

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

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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

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

We shall denote $ch_j \underline{A}$ the j -th maximum characteristic (max. ch.) root of $\underline{A}: p \times p$. i.e. $ch_1 \underline{A} \geq ch_2 \underline{A} \geq \dots \geq ch_p \underline{A}$. When $k = 1$, \underline{S}_1 , \underline{n}_1 and $\underline{\Sigma}_1$ will be denoted by \underline{S} , \underline{n} , and $\underline{\Sigma}$ respectively.

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

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

Then

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

Proof:- by definition

$$ch_p \underline{A} \leq \underline{\xi}' \underline{A} \underline{\xi} \mid \underline{\xi}' \underline{\xi} \leq ch_1 \underline{A}$$

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

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

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

Lemma 2. If $\underline{S}^* = (s^*_{ij})$ be distributed as $W(\underline{S}^* ; n, p, \underline{D}_\gamma)$ where $\underline{D}_\gamma = \text{diag.} (\gamma_1, \dots, \gamma_p)$, then (s^*_{ii} / γ_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 $\underline{\Sigma}$ is symmetric positive definite, we can find an orthogonal matrix $\underline{\Delta}$ such that $\underline{\Delta} \underline{\Sigma} \underline{\Delta}' = \underline{D}_\gamma$, a diagonal matrix with $\gamma_1 \geq \dots \geq \gamma_p > 0$. Then the distribution of $\underline{S}^* = \underline{\Delta} \underline{S} \underline{\Delta}'$ is $W(\underline{S}^* ; n, p, \underline{D}_\gamma)$.

(2.1) Let us obtain the confidence bound on γ_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}^*/γ_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

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

Since s_{11}^* depends on the nuisance parameters of Σ , we cannot carry out the exact test given by (1), but we shall obtain the confidence bound on γ_1 with probability $\geq (1-\alpha)$. We note from (2) that

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

but by lemma 1, $ch_1 \tilde{S} = ch_1 \tilde{S}^* \geq s_{11}^*$. Hence (3) gives us the following confidence bound on γ_1

$$(4) \quad \gamma_1 \leq c^{-1} ch_1 \tilde{S}$$

with probability $\geq (1-\alpha)$, and it will ^{be} less than 1 when $ch_1 \tilde{S} \leq c$.

(2.2) Applying arguments similar to (2.1), we obtain the confidence bounds on γ_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

$$(5) \quad \gamma_p \geq b^{-1} ch_p \tilde{S}$$

where b is determined from

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

(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

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

The test procedure (8) cannot be carried out in practice. Hence, noting lemma 1, we find the simultaneous confidence bounds on γ_1 or γ_p with probability $\geq (1 - \alpha)$ as

$$(10) \quad \gamma_1 \leq c_1^{-1} c_{n_1} \bar{s} \quad \text{or} \quad \gamma_p \geq b_1^{-1} c_{n_p} \bar{s},$$

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 Σ can be derived from the following considerations.

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

and so using lemma 1, we have

$$(11) \quad 1 - \alpha \leq P_{\Sigma} [b_1^{-1} c_{n_p} \bar{s} \leq \text{all ch}(\Sigma) \leq c_1^{-1} c_{n_1} \bar{s}].$$

3. Simultaneous confidence bounds on $c_{n_p} \bar{s}, \Sigma_2^{-1}$ or $c_{n_1} \bar{s}_1 \Sigma_2^{-1}$.

Lemma 2. Let A and B be two $p \times p$ symmetric positive definite matrices, and let δ_1 and δ_p be two unit ch. vectors corresponding to the max. and minimum (min.) ch. roots of A . If $A = (a_{ij})$, $A^{-1} = (a^{ij})$, $t_i = \delta_i' B \delta_i$ and $t^i = \delta_i' B^{-1} \delta_i$

for $i=1, p$, then

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

and

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

Proof. Let $\Delta : p \times p$ be an orthogonal matrix whose first column is $\delta_{\sim 1}$ and the last column is $\delta_{\sim p}$ such that $\Delta' A \Delta = D_{\sim 2}$ 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

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

Moreover, by lemma 1,

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

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

Since $\Sigma_{\sim i}$ ($i=1, 2$) are symmetric positive definite, there exists a non-singular matrix C_{\sim} such that

$$(12) \quad C_{\sim 1} \Sigma_{\sim 1} C_{\sim 1}' = D_{\sim 1}, \text{ a diagonal matrix, and } C_{\sim 2} \Sigma_{\sim 2} C_{\sim 2}' = I_{\sim 2},$$

where $\gamma_1 \geq \gamma_2 \geq \dots \geq \gamma_p > 0$, $\gamma_i = \text{ch}_i \Sigma_{\sim 1} \Sigma_{\sim 2}^{-1}$. Let $S_{\sim 1}^* = C_{\sim 1} \Sigma_{\sim 1} C_{\sim 1}'$ and $S_{\sim 2}^* = C_{\sim 2} \Sigma_{\sim 2} C_{\sim 2}'$.

