## Aplikace matematiky

## Somesh Las Gupta

On a probability inequality for multivariate normal distribution

Aplikace matematiky, Vol. 21 (1976), No. 1, 1-4
Persistent URL: http://dml.cz/dmlcz/103618

## Terms of use:

© Institute of Mathematics AS CR, 1976

Institute of Mathematics of the Czech Academy of Sciences provides access to digitized documents strictly for personal use. Each copy of any part of this document must contain these Terms of use.


This document has been digitized, optimized for electronic delivery and stamped with digital signature within the project DML-CZ: The Czech Digital Mathematics Library http://dml.cz

# ON A PROBABILITY INEQUALITY FOR MULTIVARIATE NORMAL DISTRIBUTION 

## Somesh Das Gupta*

(Received May 13, 1974)

1. Introduction. Let $P_{\lambda}$ denote the $p$-variate normal distribution $N_{p}\left(\mu, \Sigma_{\lambda}\right)$, where

$$
\begin{gather*}
\mu=\binom{\mu_{1}}{\mu_{2}} p_{1}, \quad \Sigma_{\lambda}=\left[\begin{array}{rr}
\Sigma_{11} & \lambda \Sigma_{12} \\
\lambda \Sigma_{21} & \Sigma_{22}
\end{array}\right] \begin{array}{l}
p_{1} \\
p_{2}
\end{array},  \tag{1}\\
p_{1}
\end{gather*} p_{2}, ~ \$
$$

$p_{1}+p_{2}=p, 0 \leqq \lambda \leqq 1$, and $\Sigma_{1}$ is positive-definite. Let $C_{1} \subset R^{p_{1}}, C_{2} \subset R^{p_{2}}$ be convex symmetric (about the respective origins) sets. Define

$$
\begin{equation*}
\pi(\lambda)=P_{\lambda}\left[C_{1} \times C_{2}\right] . \tag{2}
\end{equation*}
$$

Das Gupta et al. [1] have shown that

$$
\begin{equation*}
\pi(0) \leqq \pi(1) \tag{3}
\end{equation*}
$$

under the following assumptions: There exist vectors $b_{1} \in R^{p_{1}}, b_{2} \in R^{p_{2}}$ and a scalar $c$ such that
(i) $\mu_{i}=c b_{i}, i=1,2$
(ii) $\Sigma_{12}=b_{1} b_{2}^{\prime}$
(iii) $\Sigma_{i i}-b_{i} b_{i}^{\prime}(i=1,2)$ is positive definite.

The inequality (3) was proved by Khatri [2] when $\mu=0$. In this note, we shall show that

$$
\begin{equation*}
\pi(\lambda) \leqq \pi\left(\lambda^{*}\right) \tag{4}
\end{equation*}
$$

for $0 \leqq \lambda<\lambda^{*} \leqq 1$ when the above assumptions (i)-(iii) hold. For motivations and applications of the inequalities (3) and (4), one may see Das Gupta et al. [1]
*) On leave from the University of Minnesota.
and Khatri [2]. The inequality (4) was proved by Šidák [3] under the following stronger assumptions:
(a) $\mu=0$
(b) $R_{i i}=b_{i} b_{i}^{\prime}+\operatorname{diag}\left[I-b_{i} b_{i}^{\prime}\right], i=1,2$
$R_{12}=b_{1} b_{2}^{\prime}$
where $b_{1}: p_{1} \times 1, b_{2}: p_{2} \times 1$,

$$
\left[\begin{array}{ll}
R_{11} & R_{12} \\
R_{21} & R_{22}
\end{array}\right]
$$

is the correlation matrix corresponding to $\Sigma_{1}$, and for a square matrix $A$, $\operatorname{diag}(A)$ is defined to be the diagonal matrix obtained from $A$ by replacing all the off-diagonal elements of $A$ by 0 .

