

A NOTE ON THE CONFIDENCE BOUNDS  
FOR THE CHARACTERISTIC ROOTS  
OF DISPERSION MATRICES OF NORMAL VARIATES

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Institute of Statistics  
Mimeograph Series No. 406  
September 1964

UNIVERSITY OF NORTH CAROLINA

Department of Statistics

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Contract No. AF-AFOSR-84-63

This research was supported by the Mathematics Division of  
the Air Force Office of Scientific Research.

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1. Introduction. Roy and Gnanadesikan [6] have considered certain types of different alternatives in the case of two dispersion matrices, of which first three are strong alternatives, in the sense that all characteristic roots of  $\Sigma_1 \Sigma_2^{-1}$  (i.e.  $\text{ch } \Sigma_1 \Sigma_2^{-1}$ ) are less than or equal to 1, or all  $\text{ch } \Sigma_1 \Sigma_2^{-1} \geq 1$ , or union of these two alternatives. In this note, we derive the shorter confidence bounds than those given by Roy and Gnanadesikan [6] with probability  $\geq (1-\alpha)$ , when the alternative is true. Moreover, we show that when the alternative is complement of the null hypothesis, then we come across the same types of confidence bounds as derived by Anderson [1] by a different approach. Keeping in view the certain types of alternatives for more than two population dispersion matrices, we give shorter confidence bounds than those derived by Gnanadesikan [2,3], with probability  $\geq (1-\alpha)$ . All our confidence bounds are based on either  $X^2$  or  $F$  distributions. The confidence bounds that we derive have not the purpose in mind to give confidence bounds on the partials and hence, for the partials we cannot say that our confidence bounds will at all be suitable and this is true for Anderson's results [1] too. Hence, we shall not try to give or compare any types of partials from them.

Without loss of generality, we shall assume that  $S_i (i=1, 2, \dots, k)$  are independently distributed as Wishart denoted by  $W(S_i, n_i, p, \Sigma_i)$ ,  $(i=1, 2, \dots, k)$  whose density function is given by

$$\frac{1}{2^{pn_i}} \prod_{i=1}^k \frac{1}{\Gamma_p(p-1)} k^{\frac{n_i-j+1}{2}} \{ \prod_{j=1}^k F(\frac{n_i-j+1}{2}) \} |S_i|^{-\frac{1}{2} n_i} |S_i|^{\frac{1}{2} (n_i-p-1)} \exp[-\frac{1}{2} \text{tr } \Sigma_i^{-1} S_i].$$

We shall denote  $ch_j A$  the  $j$ -th maximum characteristic (max. ch.) root of  $A: p \times p$ , i.e.  $ch_1 A \geq ch_2 A \geq \dots \geq ch_p A$ . When  $k = 1$ ,  $S_1$ ,  $n_1$  and  $\Sigma_1$  will be denoted by  $S$ ,  $n$ , and  $\Sigma$  respectively.

2. Simultaneous confidence bounds on  $ch_p \Sigma$  or  $ch_1 \Sigma$ :

Lemma 1. Let  $A: p \times p$  be symmetric positive definite and  $A = (a_{ij})$ ,  $A^{-1} = (a^{ij})$ .

Then

$$ch_p A \leq [ a_{ii} , (a^{jj})^{-1} ] \leq ch_1 A$$

Proof:- by definition

$$ch_p A \leq \xi' A \xi \mid \xi' \xi \leq ch_1 A$$

all  
for non-null vectors  $\xi: p \times 1$ . Let  $\xi$  have zero elements except at the  $i$ -th place and so we get

$$ch_p A \leq a_{ii} \leq ch_1 A.$$

Similarly  $ch_p A^{-1} \leq a^{jj} \leq ch_1 A^{-1}$ . Since  $ch_j A^{-1} = (ch_{p-j+1} A)^{-1}$  for  $A$  is positive definite, we get the lemma 1.

