21}(1 - u_{11})(1 - u_{22})}{u_{11}u_{22}}.$$

4. Moments of  $W^{(2)}$ ,  $V^{(2)}$  and  $U^{(2)}$ . The first four moments of  $W^{(2)}$  will be given by

$$(4.1) \quad \mu_1'(W^{(2)}) = \left\{ 2f_1 e^{-\lambda^2/2} / (v-1) \right\} \sum_{i=0}^{\infty} a_i \left( \frac{1}{2}\lambda^2 \right)^i / i!$$

where

$$(4.2) \quad a_i = (v+i-1)/g_i$$

$$v = (f_1 + f_2) \quad \text{and} \quad g_i = v + 2i.$$

$$(4.3) \quad \mu_2'(W^{(2)}) = \left\{ 4f_1 e^{-\lambda^2/2} / (v^2-1) \right\} \sum_{i=0}^{\infty} b_i \left( \frac{1}{2}\lambda^2 \right)^i / i!$$

where

$$(4.4) \quad b_i = \{f_1 v^2 + 2(i+1)f_1^2 + (i^2 + 3i - 1)f_1 + (2i+3)f_1 f_2 + f_2^2 + (2i-1)f_2 + 2(i^2 - 1)\} / e_0,$$

where  $e_0 = g_i(g_i + 2)$  .

$$(4.5) \quad \mu_3^*(W^{(2)}) = [8f_1 e^{-\lambda^2/2} / \{(v^2 - 1)(v+3)\}] \sum_{i=0}^{\infty} c_i (\frac{1}{2}\lambda^2)^i / i! ,$$

where

$$(4.6) \quad c_i = e_1 / e_2 ,$$

and where

$$\begin{aligned} e_1 = & f_1^2 v^3 + (3i+9)f_1^4 + (6i+21)f_1^3 f_2 + (3i+15)f_1^2 f_2^2 + (3i^2 + 21i + 25)f_1^3 \\ & + (3i^2 + 30i + 41)f_1^2 f_2 + (i^3 + 18i^2 + 44i + 15)f_1^2 + (9i+18)f_1 f_2^2 \\ & + 3f_1 f_2^3 + 2f_2^3 + (12i^2 + 39i + 9)f_1 f_2 + 6i f_2^2 \\ & + (6i^3 + 30i^2 + 18i - 26)f_1 + (12i^2 + 6i - 26)f_2 + 8i^3 + 12i^2 - 20i - 24 , \end{aligned}$$

and

$$e_2 = g_i(g_i + 2)(g_i + 4) .$$

$$(4.7) \quad \mu_4^i(W^{(2)}) = [f_1 e^{-\lambda^2/2} / \{(v^2-1)(v+3)(v+5)\}] \sum_{i=0}^{\infty} d_i (\frac{1}{2}\lambda^2)^i / i! ,$$

where

$$(4.8) \quad d_i = e_3 / e_4 ,$$

and where

$$\begin{aligned} e_3 = & (v+5)[(f_1+2)(f_1+4)(v+1)(v+3)(vf_1+4f_1g_i+23f_1+6v-4g_i-30) \\ & + 4(f_1+2)(v+3)h_i(f_1v+3f_1g_i+19f_1+4v-3g_i-14) \\ & + 2(f_1^2-1)(g_i+4)(g_i+6)(2f_1g_i+3f_1v+13f_1+6v+6g_i+30) \\ & + 12(f_1-1)(g_i+6)h_i(f_1g_i+4f_1+g_i+3h_i+10) \\ & + 6h_i(h_i+2)(3f_1v+9f_1+6v+10h_i+58)] \\ & + (g_i+5)[(f_1-1)(g_i+6)\{(f_1(g_i+5)+h_i+4)(f_1(g_i+5)+12h_i+6)+45h_i(h_i+2)\} \\ & + 105h_i(h_i+2)(h_i+4)] , \end{aligned}$$

and

$$e_4 = g_i(g_i+2)(g_i+4)(g_i+6) ,$$

and where

$$h_i = f_2 + 2i .$$

It may be observed that the moments of  $V^{(2)}$  can be obtained from those of  $W^{(2)}$  using the relation  $V^{(2)} = 2 - W^{(2)}$ , which is given in terms of the  $u$ 's in (3.3).