Our proof essentially uses the inequality (3) and some suitable prior distributions of the parameters. It can also be seen that Šidák's [3] proof may be modified, incorporating the assumptions in this note, to obtain (4).
2. Proof of the inequality (4). Consider $b_{1}: p_{1} \times 1, b_{2}: p_{2} \times 1$ and $c$ satisfying the assumptions (i)-(iii). Let $X_{1}: p_{1} \times 1, X_{2}: p_{2} \times 1$ and $\theta$ be distributed as the $(p+1)$-variate normal distribution, such that conditional joint distribution of $X_{1}$ and $X_{2}$, given $\theta$, is

$$
N_{p}\left[\left(\begin{array}{ll}
\theta & b_{1} \\
\theta & b_{2}
\end{array}\right), \Gamma\right]
$$

and $\theta \sim N(c, \lambda),(\lambda$ being the variance of $\theta)$. Then the unconditional joint distribution of $X_{1}$ and $X_{2}$ is

$$
N_{p}\left[\left(\begin{array}{ll}
c & b_{1} \\
c & b_{2}
\end{array}\right), \quad \Gamma+\lambda\left(\begin{array}{ll}
b_{1} b_{1}^{\prime} & b_{1} b_{2}^{\prime} \\
b_{2} b_{1}^{\prime} & b_{2} b_{2}^{\prime}
\end{array}\right)\right] .
$$

It can be seen that the joint distribution of $X_{1}$ and $X_{2}$ id $P_{\lambda^{*}}$ or $P_{\lambda}$, according as

$$
\begin{aligned}
& \Gamma=\Gamma_{1} \equiv\left(\begin{array}{cc}
\Sigma_{11}-\lambda b_{1} b_{1}^{\prime} & \left(\lambda^{*}-\lambda\right) b_{1} b_{2}^{\prime} \\
\left(\lambda^{*}-\lambda\right) b_{2} b_{1}^{\prime} & \Sigma_{22}-\lambda b_{2} b_{2}^{\prime}
\end{array}\right) . \\
& \Gamma=\Gamma_{0} \equiv\left(\begin{array}{cc}
\Sigma_{11}-\lambda b_{1} b_{1}^{\prime} & 0 \\
0 & \Sigma_{22}-\lambda b_{2} b_{2}^{\prime}
\end{array}\right),
\end{aligned}
$$

where $0 \leqq \lambda<\lambda^{*} \leqq 1$. Applying the inequality (3) of Das Gupta et. al. [1] after verifying their assumptions, we get

$$
\begin{equation*}
P\left[X_{1} \in C_{1}, X_{2} \in C_{2} \mid \theta, \Gamma=\Gamma_{1}\right] \geqq P\left[X_{1} \in C_{1}, X_{2} \in C_{2} \mid \theta, \Gamma=\Gamma_{0}\right] \tag{5}
\end{equation*}
$$

Taking expectations of both sides of the above inequality (5) with respect to $\theta$, we get

$$
\pi\left(\lambda^{*}\right) \geqq \pi(\lambda) .
$$

Note 1. If $\mu_{1}=0, \mu_{2}=0$, rank $\left(\Sigma_{12}\right)=1$, there exist vectors $b_{1}, b_{2}$ satisfying (ii) and (iii). To satisfy (i), take $c=0$. To see this, note that there exist nonsingular matrices $A_{1}$ and $A_{2}$ such that

$$
A_{1} \Sigma_{11} A_{1}^{\prime}=I_{p_{1}}, \quad A_{2} \Sigma_{22} A_{2}^{\prime}=I_{p_{2}},
$$

and

$$
A_{1} \Sigma_{12} A_{2}^{\prime}=\left(\begin{array}{cccc}
\varrho & 0 & \ldots & 0 \\
0 & 0 & \ldots & 0 \\
\vdots & \vdots & & \vdots \\
0 & 0 & \ldots & 0
\end{array}\right): p_{1} \times p_{2}
$$

where $0 \leqq \varrho<1$.
Define

$$
b_{i}=A_{i}^{-1}[\sqrt{ }(\varrho) 0 \ldots 0]^{\prime}: p_{i} \times 1, i=1,2 .
$$

