

Further contributions to some inequalities for normal
distributions and their applications to simultaneous confidence bounds

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1. Introduction and summary. Let $\underline{x} = (x_1, \dots, x_p)'$ be distributed as multivariate normal with zero means and covariance matrix $V(\underline{x})$ and this will be denoted by $\underline{x} \sim N(0, V(\underline{x}))$. Dunn's conjecture [3], namely,

$$(1) \quad P[|x_i| \leq c_i, i=1, 2, \dots, p] \geq \prod_{i=1}^p P[|x_i| \leq c_i]$$

was established by Khatri [4], Sidak [6] and Scott [5] by using different approaches. Moreover, Khatri [4] conjectured that

$$(2) \quad P[|x_i| \geq c_i, i=1, 2, \dots, p] \geq \prod_{i=1}^p P[|x_i| \geq c_i]$$

and the proof of (2) given by Scott [5] is incorrect. The purpose of this paper is to generalize (1) and (2) in the case of convex and symmetric regions about the origin. The generalized results are mentioned as under:

Let $\underline{x}' = (y'_1, y'_2, \dots, y'_q)$ where $y'_i = (x_{p_1+\dots+p_{i-1}+1}, \dots, x_{p_1+\dots+p_i})$, $i=1, 2, \dots, q$ with $\sum_{i=1}^q p_i = p$. Moreover, let $\mathcal{D}_i(y_i)$ be convex and symmetric region in y_i about the origin in p_i -dimensional space with $-\infty < x_j < \infty$, $j = 1, 2, \dots, p_1+p_2+\dots+p_{i-1}, p_1+p_2+\dots+p_i+1, \dots, p$. Let $\overline{\mathcal{D}_i(y_i)}$ be the complementary region of $\mathcal{D}_i(y_i)$. Then, we have

$$(3) \quad P\left(\bigcap_{i=1}^q \mathcal{D}_i(y_i)\right) \geq \prod_{i=1}^q P(\mathcal{D}_i(y_i))$$

and

$$(4) \quad P\left(\bigcap_{i=1}^q \overline{\mathcal{D}_i(y_i)}\right) \geq \prod_{i=1}^q P(\overline{\mathcal{D}_i(y_i)}) .$$

Some applications of these results are given on simultaneous confidence intervals. All the results mentioned by Katri [4] are now valid omitting the structure ℓ .

2. Inequalities for multivariate normal distributions.

The following lemmas will be used in establishing (3) and (4).

Lemma 1. Let $\underline{x} \sim N(\underline{0}, V(\underline{x}))$, $\underline{x}^{(2)} = (x_2, x_3, \dots, x_p)$ and let $\mathcal{D}_1(x_1)$ and $\mathcal{D}_2(\underline{x}^{(2)})$ be two convex and symmetric regions in x_1 and $\underline{x}^{(2)}$ respectively about the origin in p -dimensional space containing axes due to other variates.

Then,

$$P(\mathcal{D}_1(x_1) \cap \mathcal{D}_2(\underline{x}^{(2)})) \geq P(\mathcal{D}_1(x_1)) P(\mathcal{D}_2(\underline{x}^{(2)})).$$

For proof, refer Katri [4].

Lemma 2. Let $\underline{x}: p \times 1 \sim N(\underline{0}, V(\underline{x}))$ and $\underline{z}: p \times 1 \sim N(\underline{0}, V(\underline{z}))$. Then, if $V(\underline{x}) - V(\underline{z})$ is positive semi-definite,

$$P(\mathcal{D}(\underline{z})) \geq P(\mathcal{D}(\underline{x}))$$

where $\mathcal{D}(\underline{w})$ is a convex and symmetric region in \underline{w} about the origin.

For proof, refer Anderson [1].

Theorem 1. Let $\underline{x} \sim N(\underline{0}, V(\underline{x}))$, $\underline{x}' = (\underline{y}'_1, \dots, \underline{y}'_q)$, $\underline{y}'_i = (x_{p_1+\dots+p_{i-1}}, \dots, x_{p_1+\dots+p_i})$

and $\mathcal{D}_i(\underline{y}_i)$ be convex and symmetric region in \underline{y}_i about the origin in p -dimensional space containing axes due to other variates, for $i=1, 2, \dots, q$.