Now consider the moments of  $U^{(2)}$ . From (3.6)

$$(4.9) \quad U^{(2)} = z_1 + z_2 + z_1 z_2 u_{21}$$

where  $z_1 = (1 - u_{11})/u_{11}$  and  $z_2 = (1 - u_{22})/u_{22}$ .

From (2.2) we get

$$(4.10) \quad f(z_1) = e^{-\lambda^2/2} z_1^{\frac{1}{2}f_2-1} {}_1F_1\left\{\frac{1}{2}v, \frac{1}{2}f_2, \left(\frac{1}{2}\lambda^2 z_1 / (1+z_1)\right)\right\} / \left\{(1+z_1)^{\frac{v}{2}} \beta\left(\frac{1}{2}f_2, \frac{1}{2}f_1\right)\right\} .$$

Similarly from (2.1) ,

$$(4.11) \quad f(z_2) = z_2^{\frac{1}{2}f_2-1} / \left\{(1+z_2)^{\frac{1}{2}(v-1)} \beta\left(\frac{1}{2}f_2, \frac{1}{2}(f_1-1)\right)\right\} .$$

Now using (4.10), (4.11) and (2.5) we obtain the first four moments of  $U^{(2)}$  as follows:

$$(4.12) \quad \begin{aligned} \mu_1'(U^{(2)}) &= \left\{2 e^{-\lambda^2/2} / (f_1-3)\right\} \sum_{i=0}^{\infty} (f_2+i) \left(\frac{1}{2}\lambda^2\right)^i / i! , \\ &= (2f_2 + \lambda^2) / (f_1-3) . \end{aligned}$$

Similarly

$$(4.13) \quad \mu_2^i(U^{(2)}) = [\lambda^4(f_1-2)+4(\lambda^2+f_2)(f_1+f_2(f_1-3)-1)]/[(f_1-2)(f_1-3)(f_1-5)],$$

$$(4.14) \quad \mu_3^i(U^{(2)}) = [\lambda^6(f_1-2)+3\lambda^4\{(f_1-2)(f_2+4)+f_2(f_1-6)+4\} \\ + (3\lambda^2+2f_2)(f_2+2)\{(f_1-2)(f_2+4) \\ + 3(f_2(f_1-6)+4)\}]/[(f_1-2)(f_1-3)(f_1-5)(f_1-7)],$$

and

$$(4.15) \quad \mu_4^i(U^{(2)}) = [\lambda^8 b + \lambda^6 \{(12+s_1)b+4B\} + \lambda^4 \{(28+6s_1+s_2)b+12(f_2+4)B+6A\} \\ + \lambda^2 \{(8+4s_1+2s_2+s_3)b+16(f_2+2)(f_2+4)B+12(f_2+2)A\} \\ + 2f_2(f_2+2)\{(f_2+4)(f_2+6)b \\ + 4(f_2+4)B+3A\}]/[(f_1-2)(f_1-3)(f_1-4)(f_1-5)(f_1-7)(f_1-9)],$$

where  $s_i$  is the  $i$ th ( $i=1,2,3$ ) elementary symmetric function in the arguments  $f_2, f_2+2, f_2+4$  and  $f_2+6$ ,

$$A = f_2^2(f_1-6)(f_1-8)+2f_2(f_1-4)(f_1-6)+16f_1-72,$$

$$B = (f_1-4)(f_2(f_1-8)+6) \quad \text{and} \quad b = (f_1-2)(f_1-4).$$

It may be observed that when  $\lambda = 0$ , the moments given in this section reduce to those obtained by Pillai [8], [9], [10], [11].