Lemma 2. If  $S^* = (s^*_{ij})$  be distributed as  $W(S^*; n, p, D_\gamma)$  where  $D_\gamma = \text{diag. } (\gamma_1, \dots, \gamma_p)$ , then  $(s^*_{ii}/\gamma_i)$  and  $(\gamma_j s^*_{jj})^{-1}$  are independently distributed as  $X^2$  with  $n$  and  $n-p+1$  degrees of freedom respectively when  $i \neq j$ .

This lemma is a special case of Bartlett's decomposition theorem (e.g. see Kshirsagar [4]), and hence the proof is omitted.

Since  $\Sigma$  is symmetric positive definite, we can find an orthogonal matrix  $\Delta$  such that  $\Delta \Sigma \Delta' = D_\gamma$ , a diagonal matrix with  $\gamma_1 \geq \dots \geq \gamma_p > 0$ . Then the distribution of  $S^* = \Delta \Sigma \Delta'$  is  $W(S^*; n, p, D_\gamma)$ .

(2.1) Let us obtain the confidence bound on  $\gamma_1$  with probability  $\geq (1-\alpha)$  by considering the pair of hypotheses  $H_0(\gamma_1=1)$  and  $H_1(\gamma_1 \leq 1)$ . We note by lemma 2,

that  $s_{11}^*/\gamma_1$  and  $(\gamma_1 s_{11}^*)^{-1}$  are distributed as  $X^2$  with  $n$  and  $n-p+1$  d.f.  $H_0(\gamma_1=1)$  against  $H_1(\gamma_1 \leq 1)$  is tested by the critical region

$$s_{11}^* \leq c, \text{ constant, or } (s_{11}^*)^{-1} \leq c, \text{ constant, and on account}$$

of

$$(s_{11}^*)^{-1} \leq s_{11}^*,$$

we shall choose the critical region

$$(1) \quad s_{11}^* \leq c$$

where  $c$  is determined from

$$(2) \quad P_{\chi^2} (X_n^2 \geq c) = 1-\alpha.$$

Since  $s_{11}^*$  depends on the nuisance parameters of  $\Sigma$ , we cannot carry out the exact test given by (1), but we shall obtain the confidence bound on  $\gamma_1$  with probability  $\geq (1-\alpha)$ . We note from (2) that

$$(3) \quad P_{\chi^2} (s_{11}^*/\gamma_1 \geq c \mid H_1) = 1-\alpha,$$

but by lemma 1,  $ch_1 S_{11}^* = ch_1 S_{11}^* \geq s_{11}^*$ . Hence (3) gives us the following confidence bound on  $\gamma_1$

$$(4) \quad \gamma_1 \leq c^{-1} ch_1 S_{11}^*$$

with probability  $\geq (1-\alpha)$ , and it will <sup>be</sup> less than 1 when  $ch_1 S_{11}^* \leq c$ .

(2.2) Applying arguments similar to (2.1), we obtain the confidence bounds on  $\gamma_p$  with probability  $\geq (1-\alpha)$  by considering the pair of hypotheses  $H_0(\gamma_p = 1)$  and  $H_2(\gamma_p \geq 1)$  as

$$(5) \quad \gamma_p \geq b^{-1} ch_p S_{pp}^*$$

where  $b$  is determined from

$$(7) \quad P_{\chi^2} (X_{n-p+1}^2 \leq b) = 1-\alpha.$$

(2.3) Now let us consider the pair of hypotheses  $H_0(\gamma_p = 1 \neq \gamma_1)$  and  $H_3 = H_1 \cup H_2$ .

By union-intersection principle [7], the critical region is

$$(8) \quad \{ b_1 \leq (s^{*pp})^{-1} \} \cup \{ s_{11}^* \leq c_1 \}$$

where, using lemma 2,  $b_1$  and  $c_1$  are obtained from

$$(9) \quad P_{\mathcal{L}}(X_{n-p+1}^2 \leq b_1) P_{\mathcal{L}}(X_n^2 \geq c_1) = 1 - \alpha.$$

The test procedure (8) cannot be carried out in practice. Hence, noting lemma 1, we find the simultaneous confidence bounds on  $\gamma_1$  or  $\gamma_p$  with probability  $\geq (1 - \alpha)$  as

$$(10) \quad \gamma_1 \leq c_1^{-1} ch_1 \underline{\Sigma} \quad \text{or} \quad \gamma_p \geq b_1^{-1} ch_p \underline{\Sigma},$$

where  $b_1$  and  $c_1$  are given by (9).