Then

$$P\left[\bigcap_{i=1}^q \mathcal{D}_i(\underline{y}_i)\right] \geq P(\mathcal{D}_1(\underline{y}_1)) P\left(\bigcap_{i=2}^q \mathcal{D}_i(\underline{y}_i)\right) \geq \prod_{i=1}^q P(\mathcal{D}_i(\underline{y}_i)).$$

Proof. When any $(q-1)$ values of $p_i, i=1,2,\dots,q$ are at the most one, then theorem 1 is established by Knatri [4] or theorem 1 is equivalent to lemma 1. Here, we assume that $p_i > 1, i = 1,2,\dots,q$. First of all, we shall consider the case when $V(\underline{x})$ is positive definite. Without loss of generality, we can write $V(\underline{x}) = \underline{A}\underline{A}'$ where $\underline{A} = (A_{ij})$ is nonsingular, $A_{ii} = 0$ for $i > i', i, i' = 1,2,\dots,q$. Let $\underline{A}^{-1}\underline{x} = \underline{z}$ and $\underline{z}' = (\underline{z}'_1, \dots, \underline{z}'_q)$ with $\underline{z}'_i = (z_{p_1+\dots+p_{i-1}+1}, \dots, z_{p_1+\dots+p_i})$. Then, it is easy to see that

$$y_i = \sum_{j=i}^q A_{ij} w_j \quad \text{for } i = 1,2,\dots,q$$

and $\underline{z} \sim N(\underline{0}, \underline{I}_p)$ or $w_i \sim IN(\underline{0}, \underline{I}_{p_i})$, $i = 1,2,\dots,q$. It is easy to see that theorem 1 will be established for $V(\underline{x})$ to be positive definite if we can establish

$$(5) \quad P \left[\bigcap_{i=1}^q \mathcal{D}_i \left(\sum_{j=i}^q A_{ij} w_j \right) \mid \underline{z} \in Q \right] >$$

$$P \left[\mathcal{D}_1(\underline{y}_1) \right] P \left[\bigcap_{j=2}^q \mathcal{D}_i \left(\sum_{j=i}^q A_{ij} w_j \mid \underline{z} \in Q \right) \right]$$

for every (p_1+1) -flat Q containing $(z_1, z_2, \dots, z_{p_1})$ -axes.

Let us take such a (p_1+1) -flat Q and let us suppose that this is determined by the set of linearly independent equations given by

$$\sum_{j=1}^{p-p_1} l_{kj} z_{j+p_1} = 0 \quad \text{for } k = 1,2,\dots,p-p_1-1$$

where without loss of generality, take $\sum_{j=1}^{p-p_1} l_{kj}^2 = 1$ and $\sum_{j=1}^{p-p_1} l_{kj} l_{k'j} = 0$

for $k \neq k'$. Let $\underline{L}_1 = (l_{kj}) : (p-p_1-1) \times (p-p_1)$. Then, we can complete \underline{L}_1 by a vector \underline{l} such that $L' = \begin{pmatrix} \underline{l} \\ \underline{L}_1 \end{pmatrix}$ is an orthogonal matrix. Now use the transformation

$$L(\underline{w}'_2, \dots, \underline{w}'_q)' = (v_1, v_2, \dots, v_{p-p_1})'$$

Then it is obvious that the (p_1+1) -flat Q will have the coordinate system given by $(\underline{w}'_1, v_1, v_i = 0 \text{ for } i = 2, \dots, p-p_1)$ and $v_1 \sim N(0,1)$ and $\underline{w}'_1 \sim N(0, I_{p_1})$ and they are independently distributed. Hence, using this system of coordinates in the left side of (5), we get

$$(5) \quad I(Q) = P\left[\bigcap_{i=1}^q \mathcal{N}_i\left(\sum_{j=i}^q A_{ij} \underline{w}_j\right) \mid Z \in Q\right] = P\left[\mathcal{N}_1(A_{11}\underline{w}_1 + \delta_1 v_1) \cap \left(\bigcap_{j=2}^q \mathcal{N}_j(\delta_j v_1)\right)\right]$$