5. Approximations to the distributions of  $W^{(2)}$ ,  $V^{(2)}$  and  $U^{(2)}$ .

On the basis of the moments presented in the preceding section, the following approximation to the distribution of  $W^{(2)}$  is suggested for small values of  $\lambda$ :

$$(5.1) \quad g_1(W^{(2)}) = (W^{(2)})^{p_1-1} (1-W^{(2)}/2)^{q_1-1} / [2^{p_1} \beta(p_1, q_1)], 0 < W^{(2)} < 2,$$

where

$$p_1 = [(2K_1 - K_2)K_1] / [2(K_2 - K_1^2)],$$

$$q_1 = [(2 - K_1)(2K_1 - K_2)] / [2(K_2 - K_1^2)],$$

where

$$K_1 = 2f_1\{1 - (\lambda^2/2)/(v+2)\} / v$$

and

$$K_2 = 4f_1(f_1 v + v - 2)\{1 - \lambda^2/(v+4)\} / \{(v-1)v(v+2)\}.$$

A comparison of the lower order moments from (5.1) with the respective exact ones may be made from Table 1.

Since  $V^{(2)} = 2 - W^{(2)}$ , an approximation to the distribution of  $V^{(2)}$  can be obtained from (5.1) in the following form:

$$(5.2) \quad g_2(V^{(2)}) = (V^{(2)})^{q_1-1} (1-V^{(2)}/2)^{p_1-1} / [2^{q_1} \beta(q_1, p_1)], 0 < V^{(2)} < 2.$$

Table 1

Moments (central) of  $W^{(2)}$  from the exact and approximate distributions for different values of  $f_1$  and  $f_2$  and  $\lambda = 2$ .

Moments	$f_1 = 10 \quad f_2 = 5$			$f_1 = 100 \quad f_2 = 5$		
	Exact	Approximate	Ratio (A/E)	Exact	Approximate	Ratio (A/E)
$\mu_1$	1.2134	1.1765	.9696	1.8708	1.8692	.9991
$\mu_2$	0.0506	0.0578	1.1404	$0.0^2 269$	$0.0^2 284$	1.0560
$\mu_3$	$-0.0^2 151$	$-0.0^2 229$	1.5230	$-0.0^4 919$	$-0.0^3 113$	1.2319
$\mu_4$	$0.0^2 731$	$0.0^2 907$	1.2405	$0.0^4 255$	$0.0^4 303$	1.1900
$\sqrt{\mu_2}$	0.2250	0.2403	1.0679	0.0518	0.0533	1.0276
$\beta_1$	0.0175	0.0273	1.5639	0.4350	0.5605	1.2886
$\beta_2$	2.8510	2.7192	0.9538	3.5265	3.7632	1.0671
Moments	$f_1 = 20 \quad f_2 = 20$			$f_1 = 100 \quad f_2 = 80$		
	Exact	Approximate	Ratio (A/E)	Exact	Approximate	Ratio (A/E)
$\mu_1$	0.9575	0.9524	.9947	1.0992	1.0989	0.9997
$\mu_2$	0.0229	0.0232	1.0103	$0.0^2 5419$	$0.0^2 5425$	1.0012
$\mu_3$	$0.0^3 112$	$0.0^3 100$	0.8899	$-0.0^4 109$	$-0.0^4 117$	1.0722
$\mu_4$	$0.0^2 151$	$0.0^2 154$	1.0148	$0.0^4 872$	$0.0^4 874$	1.0020
$\sqrt{\mu_2}$	0.1514	0.1522	1.0052	0.0736	0.0737	1.0006
$\beta_1$	$0.0^2 105$	$0.0^3 806$	0.7678	$0.0^3 748$	$0.0^3 856$	1.1456
$\beta_2$	2.8849	2.8681	0.9942	2.9695	2.9688	0.9997
Moments	$f_1 = 5 \quad f_2 = 20$			$f_1 = 5 \quad f_2 = 100$		
	Exact	Approximate	Ratio (A/E)	Exact	Approximate	Ratio (A/E)
$\mu_1$	0.3751	0.3704	0.9875	0.0935	0.0935	0.9991
$\mu_2$	0.0208	0.0203	0.9782	$0.0^2 162$	$0.0^2 162$	0.9976
$\mu_3$	$0.0^2 160$	$0.0^2 167$	1.0437	$0.0^4 522$	$0.0^4 530$	1.0136
$\mu_4$	$0.0^2 139$	$0.0^2 136$	0.9806	$0.0^4 102$	$0.0^4 103$	1.0075
$\sqrt{\mu_2}$	0.1442	0.1426	0.9890	0.0403	0.0403	0.9988
$\beta_1$	0.2848	0.3314	1.1637	0.6368	0.6591	1.0350
$\beta_2$	3.2123	3.2920	1.0248	3.8780	3.9262	1.0124