Note 2. Suppose $\mu_{1} \neq 0, \mu_{2} \neq 0$. Assume the following: There exists a positive scalar $k$ such that
(ii') $\Sigma_{12}=k \mu_{1} \mu_{2}^{\prime}$
(iii') $k^{-1}>\max \left(\mu_{1}^{\prime} \Sigma_{11}^{-1} \mu_{1}, \mu_{2}^{\prime} \Sigma_{22}^{-1} \mu_{2}\right)$.
We shall show that there exist $b_{1}, b_{2}$ and $c$ satisfying (i)-(iii). There exist orthogonal matrices $L_{1}$ and $L_{2}$ such that

$$
\mu_{i}^{\prime}=\left(\delta_{i} 0 \ldots 0\right) L_{i} \Sigma_{i i}^{1 / 2}, \quad i=1,2
$$

where

$$
\delta_{i}=\left(\mu_{i}^{\prime} \Sigma_{i i}^{-1} \mu_{i}\right)^{1 / 2} .
$$

Define

$$
c=k^{-1 / 2}, \quad b_{i}=\mu_{i} / c, \quad i=1,2
$$

Note 3. When $\Sigma_{i i}(i=1,2)$ is p. d., rank $\left(\Sigma_{12}\right)=1$, but $\Sigma$ is p.s.d., the above proof is also valid for showing

$$
\pi(\lambda) \leqq \pi\left(\lambda^{*}\right)
$$

where $0 \leqq \lambda<\lambda^{*}<1$. In that case we need the following assumption:

$$
\begin{equation*}
\Sigma_{i i}-\lambda^{*} b_{i} b_{i}^{\prime} \quad(i=1,2,) \text { is p.d. } \tag{iiia}
\end{equation*}
$$

instead of the assumption (iii). Correspondingly the assumption (iii') in Note 2 can be changed. However the proof is no longer tenable for showing $\pi(\lambda) \leqq \pi(1), 0 \leqq \lambda<1$ when $\Sigma$ is p.s.d. subject to the assumptions made in the beginning of Note 3 . This
may apparently follow from the result of Das Gupta et. al. [1] who claimed to prove (3) under the assumption: $\Sigma_{i i}-b_{i} b_{i}^{\prime}$ is p.s.d. $(i=1,2)$, instead of (iii); however their proof is not complete.

Acknowledgment. I am thankful to Professor Z. Šidák for bringing his paper to my attention and going through the proof in this paper.

## References

[1] Das Gupta, S., Eaton, M. L., Olkin, I., Perlman, M. D., Savage, L. J., and Sobel, M.: Inequalities on the probability content of convex regions for elliptically contoured distributions. Proc. Sixth Berk. Symp. on Math. Stat. and Prob. Vol. II, (1972), 241-265.
[2] Khatri, C. G.: On certain inequalities for normal distributions and their applications to simultaneous confidence bounds. Ann. Math. Statist. 38, (1967), 1853-1867.
[3] Šidák, Z.: On probabilities in certain multivariate distributions: their dependence on correlations. Aplikace Matematiky 18, (1973), 128-135.

Souhrn

## O JEDNÉ NEROVNOSTI PRO PRAVDĚPODOBNOSTI V MNOHOROZMĚRNÉM NORMÁLNÍM ROZLOŽENÍ

## Somesh Das Gupta

Necht̛ $P_{\lambda}$ označuje $p$-rozměrné normální rozložení $N_{p}\left(\mu, \Sigma_{\lambda}\right)$, kde

$$
\Sigma_{\lambda}=\left[\begin{array}{cc}
\Sigma_{11} & \lambda \Sigma_{12} \\
\lambda \Sigma_{21} & \Sigma_{22}
\end{array}\right]
$$

je rozdělena na bloky s $p_{1}, p_{2}$ řádky a sloupci, přičemž $p_{1}+p_{2}=p, 0 \leqq \lambda \leqq 1$, a $\Sigma_{1}$ je positivně definitní. Budtež $C_{1} \subset R^{p_{1}}, C_{2} \subset R^{p_{2}}$ konvexní symetrické množiny. V článku je za určitých předpokladủ o $\mu$ a $\Sigma_{1}$ dokázáno, že pro $0 \leqq \lambda<\lambda^{*} \leqq 1$ je $P_{\lambda}\left[C_{1} \times C_{2}\right] \leqq P_{\lambda *}\left[C_{1} \times C_{2}\right]$, což zobecňuje dříivéjší výsledky Das Gupty aj. [1], Khatriho [2] a Šidáka [3].

Author's address: Professor Somesh Das Gupta, Department of Statistics, Stanford University, Stanford, California 94305, U.S.A.