where if $\underline{l}' = (\underline{l}'_2, \underline{l}'_3, \dots, \underline{l}'_q)$ with $\underline{l}'_i : p_i \times 1$, $\delta_i = \sum_{j=i}^q A_{ij} \underline{l}_j$ for $i = 2, 3, \dots, q$ and $\delta_1 = \sum_{j=2}^q A_{1j} \underline{l}_j$. Since $\delta_i, i = 2, \dots, q$ are fixed vector and $\mathcal{N}_i(\delta_i v_1), i = 2, 3, \dots, q$ are convex and symmetric in $\delta_i v_1$ about the origin and hence $\mathcal{N}(v_1) = \bigcap_{i=2}^q \mathcal{N}_i(v_1)$ is convex and symmetric in v_1

about the origin. Then, using this in (6) and then lemma 1, we get

$$(7) \quad I(Q) \geq P\left[\mathcal{N}_1(A_{11}\underline{w}_1 + \delta_1 v_1)\right] P(\mathcal{N}(v_1)) = \prod_{i=2}^q \mathcal{N}_i(\delta_i v_1)$$

Note that

$$(8) \quad P\left(\bigcap_{i=2}^q \mathcal{N}_i(\delta_i v_1)\right) = P\left(\bigcap_{i=2}^q \mathcal{N}_i\left(\sum_{j=i}^q A_{ij} \underline{w}_j\right) \mid Z \in Q\right)$$

and by using lemma 2,

$$(9) \quad P\left[\mathfrak{D}_1(A_{11}w_1 + \delta_1 v_1)\right] \geq P\left[\mathfrak{D}_1(y_1)\right]$$

$$\text{for } V(y_1) - V(A_{11}w_1 + \delta_1 v_1) = \sum_{j=1}^q A_{1j}A'_{1j} - (A_{11}A'_{11} + \delta_1 \delta'_1) =$$

$$\sum_{j=2}^q A_{1j}(\mathbb{I}_{p_j} - \ell_j \ell'_j) A'_{1j} \text{ is positive semi-definite.}$$

Using (8) and (9) in (7), we get (5). Thus, theorem 1 is established when $V(\underline{x})$ is nonsingular.

Let $V(\underline{x})$ be positive semi-definite. Let $\underline{u}: p \times 1 \sim N(0, n \mathbb{I}_p)$ and let \underline{u} and \underline{x} be independently distributed. Then $\underline{x} + \underline{u} = \underline{t} \sim N(0, n \mathbb{I}_p + V(\underline{x}))$ and $V(\underline{x}) + n \mathbb{I}_p$ is positive definite. Hence from the result for positive definite covariance matrix, we get

$$(10) \quad P\left[\bigcap_{i=1}^q \mathfrak{D}_i(t_i)\right] \geq \prod_{i=1}^q P\left[\mathfrak{D}_i(t_i)\right].$$

Taking limits as $n \rightarrow 0^+$, we get the result for the singular case, for

$$\lim_{n \rightarrow 0^+} P\left[\mathfrak{D}(t_i)\right] = P\left[\mathfrak{D}(y_i)\right]$$

if $\mathfrak{D}(y)$ is a convex and symmetric region in y about the origin.

Theorem 2. Under the notations of theorem 1, we have

$$P\left(\bigcap_{i=1}^q \overline{\mathfrak{D}_i(y_i)}\right) \geq P(\overline{\mathfrak{D}_1(y_1)}) P\left(\bigcap_{i=2}^q \overline{\mathfrak{D}_i(y_i)}\right) \geq \prod_{i=1}^q P(\overline{\mathfrak{D}_i(y_i)}),$$

where $\overline{\mathfrak{D}(w)}$ is the complement of $\mathfrak{D}(w)$.