Again, consider  $U^{(2)}$ . An approximation to the distribution of  $U^{(2)}$  for  $f_1 > f_2$  and which is good even for very small values of  $f_2$  is given below:

$$(5.3) \quad g_3(U^{(2)}) = (U^{(2)})^{p_2-1} / \{ (1+U^{(2)})/K_3 \}^{p_2+q_2+1} K_3^{p_2} \beta(p_2, q_2+1) \},$$

$$0 < U^{(2)} < \infty,$$

where

$$p_2 = 2q_2 / \{q_2(h-1) - 2h\}$$

$$q_2 = 2\{c^2(f_1-5)h - (c+d)^2(f_1-3)\} / \{c^2(f_1-5)(h+1) - 2(c+d)^2(f_1-3)\},$$

$$K_3 = c\{q_2(h-1) - 2h\} / \{2(f_1-3)\},$$

$$h = (c+1.99d)^3(f_1-3) / \{(c+d)^2(f_1-7)c\},$$

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

A comparison of the moments from (5.3) with the respective exact ones may be made from Table 2.

6. Power functions of tests of hypothesis:  $\lambda = 0$  against  $\lambda > 0$  based on  $V^{(2)}$ ,  $U^{(2)}$  and  $\Lambda$ . Using the results on the moments of  $W^{(2)}$  in section 4, and the relation  $V^{(2)} = 2W^{(2)}$ , the central moments,  $\mu_2$ ,  $\mu_3$ , and  $\mu_4$ , and the moment quotients,  $\beta_1$  and  $\beta_2$ , were computed for various values of  $f_1$ ,  $f_2$ , and  $\lambda$ . Similar computations were made for  $U^{(2)}$  and Wilks' criterion, using the expressions in section 4 for the

Table 2

Moments (central) of  $U^{(2)}$  from the exact and approximate distributions for different values of  $f_1 > f_2$  and  $\lambda = 1, 3, \text{ and } 5$