We note that the confidence bounds given by (4), (5) and (10) are shorter than those which can be derived by Roy and Gnanadesikan's technique [5].

(2.4) Now, we note that Anderson's confidence bound [1] on all  $ch \underline{\Sigma}$  can be derived from the following considerations.

$$\begin{aligned} (1-\alpha) &= P_{\mathcal{L}} \{ X_{n-p+1}^2 \leq b_1 \} P_{\mathcal{L}}(X_n^2 \geq c_1) = P_{\mathcal{L}} [ (\gamma_p s^{*pp})^{-1} \leq b_1 ] P_{\mathcal{L}} ( s_{11}^* / \gamma_1 \geq c_1 ) \\ &= P_{\mathcal{L}} [ c_1^{-1} (s^{*pp})^{-1} \leq \text{all } ch(\underline{\Sigma}) \leq b_1^{-1} s_{11}^* ] \text{ by lemma 2,} \end{aligned}$$

and so using lemma 1, we have

$$(11) \quad 1 - \alpha \leq P_{\mathcal{L}} [ b_1^{-1} ch_p \underline{\Sigma} \leq \text{all } ch(\underline{\Sigma}) \leq c_1^{-1} ch_1 \underline{\Sigma} ].$$

3. Simultaneous confidence bounds on  $ch_p \underline{\Sigma}, \underline{\Sigma}_2^{-1}$  or  $ch_1 \underline{\Sigma}_1 \underline{\Sigma}_2^{-1}$ .

Lemma 2. Let  $\underline{A}$  and  $\underline{B}$  be two  $p \times p$  symmetric positive definite matrices, and let  $\delta_{-1}$  and  $\delta_p$  be two unit  $ch.$  vectors corresponding to the max. and minimum (min.)  $ch.$  roots of  $\underline{A}$ . If  $\underline{A} = (a_{ij})$ ,  $\underline{A}^{-1} = (a^{ij})$ ,  $t_i = \delta_{-1}' \underline{B} \delta_{-1}$  and  $t^i = \delta_{-1}' \underline{B}^{-1} \delta_{-1}$

for  $i=1, p$ , then

$$\text{ch}_{p \sim \sim} AB^{-1} \leq [ a_{ii} t^p, a_{ii}/t_p, t^p/a^{jj}, (t_p a^{jj})^{-1} ]$$

and

$$\text{ch}_{1 \sim \sim} AB^{-1} \geq [ a_{ii} t^1, a_{ii}/t_1, t^1/a^{jj}, (t_1 a^{jj})^{-1} ].$$

Proof. Let  $\Delta : p \times p$  be an orthogonal matrix whose first column is  $\delta_1$  and the last column is  $\delta_p$  such that  $\Delta' A \Delta = D_z$  is a diagonal matrix. Let  $U = \Delta' B \Delta$ . Then  $u_{ii} = t_i$  and  $u^{ii} = t^i$  for  $i=1, p$ . Now, we shall only prove the first part of the lemma 3, for the other part can similarly be proved. We note that

$$\text{ch}_{p \sim \sim} AB^{-1} = \text{ch}_p (D_z^{\frac{1}{2}} U^{-1} D_z^{\frac{1}{2}}) \leq [ z_p t^p, z_p/t_p ] \text{ by using lemma 1.}$$

Moreover, by lemma 1,

$$\text{ch}_p A = z_p \leq [ a_{ii}, a^{jj} ]$$

and so the first part of the lemma 3 is proved, and this proves the lemma.