Proof. Let us consider the case when $V(\underline{x})$ is positive definite and we proceed in the same manner as in theorem 1 in considering

$$P\left[\mathfrak{D}_1(\underline{y}_1) \cap \left(\bigcup_{i=2}^q \mathfrak{D}_i(\underline{y}_i)\right)\right].$$

Using the same arguments as those in theorem 1, we get

$$(11) \quad P\left[\mathfrak{D}_1\left(\sum_{j=1}^q A_{1j} w_j\right) \cap \left(\bigcup_{i=2}^q \mathfrak{D}_i\left(\sum_{j=i}^q A_{ij} w_j\right)\right) \mid \underline{Z} \in \mathcal{Q}\right]$$

$$= P\left[\mathfrak{D}_1(A_{11} w_1 + \delta_1 v_1) \cap \left(\bigcup_{i=2}^q \mathfrak{D}_i'(v_1)\right)\right].$$

Now $\bigcup_{i=2}^q \mathfrak{D}_i'(v_1) \iff |v_1| \leq \alpha$ for some $\alpha \geq 0$. Hence, using this in

(11) and using theorem 1, we get

$$(12) \quad P\left[\mathfrak{D}_1\left(\sum_{j=1}^q A_{1j} w_j\right) \cap \left(\bigcup_{i=2}^q \mathfrak{D}_i\left(\sum_{j=i}^q A_{ij} w_j\right)\right) \mid \underline{Z} \in \mathcal{Q}\right]$$

$$\geq P\left[\mathfrak{D}_1(A_{11} w_1 + \delta_1 v_1)\right] P\left[|v_1| \leq \alpha\right]$$

and using $P\left[|v_1| \leq \alpha\right] = P\left[\bigcup_{i=2}^q \mathfrak{D}_i'(v_1)\right] = P\left[\bigcup_{i=2}^q \mathfrak{D}_i\left(\sum_{j=i}^q A_{ij} w_j\right) \mid \underline{Z} \in \mathcal{Q}\right]$ and

$P\left[\mathfrak{D}_1(A_{11} w_1 + \delta_1 v_1)\right] \geq P(\mathfrak{D}_1(\underline{y}_1))$, we get

$$(13) \quad P\left[\mathfrak{D}_1\left(\sum_{j=1}^q A_{1j} w_j\right) \cap \left(\bigcup_{i=2}^q \mathfrak{D}_i\left(\sum_{j=i}^q A_{ij} w_j\right)\right) \mid \underline{Z} \in \mathcal{Q}\right]$$

$$\geq P(\mathfrak{D}_1(\underline{y}_1)) P\left[\bigcup_{i=2}^q \mathfrak{D}_i\left(\sum_{j=i}^q A_{ij} w_j\right) \mid \underline{Z} \in \mathcal{Q}\right].$$

Then (13) gives us

$$(14) \quad P\left[\mathfrak{D}_1(\underline{y}_1) \cap \left(\bigcup_{i=2}^q \mathfrak{D}_i(\underline{y}_i)\right)\right] \geq P(\mathfrak{D}_1(\underline{y}_1)) P\left(\bigcup_{i=2}^q \mathfrak{D}_i(\underline{y}_i)\right).$$

We note that if \mathcal{R}_1 and \mathcal{R}_2 be two regions, then

$$P(\mathcal{R}_1) = P(\mathcal{R}_1 \cap \mathcal{R}_2) + P(\mathcal{R}_1 \cap \bar{\mathcal{R}}_2) .$$

Moreover, we have $\overline{\left\{ \bigcup_{i=2}^q \mathcal{D}_i(\underline{y}_i) \right\}} = \bigcap_{i=2}^q \overline{\mathcal{D}_i(\underline{y}_i)}$. Using these in (14), we get

$$(15) \quad P\left[\mathcal{D}_1(\underline{y}_1) \cap \left(\bigcap_{i=2}^q \overline{\mathcal{D}_i(\underline{y}_i)}\right)\right] \leq P(\mathcal{D}_1(\underline{y}_1)) P\left(\bigcap_{i=2}^q \overline{\mathcal{D}_i(\underline{y}_i)}\right)$$

and this implies

$$(16) \quad P\left(\bigcap_{i=1}^q \overline{\mathcal{D}_i(\underline{y}_i)}\right) \geq P(\mathcal{D}_1(\underline{y}_1)) P\left(\bigcap_{i=2}^q \overline{\mathcal{D}_i(\underline{y}_i)}\right) .$$

Thus, theorem 2 is proved when $V(\underline{x})$ is positive definite. When $V(\underline{x})$ is singular, we can argue in the same manner as in theorem 1. This completes the proof of theorem 2.