Moments	$f_1 = 10, f_2 = 2, \lambda = 1$			$f_1 = 15, f_2 = 5, \lambda = 5$		
	Exact	Approximate	Ratio (A/E)	Exact	Approximate	Ratio (A/E)
$\mu_1$	0.7143	0.7143	1.0000	2.9167	2.9167	1.0000
$\mu_2$	0.5041	0.4760	0.9442	2.3937	2.1092	0.8812
$\mu_3$	1.5792	1.5333	0.9709	8.0457	6.8781	0.8549
$\mu_4$	25.7893	27.0736	1.0498	82.9146	67.0244	0.8084
$\sqrt{\mu_2}$	0.7100	0.6899	0.9717	1.5472	1.4523	0.9387
$\beta_1$	19.4703	21.8049	1.1199	4.7198	5.0415	1.0682
$\beta_2$	101.4935	119.5121	1.1775	14.4708	15.0656	1.0411
Moments	$f_1 = 50, f_2 = 10, \lambda = 1$			$f_1 = 100, f_2 = 10, \lambda = 1$		
	Exact	Approximate	Ratio (A/E)	Exact	Approximate	Ratio (A/E)
$\mu_1$	0.4468	0.4468	1.0000	0.2165	0.2165	1.0000
$\mu_2$	0.0258	0.0253	0.9823	$0.0^2 532$	$0.0^2 522$	0.9798
$\mu_3$	$0.0^2 373$	$0.0^2 407$	1.0911	$0.0^3 295$	$0.0^3 309$	1.0467
$\mu_4$	$0.0^2 296$	$0.0^2 314$	1.0616	$0.0^3 112$	$0.0^3 114$	1.0114
$\sqrt{\mu_2}$	0.1605	0.1591	0.9911	0.0730	0.0722	0.9898
$\beta_1$	0.8126	1.0207	1.2561	0.5772	0.6724	1.1648
$\beta_2$	4.4566	4.9032	1.1002	3.9610	4.1731	1.0535
Moments	$f_1 = 100, f_2 = 20, \lambda = 3$			$f_1 = 100, f_2 = 20, \lambda = 5$		
	Exact	Approximate	Ratio (A/E)	Exact	Approximate	Ratio (A/E)
$\mu_1$	0.5052	0.5052	1.0000	0.6701	0.6701	1.0000
$\mu_2$	0.0155	0.0140	0.9031	0.0252	0.0209	0.8292
$\mu_3$	$0.0^2 116$	$0.0^2 106$	0.9199	$0.0^2 236$	$0.0^2 186$	0.7886
$\mu_4$	$0.0^3 874$	$0.0^3 736$	0.8417	$0.0^2 231$	$0.0^2 161$	0.6995
$\sqrt{\mu_2}$	0.1246	0.1184	0.9503	0.1587	0.1446	0.9106
$\beta_1$	0.3578	0.4111	1.1488	0.3467	0.3782	1.0907
$\beta_2$	3.6295	3.7458	1.0320	0.3631	0.3694	1.0173

moments of the former, and deriving the expressions for the moments of the latter as the product of the respective moments of  $u_{11}$  and  $u_{22}$ .

For a given size  $\alpha$ , using the  $\beta_1$  and  $\beta_2$  values computed for fixed  $f_1$  and  $f_2$  for  $\lambda = 0$ , the critical region was determined for each criterion referring to tables of "Percentage points of Pearson curves for  $\beta_1$  and  $\beta_2$  expressed in standardized measure" [7] .

Further, for the same values of  $f_1$  and  $f_2$  and a value of  $\lambda > 0$ , the computed values of  $\beta_1$  and  $\beta_2$  were used to determine from the same table by interpolation the power of the test based on the critical region determined previously. The following table presents the results of these computations.

Table 3

Powers of tests of hypothesis:  $\lambda = 0$  against  $\lambda > 0$   
based on  $V^{(2)}$ ,  $U^{(2)}$  and  $\Lambda$  .

$f_1$	$f_2$	$\alpha$	$\lambda$	$V^{(2)}$	Power $U^{(2)}$	$\Lambda$
50	10	.005	1	.0076	.0076	.0113
50	10	.005	2	.0215	.0217	.0470
100	10	.01	1	.0156	.0156	.0217
100	20	.025	1	.0306	.0306	.0345
100	30	.025	1	.0300	.0300	.0303
100	50	.025	1	.0278	.0277	.0280
100	100	.025	1	.0270	.0270	.0270
50	50	.005	1	.0057	.0056	.0056
50	50	.005	2	.0085	.0083	.0085
50	50	.005	3	.0151	.0151	.0153

Table 3 shows that a) there is practically very little difference between the powers of tests based on  $V^{(2)}$  and  $U^{(2)}$  and b) for small values of  $f_2$  Wilks' criterion seems to have marked power compared to both  $V^{(2)}$  and  $U^{(2)}$ . This point needs further investigation.