Since  $\Sigma_i$  ( $i=1, 2$ ) are symmetric positive definite, there exists a non-singular matrix  $C$  such that

$$(12) \quad C \Sigma_1 C' = D_\gamma, \text{ a diagonal matrix, and } C \Sigma_2 C' = I,$$

where  $\gamma_1 \geq \gamma_2 \geq \dots \geq \gamma_p > 0$ ,  $\gamma_i = \text{ch}_i \Sigma_1 \Sigma_2^{-1}$ . Let  $S_1^* = C \Sigma_1 C'$  and  $S_2^* = C \Sigma_2 C'$ .

Then  $S_1^*$  and  $S_2^*$  are independently distributed as

$$(13) \quad W(S_1^*; n_1, p, D_\gamma) \text{ and } W(S_2^*; n_2, p, I).$$

Lemma 4. Let  $\xi_1$  and  $\xi_p$  be the unit ch. vectors corresponding to the max. and min. ch. root of  $S_1^* = (s_{ij}^*)$ , and let  $S_1^*$  and  $S_2^*$  be distributed as (13).

Then  $s_{11}^* (\xi_1' S_2^{*-1} \xi_1) (n_2 - p + 1) / n_1 \gamma_1$  and  $n_2 / \{ s^{*pp} (\xi_p' S_2^* \xi_p) \gamma_p (n_1 - p + 1) \}$ ,

where  $S_1^{*-1} = (s^{*ij})$ , are independently distributed as  $F$  with  $n_1, n_2 - p + 1$  and

$n_1 - p + 1$ ,  $n_2$  d.f. respectively.

Proof. Let  $\xi$  be an orthogonal matrix, whose first and the last columns are  $\xi_1$  and  $\xi_p$ , such that  $\xi' S_1^* \xi$  is a diagonal matrix. Then it is easy to see that  $S_1^*$  and  $\xi' S_2^* \xi = U$  are independently distributed as

$$W(S_1^*; n_1, p, D_\gamma) \text{ and } W(U; n_2, p, I) .$$

We note that if  $U^{-1} = (u^{ij})$  and  $U = (u_{ij})$ ,  $u^{11} = (\xi_1' S_2^{*-1} \xi_1)$  and  $u_{pp} = \xi_p' S_2^* \xi_p$ . Using lemma 2, it can be seen that  $(u^{11})^{-1}$ ,  $u_{pp}$ ,  $s_{11}^* / \gamma_1$ ,  $(s_{pp}^* \gamma_p)^{-1}$  are independently distributed as  $X^2$  with respective d.f.  $n_2 - p + 1$ ,  $n_2, n_1, n_2 - p + 1$ . From this, lemma 4 is obvious.

(2.1) Let us obtain the confidence bound on  $\gamma_1 = \text{ch}_1 \Sigma_1 \Sigma_2^{-1}$  by considering the pair of hypotheses  $H_0(\gamma_1 = 1)$  and  $H_1(\gamma_1 \leq 1)$ . Then, the critical region as in section (2.1) is

$$s_{11}^* \leq \text{a constant} .$$

Here, we note that  $s_{11}^*$  contains unknown parameters depending on  $\Sigma_2$  and so, we require a function depending on  $S_2^*$ . On account of  $(\xi_1' S_2^{*-1} \xi_1) \geq (\xi_1' S_2^* \xi_1)^{-1}$  and lemma 3, we shall choose the critical region

$$(14) \quad s_{11}^* (\xi_1' S_2^{*-1} \xi_1) (n_2 - p + 1) / n_1 \leq c$$

where  $c$  is to be determined from

$$(15) \quad P_{\gamma_1} (F_{n_1, n_2 - p + 1} \geq c) = 1 - \alpha .$$

We note that even though we cannot carry out the test procedure (14), but we can make a confidence statement on  $\gamma_1$  with probability  $\geq (1 - \alpha)$ . By lemma 4, we have

$$(16) \quad (1 - \alpha) = P_{\gamma_1} [ s_{11}^* (\xi_1' S_2^{*-1} \xi_1) (n_2 - p + 1) / n_1 \gamma_1 \geq c ] \\ \leq P_{\gamma_1} [ (\text{ch}_1 S_1 S_2^{-1}) (n_2 - p + 1) / (n_1 c) \geq \gamma_1 ] , \text{ by lemma 3} .$$



Thus, (16) gives the confidence bound on  $\gamma_1$  with probability  $\geq (1-\alpha)$  considering the pair  $H_0(\gamma_1=1)$  and  $H_1(\gamma_1 \leq 1)$ .