Corollary 1. Let $\underline{x}_j \sim N(\underline{0}, V(\underline{x}_j))$, $j = 1, 2, \dots, n$ and let them be independent.

Let $\mathcal{D}_i = \mathcal{D}_i(\underline{y}_{i1}, \underline{y}_{i2}, \dots, \underline{y}_{in})$ be convex and separately symmetric in $\underline{y}_{i1}, \underline{y}_{i2}, \dots, \underline{y}_{in}$ about the origin for $i = 1, 2, \dots, q$ and $\bar{\mathcal{D}}_i$ be the complement of \mathcal{D}_i . Then

$$P\left(\bigcap_{i=1}^q \mathcal{D}_i\right) \geq \prod_{i=1}^q P(\mathcal{D}_i) \quad \text{and} \quad P\left(\bigcap_{i=1}^q \bar{\mathcal{D}}_i\right) \geq \prod_{i=1}^q P(\bar{\mathcal{D}}_i) .$$

(For the definition of separately symmetric, see Knaatri [4].)

Proof. We shall only indicate the proof for one case as under:

Let $\underline{w}_{ij} \sim N(\underline{0}, V(\underline{y}_{ij}))$, $j = 1, 2, \dots, n$ and $i = 1, 2, \dots, q$ and let them be independent and independent of $\underline{x}_1, \underline{x}_2, \dots, \underline{x}_n$. By theorem 1, it is easy to see that

$$(17) \quad \mathbb{P} \left[\bigcap_{i=1}^q \mathcal{D}_i(\underline{y}_{i1}, \dots, \underline{y}_{in}) \mid \underline{x}_1, \dots, \underline{x}_{n-1} \right] \\ \geq \mathbb{P} \left[\bigcap_{i=1}^q \mathcal{D}_i(\underline{y}_{i1}, \dots, \underline{y}_{in-1}) \mid \underline{x}_1, \dots, \underline{x}_{n-1} \right]$$

because $\mathcal{D}_i(\underline{y}_{i1}, \dots, \underline{y}_{in})$ is convex and symmetric in \underline{y}_{in} about the origin when $\underline{y}_{i1}, \dots, \underline{y}_{in-1}$ are fixed. From (17), we get

$$(18) \quad \mathbb{P} \left[\bigcap_{i=1}^q \mathcal{D}_i(\underline{y}_{i1}, \dots, \underline{y}_{in}) \right] \geq \mathbb{P} \left[\bigcap_{i=1}^q \mathcal{D}_i(\underline{y}_{i1}, \dots, \underline{y}_{in-1}, \underline{w}_{in}) \right].$$

Proceeding in the same manner for $\underline{x}_{n-1}, \underline{x}_{n-2}, \dots, \underline{x}_1$, we get the final result as

$$(19) \quad \mathbb{P} \left[\bigcap_{i=1}^q \mathcal{D}_i \right] \geq \mathbb{P} \left[\bigcap_{i=1}^q \mathcal{D}_i(\underline{w}_{i1}, \dots, \underline{w}_{in}) \right] = \prod_{i=1}^q \mathbb{P} \left[\mathcal{D}_i(\underline{w}_{i1}, \dots, \underline{w}_{in}) \right] \\ = \prod_{i=1}^q P(\mathcal{D}_i).$$

This proves the first part of corollary 1. The second part can be proved in the same manner.

Corollary 2. Let $\underline{x}_j \sim N(0, V(\underline{x}_j))$, $j=1,2,\dots,n$ and be independent. Let us suppose that $V(\underline{x}_j) = (A_{ii',j})$, $A_{ii',j} = \sigma_{i,j}^2 I_{\alpha_i}$ for $\alpha_i = 1,2,\dots,r$

and $\sum_{\alpha=\alpha_1+\dots+\alpha_{i-1}+1}^{\alpha_1+\dots+\alpha_i} x_{\alpha,j}^2$ for $i = 1,2,\dots,r$

$\underline{y}'_{t,j} = (Z_{r_1+\dots+t_{t-1}+1}, \dots, Z_{r_1+r_2+\dots+r_t})$, $t = 1,2,\dots,q$, $\sum_{t=1}^q r_t = r$ and