7. Power functions for tests of hypothesis:  $\rho = 0$  against  $\rho > 0$  based on  $V^{(2)}$ ,  $U^{(2)}$ ,  $\Lambda$  and the largest root. In the case of relation between a  $p$ -set of variates,  $x' = (x_1, \dots, x_p)$ , and a  $q$ -set,  $y' = (y_1, \dots, y_q)$ , from a  $(p+q)$ -variate normal population, where there is only one non-null population canonical correlation coefficient,  $\rho$ , and  $p \leq q$ ,  $(p+q) < n'$  where  $n'$  is the sample size,

$$(7.1) \quad \lambda^2 = \rho^2 \sum_{t=1}^v y_{1t}^2 / (1-\rho^2)$$

where  $y_{1t}$  ( $t=1, \dots, v$ ) are related to the sample observations of  $y_1$ , and  $y$ , here, is considered fixed [6]. Further,  $f_2 = q$  and  $f_1 = n' - q - 1$  such that  $v = f_1 + f_2$ . If, however,  $y$  is not fixed, then

$\sum_{t=1}^v y_{1t}^2$  in  $\lambda^2$  of (7.1) is a chi-square with  $v$  degrees of freedom and,

therefore, for obtaining the moments of  $W^{(2)}$  in this case the following changes may be made in the moments of  $W^{(2)}$  given in section 4:

$$(7.2) \quad e^{-\frac{1}{2}\lambda^2} \longrightarrow (1-\rho^2)^{v/2}, \quad (\lambda^2)^i \longrightarrow (\rho^2)^i \quad \text{and}$$

$$(a_i, b_i, c_i, d_i) \longrightarrow [v(v+2) \dots (v+2(i-1))] (a_i, b_i, c_i, d_i) .$$

Similar changes apply for Wilks' criterion. But for  $U^{(2)}$ ,  $(\lambda^2)^1$  is replaced by  $(2\rho^2/(1-\rho^2))^1 \Gamma(\frac{1}{2}v+1)/\Gamma(\frac{v}{2})$ . Now for the test of the hypothesis:  $\rho = 0$  against  $\rho > 0$  using  $V^{(2)}$ ,  $U^{(2)}$  and  $\Lambda$ , powers were evaluated for  $\rho = .05$  and  $\rho = .1$  for certain values of  $f_1$  and  $f_2$  using the method discussed in the foregoing section. For the largest root, the power was computed using Constantine's form of the distribution of the canonical correlation coefficients [4], [5] in the following manner:

First the joint distribution for  $p = 2$  and a single nonzero  $\rho$  was obtained as a series of determinants using a lemma by Pillai [12]. Further taking into account the first seven terms of the series and integrating out the smallest root by employing Pillai's method [8, 10], the following expression was obtained for the cdf of the largest canonical correlation coefficient,  $r_2^2$ .

$$\begin{aligned} \Pr\{r_2^2 \leq x\} = K_{\frac{1}{2}} \left\{ -I_0(m+1, n+1) \left[ I(x; m, n) \left\{ \sum_{j=0}^6 (B_j x^{6-j} / (m+n+8-j)) \right\} \right. \right. \\ \left. \left. - x I(x; m+1, n) \left\{ \sum_{j=0}^4 (C_j x^{4-j} / (m+n+7-j)) \right\} \right. \right. \\ \left. \left. - x^2 I(x; m+2, n) \left\{ \sum_{j=0}^2 (D_j x^{2-j} / (m+n+6-j)) \right\} \right. \right. \\ \left. \left. - x^3 I(x; m+3, n) E / (m+n+5) \right] \right\} \end{aligned}$$

$$\begin{aligned}
& +2I(x; 2m+7, 2n+1) \left[ \{B_0/(m+n+8)\} - \{C_0/(m+n+7)\} - \{D_0/(m+n+6)\} - \{E_0/(m+n+5)\} \right] \\
& +2I(x; 2m+6, 2n+1) \left[ \{B_1/(m+n+7)\} - \{C_1/(m+n+6)\} - \{D_1/(m+n+5)\} \right] \\
& +2I(x; 2m+5, 2n+1) \left[ \{B_2/(m+n+6)\} - \{C_2/(m+n+5)\} - \{D_2/(m+n+4)\} \right] \\
& +2I(x; 2m+4, 2n+1) \left[ \{B_3/(m+n+5)\} - \{C_3/(m+n+4)\} \right] \\
& +2I(x; 2m+3, 2n+1) \left[ \{B_4/(m+n+4)\} - \{C_4/(m+n+3)\} \right] \\
& +2I(x; 2m+2, 2n+1) \{B_5/(m+n+3)\} + 2I(x; 2m+1, 2n+1) \{B_6/(m+n+2)\} \}
\end{aligned}$$