(3.2) Similarly, we obtain the confidence bound on  $\gamma_p = \text{ch}_p(\Sigma_1 \Sigma_2^{-1})$  with probability  $\geq (1-\alpha)$  by considering the pair of hypotheses  $H'_0(\gamma_p=1)$  and  $H_2(\gamma_p \geq 1)$  as

$$(17) \quad (\text{ch}_{p-1} s_1 s_2^{-1}) n_2 / b(n_1 - p + 1) \leq \gamma_p$$

where  $b$  is to be determined from

$$(18) \quad P_2(F_{n_1-p+1, n_2} \leq b) = 1-\alpha.$$

(3.3). Here, we give the confidence bound on  $\gamma_1$  or  $\gamma_p$  with probability  $\geq (1-\alpha)$  by considering the hypotheses  $H_0(\gamma_1=\gamma_p=1)$  and  $H_3 = H_1 U H_2$  as

$$(19) \quad \gamma_1 \leq c_1^{-1} (\text{ch}_{p-1} s_1 s_2^{-1}) (n_2 - p + 1) n_1^{-1} \text{ or } \gamma_p \geq b_1^{-1} (n_1 + p + 1)^{-1} n_2 (\text{ch}_{p-1} s_1 s_2^{-1})$$

where  $c_1$  and  $b_1$  are given by

$$(20) \quad P_2(F_{n_1-p+1, n_2} \leq b_1) P_2(F_{n_1, n_2-p+1} \geq c_1) = 1-\alpha$$

The confidence bounds given by (16), (17) and (19) are shorter than those given by Roy and Gnanadesikan [6].

(3.4) Now, we note that Anderson's result [1] on all  $(\text{ch}_{p-1} \Sigma_1 \Sigma_2^{-1})$  can be derived from the following considerations.

$$\begin{aligned} (1-\alpha) &= P_2(F_{n_1-p+1, n_2} \leq b_1) P_2(F_{n_1, n_2-p+1} \geq c_1) \\ &= P_2[n_2 / \{(n_1-p+1) \gamma_p s^{*PP}(\xi_p^* S_p^* \xi_p)\} \leq b_1] P_2[s_{11}^*(\xi_1^* S_2^{*-1} \xi_1)(n_2-p+1)/n_1 \gamma_1 \geq c_1] \\ &= P_2[n_2 / \{(n_1-p+1) s^{*PP}(\xi_p^* S_p^* \xi_p) b_1\} \leq \text{all ch}(\Sigma_1 \Sigma_2^{-1}) \leq s_{11}^*(\xi_1^* S_2^{*-1} \xi_1)(n_2-p+1)/n_1 c_1] \end{aligned}$$

Hence, using lemma 3 we have with probability  $\geq (1-\alpha)$ ,

$$(21) \quad (ch_p S_p^{-1}) n_2 / (n_2 - p + 1) b_1 \leq \text{all } ch (\Sigma_1^{-1}) \leq (ch_1 S_1^{-1}) (n_2 - p + 1) / n_1 c_1,$$

where  $b_1$  and  $c_1$  are given by (20).

#### 4. Simultaneous confidence bounds on $ch_t \Sigma_i$ ( $i=1, 2, \dots, k; t=1, p$ ).

Since  $\Sigma_i$  is symmetric positive definite, there exists an orthogonal matrix  $\Delta_i$  such that  $\Delta_i \Sigma_i \Delta_i' = D_{i,\beta}$ , a diagonal matrix, with diagonal elements  $\beta_{i,1} \geq \beta_{i,2} \geq \dots \geq \beta_{i,p} > 0$  ( $i=1, 2, \dots, k$ ). Let  $S_i^* = \Delta_i S_i \Delta_i'$ . Then  $S_i^*$  are independently distributed as  $W_{\Sigma_i^*}^*(S_i^*; n_i, p, D_{i,\beta})$ , ( $i=1, 2, \dots, k$ ).