$\mathcal{D}_t = \mathcal{D}_t(\underline{y}_{t,1}, \underline{y}_{t,2}, \dots, \underline{y}_{t,n})$ about the origin for $t = 1,2,\dots,q$. Then,

$$P\left(\bigcap_{t=1}^q \mathcal{D}_t\right) \geq \prod_{t=1}^q P(\mathcal{D}_t) \quad \text{and} \quad P\left(\bigcap_{t=1}^q \bar{\mathcal{D}}_t\right) \geq \prod_{t=1}^q P(\bar{\mathcal{D}}_t).$$

This follows from corollary 1.

Note: In corollaries 1 and 2, if some observations are missing on \underline{y}_1 or \underline{y}_2, \dots or \underline{y}_q , we have to omit these from the convex and symmetric regions \mathcal{D}_i , $i = 1, 2, \dots, q$. e.g. Suppose on \underline{y}_1 , the only observations available are $\underline{y}_{1,j}$, $j = 1, 2, \dots, n_1$ ($n_1 < n$). Then, $\mathcal{D}_1 = \mathcal{D}_1(\underline{y}_{1,1}, \dots, \underline{y}_{1,n_1})$ is convex and symmetric region in $\underline{y}_{1,1}, \dots, \underline{y}_{1,n_1}$ about the origin.

3. Direct applications.

(3.1) Confidence bounds for means.

Let us suppose that $\underline{x}_j \sim N(\underline{\xi}, V(\underline{x}))$ for $j = 1, 2, \dots, n$ and let them be independent. Let us assume that $V(\underline{x}) = (A_{ii})$, $A_{ii} = \sigma_i^2 \frac{1}{\alpha_i}$, $i=1, 2, \dots, r$ and $\underline{\xi}' = (\underline{\xi}'_1, \underline{\xi}'_2, \dots, \underline{\xi}'_r)$, with $\underline{\xi}'_i: \alpha_i \times 1$.

Let $\underline{x}'_j = (\underline{y}'_{1,j}, \dots, \underline{y}'_{r,j})$, $\underline{y}'_i: \alpha_i \times 1$, $\bar{\underline{y}}_i = \sum_{j=1}^n \underline{y}_{i,j} / n$ and

$\mathcal{D}_{ii} = \sum_{j=1}^n \underline{y}'_{i,j} \underline{y}_{i,j} - n \bar{\underline{y}}_i' \bar{\underline{y}}_i$. Then, by corollary 2, it is easy to see that

$$(20) \quad \begin{aligned} P\left[\left(\bar{\underline{y}}_i - \underline{\xi}'_i\right)' \left(\bar{\underline{y}}_i - \underline{\xi}'_i\right) \leq c_i \mathcal{D}_{ii}, i = 1, 2, \dots, r\right] \\ \geq \prod_{i=1}^r P\left[\left(\bar{\underline{y}}_i - \underline{\xi}'_i\right)' \left(\bar{\underline{y}}_i - \underline{\xi}'_i\right) \leq c_i \mathcal{D}_{ii}\right] \end{aligned}$$

because $\bar{\underline{y}}_i$ and \mathcal{D}_{ii} are independently distributed and $\mathcal{D}_{ii} = \sum_{j=1}^{n-1} Z'_{i,j} Z_{i,j}$

with $(\bar{\underline{y}}_1', \dots, \bar{\underline{y}}_r')' \sim N(\underline{\xi}, V(\underline{x})/n)$ and $\underline{Z}_j = (Z'_{1,j}, \dots, Z'_{r,j})' \sim IN(0, V(\underline{x}))$.