where

$$f_1 = 2n+3, \quad f_2 = 2m+3,$$

$$K_{\pm} = (1-\rho^2)^{v/2} C(2, m, n),$$

$$C(2, m, n) = \Gamma(2m+2n+5) / \{4\Gamma(2m+2)\Gamma(2n+2)\},$$

$$I_0(m+1, n+1) = x^{m+1}(1-x)^{n+1}, \quad I(x; c', d') = \int_0^x \theta^{c'} (1-\theta)^{d'} d\theta,$$

$$B_0 = 231A_6, \quad B_1 = 63A_5 + (m+7)B_0/(m+n+8), \quad B_2 = 35A_4 + (m+6)B_1/(m+n+7),$$

$$B_3 = 5A_3 + (m+5)B_2/(m+n+6), \quad B_4 = 3A_2 + (m+4)B_3/(m+n+5),$$

$$B_5 = A_1 + (m+3)B_4/(m+n+4), \quad B_6 = 1 + (m+2)B_5/(m+n+3),$$

$$C_0 = 105A_6, \quad C_1 = 28A_5 + (m+6)C_0/(m+n+7), \quad C_2 = 15A_4 + (m+5)C_1/(m+n+6),$$

$$C_3 = 2A_3 + (m+4)C_2/(m+n+5), \quad C_4 = A_2 + (m+3)C_3/(m+n+4)$$

$$D_0 = 21A_6, \quad D_1 = 5A_5 + (m+5)D_0/(m+n+6), \quad D_2 = 2A_4 + (m+4)D_1/(m+n+5),$$

$$\begin{aligned}
E_0 &= 5A_6, \quad A_1 = v^2 \rho^2 / 2^2 f_2, \\
A_2 &= [v(v+2)]^2 \rho^4 / [f_2(f_2+2)2^6], \\
A_3 &= A_2(v+4)^2 \rho^2 / [(f_2+4)2.3!], \\
A_4 &= A_3(v+6)^2 \rho^2 / \{2^6(f_2+6)\}, \\
A_5 &= A_4(v+8)^2 \rho^2 / \{2^2.5(f_2+8)\}, \quad \text{and} \\
A_6 &= A_5(v+10)^2 \rho^2 / \{2^3.6(f_2+10)\}.
\end{aligned}$$

For  $\rho = 0$ , upper 1% points of the largest root were taken from Pillai's tables [11] for values of  $m = 2$  and  $5$  and  $n = 10, 15, 20, 25, 30, 40$  and  $60$ . Using these  $x_{.99}$  values to determine the critical region, the powers of the largest root test were computed for  $\rho = .05$  and  $\rho = .1$  for values of  $m$  and  $n$  given above. These are shown in Table 4.

Table 4

Powers of the largest root test for testing  $\rho = 0$   
against  $\rho = .05$  and  $\rho = .1$  and  $\alpha = .01$

n	Power			
	$\rho = .05$		$\rho = .1$	
	m = 2	m = 5	m = 2	m = 5
10	.010321	.010247	.011254	.010898
15	.010366	.010311	.011798	.011294
20	.010621	.010394	.012612	.011728
25	.010744	.010518	.013310	.012226
30	.010779	.010611	.013927	.012705
40	.011437	.010812	.016011	.013735
60	.011625	.011248	.019072	.016063

Now, a comparison of the powers of the test of hypothesis:  
 $\rho = 0$  against  $\rho > 0$  based on  $V^{(2)}$ ,  $U^{(2)}$ ,  $\Lambda$  and the largest root  
 may be made from Table 5.