Then $S_{\sim 1}^*$ and $S_{\sim 2}^*$ are independently distributed as

$$(13) \quad W(S_{\sim 1}^*; n_1, p, D_{\sim 1}, \gamma) \text{ and } W(S_{\sim 2}^*; n_2, p, I_{\sim 2}).$$

Lemma 4. Let $\epsilon_{\sim 1}$ and $\epsilon_{\sim p}$ be the unit ch. vectors corresponding to the max. and min. ch. root of $S_{\sim 1}^* = (s_{ij}^*)$, and let $S_{\sim 1}^*$ and $S_{\sim 2}^*$ be distributed as (13).

Then $s_{11}^* (\epsilon_{\sim 1}' S_{\sim 2}^{*-1} \epsilon_{\sim 1}) (n_2 - p + 1) / n_1 \gamma_1$ and $n_2 / \{ s_{pp}^{*pp} (\epsilon_{\sim p}' S_{\sim 2}^* \epsilon_{\sim p}) \gamma_p (n_1 - p + 1) \}$,

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

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

Proof. Let ξ be an orthogonal matrix, whose first and the last columns are ξ_1 and ξ_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}^* / γ_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.

(3.1) Let us obtain the confidence bound on $\gamma_1 = \text{ch}_{11} \Sigma_1 \Sigma_2^{-1}$ 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 Σ_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

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

We note that even though we cannot carry out the test procedure (14), but we can make a confidence statement on γ_1 with probability $\geq (1 - \alpha)$. By lemma 4, we have

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

Thus, (16) gives the confidence bound on γ_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 γ_1 or γ_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.

$$\begin{aligned} (1-\alpha) &= P_2 (F_{n_1-p+1, n_2} \leq b_1) P_2 (F_{n_1, n_2-p+1} \geq c_1) \\ &= P_2 [n_2 / ((n_1-p+1) \gamma_p s^{*pp}(\xi_p^* S_p^* \xi_p)) \leq b_1] P_2 [s_{11}^* (\xi_1^* S_1^{*-1} \xi_1) (n_2-p+1) / n_1 \gamma_1 \geq c_1] \\ &= P_2 [n_2 / ((n_1-p+1) s^{*pp}(\xi_p^* S_p^* \xi_p) b_1) \leq \text{all ch}(\Sigma_1 \Sigma_2^{-1}) \leq s_{11}^* (\xi_1^* S_1^{*-1} \xi_1) (n_2-p+1) / n_1 c_1] \end{aligned}$$

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

$$(21) \quad (c_{n_1} S_{p-1} S_2^{-1}) n_2 / (n_2 - p + 1) b_1 \leq \text{all } c_n (\Sigma_1 \Sigma_2^{-1}) \leq (c_{n_1} S_{p-1} S_2^{-1}) (n_2 - p + 1) / n_1 c_1,$$

where b_1 and c_1 are given by (20).

4. Simultaneous confidence bounds on $c_{n_t} \Sigma_i$ ($i=1, 2, \dots, k; t=1, p$).

Since Σ_i is symmetric positive definite, there exists an orthogonal matrix Δ_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 $W(S_i^*; n_i, p D_{i,\beta})$, ($i=1, 2, \dots, k$).

(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

$$(23) \quad P_{H_0} (F_{n_i, n_k - p + 1} \geq d_i') = 1 - \alpha.$$

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

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

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

$$(25) \quad \beta_{i,1} | \beta_{k,p} \leq d_i'^{-1} (c_{n_1} S_i) (c_{n_p} S_k)^{-1} (n_k - p + 1) n_i^{-1}, \quad i=1, \dots, k-1$$

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

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

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

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

$$= P_{\chi} \left[n_k | \{ s_{i,1}^{*pp} s_{k,11}^* (n_i - p + 1) e_i \} \leq \beta_{i,p} | \beta_{k,1}, i=1, \dots, k-1 \right]$$

$$P_{\chi} \left[\beta_{i,1} | \beta_{k,p} \leq s_{i,11}^* s_{k,11}^{*pp} (n_i - p + 1) (n_i d_i)^{-1}, i=1, 2, \dots, k-1 \right]$$

$$= P_{\chi} \left[\frac{n_k}{s_{i,11}^{*pp} s_{k,11}^* (n_i - p + 1) e_i} \leq \frac{\beta_{i,p}}{\beta_{k,1}} \leq \frac{\beta_{i,1} | \beta_{k,p}}{\beta_{k,1}} \leq \frac{s_{i,11}^{*pp} (n_k - p + 1)}{n_i d_i}, i=1, 2, \dots, k-1 \right]$$

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

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

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 ⁱⁿ some sense [^] better than those given by Anderson [1].

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