Now it is easy to see that $n(n-1)(\bar{y}_i - \xi_i)'(\bar{y}_i - \xi_i)/\rho_{ii}$ is distributed as $F_{\alpha_i, (n-1)\alpha_i}$ with α_i and $(n-1)\alpha_i$ degrees of freedom for $i = 1, 2, \dots, r$. Hence, we can find c_1, c_2, \dots, c_r such that

$$(21) \quad \prod_{i=1}^r P\left[\frac{(\bar{y}_i - \xi_i)'(\bar{y}_i - \xi_i)}{\rho_{ii}} \leq c_i \rho_{ii}\right] = 1 - \alpha.$$

One choice of choosing c_1, c_2, \dots, c_r is to take

$$P\left[\frac{(\bar{y}_i - \xi_i)'(\bar{y}_i - \xi_i)}{\rho_{ii}} \leq c_i \rho_{ii}\right] = (1 - \alpha)^{1/r}.$$

Using (21) in (20), we can find simultaneous confidence bounds on ξ_i , $i = 1, 2, \dots, r$ with confidence greater than $(1 - \alpha)$ as

$$(22) \quad a_i' \bar{y}_i - \left\{c_i \rho_{ii} (a_i' a_i)\right\}^{1/2} \leq a_i' \xi_i \leq a_i' \bar{y}_i + \left\{c_i \rho_{ii} (a_i' a_i)\right\}^{1/2}$$

for all $i = 1, 2, \dots, r$ and for all non-null vectors $a_i: \alpha_i \times 1$, $i = 1, 2, \dots, r$.

(3.2) One sided confidence bounds on variances.

Let us suppose that $\underline{x} = (y_1', \dots, y_q')' \sim N(0, V(\underline{x}))$ and let us have n independent observations on \underline{x} . Out of these n observations, it is found that n_i observations are missing on y_i , $i = 1, 2, \dots, q$. Let $V(\underline{x}) = (A_{ii})$, and S_i , the sample sum of squares matrix due to available observations on y_i , $i = 1, 2, \dots, q$.

If $\mathfrak{D}_i = \mathfrak{D}_i \left[\text{ch}_{\max} (A_{ii}^{-1} S_i) \leq c_i \right]$, then \mathfrak{D}_i is section-wise convex and separately symmetric in available observations about the origin (see DasGupta, Mudholkar and Anderson [2]). Hence, by corollary 1, we get

$$(23) \quad P\left[\bigcap_{i=1}^q \mathfrak{D}_i\right] \geq \prod_{i=1}^q P(\mathfrak{D}_i) \quad \text{and} \quad P\left[\bigcap_{i=1}^q \bar{\mathfrak{D}}_i\right] \geq \prod_{i=1}^q P(\bar{\mathfrak{D}}_i).$$

In order to obtain the lower bounds on the parameters \underline{A}_{ii} , $i = 1, 2, \dots, q$, we use the first part of (22). Let us choose c_1, c_2, \dots, c_q such that

$$(24) \quad \prod_{i=1}^q P \left[\text{ch}_{\max} (\underline{A}_{ii}^{-1} S_i) \leq c_i \right] = 1 - \alpha .$$

Using (24) in the first part of (23), we get simultaneous lower bounds on \underline{A}_{ii} , $i = 1, 2, \dots, q$ with confidence greater than or equal to $(1 - \alpha)$ as

$$(25) \quad \underline{a}'_i \underline{A}_{ii} \underline{a}_i \geq \underline{a}'_i S_i \underline{a}_i / c_i, \quad i = 1, 2, \dots, q$$

for all non-null vectors $\underline{a}_1, \underline{a}_2, \dots, \underline{a}_q$.

Similarly, by choosing c'_i , $i = 1, 2, \dots, q$ from

$$(26) \quad \prod_{i=1}^q P \left[\text{ch}_{\max} (\underline{A}_{ii}^{-1} S_i) > c'_i \right] = 1 - \alpha ,$$

we find the simultaneous upper bounds on \underline{A}_{ii} , $i = 1, 2, \dots, q$ with confidence greater than or equal to $(1 - \alpha)$ as

$$(27) \quad \underline{a}'_i \underline{A}_{ii} \underline{a}_i \leq c'_i (\underline{a}'_i S_i \underline{a}_i), \quad i = 1, 2, \dots, q$$

for all non-null vectors \underline{a}_i , $i = 1, 2, \dots, q$.

By combining (25) and (27), we get the simultaneous confidence bounds on \underline{A}_{ii} , $i = 1, 2, \dots, q$ with confidence greater than or equal to $(1 - \alpha)^2$.

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