(4.1) Let us suppose that the  $k$ -th population is standard. We shall try to obtain the confidence bounds on  $\beta_{i,1} | \beta_{k,p}$  ( $i=1, 2, \dots, k-1$ ) with probability  $\geq (1-\alpha)$  by considering the hypotheses  $H_0 (\beta_{1,j} = \beta_{k,p} \quad j = 1, 2, \dots, k-1)$  and  $H_1 = \bigcup_{i=1}^{k-1} H_{1,i} (\beta_{i,1} \leq \beta_{k,p})$ . We note that for testing  $H_{0,i} (\beta_{i,1} = \beta_{k,p})$  against  $H_{1,i} (\beta_{i,1} \leq \beta_{k,p})$ , we have the critical region

$$(22) \quad w_i : s_{i,11}^* s_k^{*pp} (n_k - p + 1) / n_i \leq d_i'$$

where  $S_k^{*-1} = (s_k^{*ij})$  and  $d_i'$  is to be determined from

$$(23) \quad P_{\Sigma} (F_{n_i, n_k - p + 1} \geq d_i) = 1 - \alpha.$$

By the union intersection principle [7], the critical region for testing  $H_0$  against  $H_1$  is

$$(24) \quad w = \bigcup_{i=1}^k w_i \text{ such that } P_{\Sigma} (x \in w | H_0) = \alpha.$$

Note that we cannot carry out the test procedure (24). Hence, we obtain the confidence bounds on  $(\beta_{i,1} | \beta_{k,p})$  with probability  $\geq (1-\alpha)$  from (24) as

$$(25) \quad \beta_{i,1} | \beta_{k,p} \leq d_i^{-1} (ch_1 S_i) (ch_p S_k)^{-1} (n_k - p + 1) n_i^{-1}, \quad i=1, \dots, k-1$$

where  $d_1, d_2, \dots, d_{k-1}$  are to be calculated from

$$(26) \quad P_{\chi} (F_{n_i, n_k - p + 1} \geq \alpha_i ; i=1, 2, \dots, k-1) = 1 - \alpha$$

(4.2) Similarly, the confidence bounds on  $(\beta_{i,p} | \beta_{k,1})$  with probability  $\geq$

$(1 - \alpha)$  by considering the hypotheses  $H_0(\beta_{i,p} = \beta_{k,1}, i=1, 2, \dots, k)$  and

$H_2 = \bigcup_{i=1}^{k-1} H_{2,i}(\beta_{i,p} \geq \beta_{k,1})$ , can be given by

$$(27) \quad \beta_{i,p} | \beta_{k,1} \geq e_i^{-1} (c_{n_i} s_{p,i}) (c_{n_k} s_{1,k})^{-1} n_k (n_i - p + 1)^{-1}, \quad i=1 \text{ or } 2 \text{ or } \dots$$

where  $e_1, \dots, e_{k-1}$  are to be determined from

$$(28) \quad P_{\chi} (F_{n_i - p + 1, n_k} \leq e_i ; i=1, 2, \dots, k-1) = 1 - \alpha$$

The values of  $d_i$  and  $e_i$ ,  $(i=1, 2, \dots, k-1)$  can be determined from Nair's tables [5, p.164] in some cases. The result similar to (2.3) and (3.3) can be written down for this case too, but we are not giving it, because it is very straightforward.

(4.3) Let us consider

$$(29) \quad 1 - \alpha = P_{\chi} (F_{n_i - p + 1, n_k} \leq e_i ; i=1, 2, \dots, k-1) P_{\chi} (F_{n_i, n_k - p + 1} \geq d_i ; i=1, 2, \dots, k-1)$$

$$= P_{\chi} \left[ n_k \mid \{ s_{i,1}^{*pp} s_{k,1}^{*pp} (n_i - p + 1) e_i \} \leq \beta_{i,p} | \beta_{k,1}, i=1, \dots, k-1 \right]$$

$$P_{\chi} \left[ \beta_{i,1} | \beta_{k,p} \leq s_{i,1}^{*pp} s_{k,1}^{*pp} (n_i - p + 1) d_i^{-1}, i=1, 2, \dots, k-1 \right]$$

$$= P_{\chi} \left[ \frac{n_k}{s_{i,1}^{*pp} s_{k,1}^{*pp} (n_i - p + 1) e_i} \leq \frac{\beta_{i,p} | \beta_{k,1}^*}{\beta_{k,1}} \frac{s_{i,1}^{*pp} (n_i - p + 1)}{n_i d_i} ; i=1, 2, \dots, k-1 \right]$$