Table 5

Powers of the test  $\rho = 0$  against  $\rho > 0$   
 based on  $V^{(2)}$ ,  $U^{(2)}$ ,  $\Lambda$  and largest root  
 for  $\rho = .05$  and  $\rho = .1$ , and  $\alpha = .01$

$f_1$	Power							
	$\rho = .05$							
	$f_2 = 7$				$f_2 = 13$			
	$V^{(2)}$	$U^{(2)}$	$\Lambda$	largest root	$V^{(2)}$	$U^{(2)}$	$\Lambda$	largest root
53	.0108	.0107	.0118	.0107	.0106	.0106	.0110	.0105
83	.0115	.0115	.0140	.0114	.0112	.0109	.0122	.0108
123	.0123	.0120	.0155	.0116	.0115	.0115	.0130	.0112
$\rho = .1$								
53	.0135	.0135	.0190	.0133	.0125	.0125	.0135	.0122
83	.0165	.0165	.0280	.0160	.0144	.0142	.0180	.0137
123	.0202	.0200	.0440	.0190	.0170	.0170	.0240	.0161

Table 5 shows that a) the largest root has comparatively less power than the other test criteria b)  $V^{(2)}$  and  $U^{(2)}$  practically have equal power and c) Wilks' criterion as in the previous case seems to have greater power for the (small) values of  $f_2$  considered here. Further investigation is being made to clear this point.

The author wishes to thank Mrs. Louise Mao Lui, Statistical Laboratory, Purdue University, for the excellent programming of the material for the computations in this paper carried out on the IBM 7094 Computer, Purdue University's Computer Science's Center.



## REFERENCES

- [1] Anderson, T.W. and Girshick, M.A. (1944). 'Some extensions of the Wishart distribution.' Ann. Math. Statist., 15, 345-357.
- [2] Anderson, T.W. (1946). The non-central Wishart distribution and certain problems of multivariate statistics. Ann. Math. Statist., 17, 409-431.
- [3] Anderson, T.W. (1958). An Introduction to Multivariate Statistical Analysis. John Wiley and Sons, New York.
- [4] Constantine, A.G. (1963). Some non-central distribution problems in multivariate analysis. Ann. Math. Statist., 34, 1270-1285.
- [5] James, Alan T. (1964). Distributions of matrix variates and latent roots derived from normal samples. Ann. Math. Statist., 35, 475-501.
- [6] Kshirsagar, A.M. (1961). The non-central multivariate beta distribution. Ann. Math. Statist., 32, 104-111.
- [7] Pearson, E.S. and Hartley, H.O. (1956). Biometrika Tables for Statisticians, 1. Cambridge University Press for the Biometrika Trustees.
- [8] Pillai, K.C. Sreedharan. (1954). On some distribution problems in multivariate analysis. Mimeograph Series No. 88, Institute of Statistics, University of North Carolina.
- [9] Pillai, K.C. Sreedharan (1955). Some new test criteria in multivariate analysis. Ann. Math. Statist., 26, 117-121.
- [10] Pillai, K.C. Sreedharan (1956). Some results useful in multivariate analysis. Ann. Math. Statist., 27, 1106-1114.
- [11] Pillai, K.C. Sreedharan (1960). Statistical Tables for Tests of Multivariate Hypotheses. The Statistical Center, Manila.
- [12] Pillai, K.C. Sreedharan. (1964). On the moments of elementary symmetric functions of the roots of two matrices. Ann. Math. Statist., 35, 1704-1712.
- [13] Pillai, K.C. Sreedharan and Mijares, Tito A. (1959). On the moments of the trace of a matrix and approximations to its distribution. Ann. Math. Statist., 30, 1135-1140.
- [14] Pillai, K.C. Sreedharan and Samson, Pablo (1959). On Hotelling's generalization of  $T^2$ . Biometrika, 46, 160-168.
- [15] Roy, S.N. (1957). Some Aspects of Multivariate Analysis, John Wiley and Sons, New York.
- [16] Wilks, S.S. (1932). Certain generalizations in the analysis of variance. Biometrika, 24, 471-494.