Hence using lemma 1, we have with probability  $\geq (1 - \alpha)$ ,

$$(30) \quad \frac{n_k c_{n_i} s_{p,i}}{e_i (n_i - p + 1) c_{n_k} s_{p,k}} \leq \frac{\beta_{i,p}}{\beta_{k,1}} \leq \frac{\beta_{i,1}}{\beta_{k,p}} \leq \frac{(n_k - p + 1) c_{n_i} s_{1,i}}{d_i n_i c_{n_k} s_{p,k}} \quad \text{for } i=1, 2, \dots, k-1$$

where  $e_i$  and  $d_i$  ( $i=1, 2, \dots, k-1$ ) are given by (29).

We note that the confidence bounds given by (30) are shorter than those given by Gnanadesikan [2,3]. Our confidence bounds are not meant for deriving the confidence bounds on the partials. The confidence bounds given by (30) are <sup>in</sup> some sense better than those given by Anderson [1].

REFERENCES

- [1] Anderson, T. W. (1963). Some optimum confidence bounds for roots of determinantal equations. Mimeo series, Columbia University New York.
- [2] Gnanadesikan, R. (1959). Equality of more than two variances and more than two dispersion matrices against certain alternatives. Ann. Math. Stat., 30, 177-184.
- [3] Gnanadesikan, R. (1960). Correction to and comment on "Equality of more than two variances and of more than two dispersion matrices against certain alternatives", Ann. Math. Statist. 31 227-228.
- [4] Kshirsagar, A. M. (1959). Bartlett Decomposition and Wishart distribution. Ann. Math. Statist. 30, 239-241.
- [5] Pearson, E.S. and Hartley, H.O. (1958). Biometrika Tables for Statisticians, Vol. I. Second edition, Cambridge University press.
- [6] Roy, S.N. and Gnanadesikan, R. (1962). Two sample comparisons of dispersion matrices for alternatives of intermediate specificity. Ann. Math. Statist. 33, 432-437.
- [7] Roy, S.N. (1953). On a heuristic method of test construction and its use in multivariate Analysis. Ann. Math. Statist. 24, 220-228.

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344. Roberts, Charles D. An asymptotically optimal sequential design for comparing several experimental categories with a standard or control. 1963.
345. Novick, M. R. A Bayesian indifference procedure. 1963.
346. Johnson, N. L. Cumulative sum control charts for the folded normal distribution. 1963.
347. Potthoff, Richard F. On testing for independence of unbiased coin tosses lumped in groups too small to use  $\chi^2$ .
348. Novick, M. R. A Bayesian approach to the analysis of data for clinical trials. 1963.
349. Sethuraman, J. Some limit distributions connected with fixed interval analysis. 1963.
350. Sethuraman, J. Fixed interval analysis and fractile analysis. 1963.
351. Potthoff, Richard F. On the Johnson-Neyman technique and some extensions thereof. 1963.
352. Smith, Walter L. On the elementary renewal theorem for non-identically distributed variables. 1963.
353. Naor, P. and Yadin, M. Queueing systems with a removable service stations. 1963.
354. Page, E. S. On Monte Carlo methods in congestion problems—I. Searching for an optimum in discrete situations. February, 1963.
355. Page, E. S. On Monte Carlo methods in congestion problems—II. Simulation of queueing systems. February, 1963.
356. Page, E. S. Controlling the standard deviation by cusums and warning lines.
357. Page, E. S. A note on assignment problems. March, 1963.
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