Monotonicity of the Modified Likelihood Ratio Test for a Covariance Matrix

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1. Introduction and Summary

In our previous paper [4], we have proved that the modified likelihood ratio test (=modified LR test) for the equality of a covariance matrix Σ to a given one Σ_0 in a *p*-variate normal distribution is unbiased. The power function of this test depends only on the characteristic roots of $\Sigma \Sigma_0^{-1}$. In this note we prove that this power function is a monotonically increasing (decreasing) function of each of the characteristic roots of $\Sigma \Sigma_0^{-1}$, when it is greater (less) than one, that is, it has the monotonicity property.

2. The monotonicity of the test

Let $p \times 1$ vectors $X_1, \dots, X_N (N > p)$ be a random sample from a multivariate normal distribution with unknown mean vector μ and unknown covariance matrix $\sum (\det \Sigma \neq 0)$. We wish to test the hypothesis $H: \Sigma = \sum_0$ against the alternatives $K: \Sigma \neq \sum_0$ where μ is unknown and \sum_0 is a given positive definite matrix (p.d. matrix). The LR critical region for this problem is given by, as in Anderson [1],

(2.1)
$$\boldsymbol{\omega}' = \left\{ \boldsymbol{S} \mid \boldsymbol{S} \text{ is p.d. and } \mid \boldsymbol{S} \boldsymbol{\Sigma}_{0}^{-1} \mid \frac{N}{2} \operatorname{etr} \left[-\frac{1}{2} \boldsymbol{\Sigma}_{0}^{-1} \boldsymbol{S} \right] \leq c_{\alpha} \right\},$$

where the symbol etr means exptr, $S = \sum_{\alpha=1}^{N} (X_{\alpha} - \bar{X}) (X_{\alpha} - \bar{X})'$ and $\bar{X} = N^{-1} \sum_{\alpha=1}^{N} X_{\alpha}$. The constant c_{α} is determined such that the level of this test is α . By replacing $|S \sum_{0}^{-1}|^{N/2}$ to $|S \sum_{0}^{-1}|^{(N-1)/2}$ as in our previous paper [4], we can prove the following theorem.

THEOREM 1. For testing the hypothesis $H: \sum = \sum_0$ against the alternatives $K: \sum i \sum_0 for$ unknown mean μ , the following modified LR critical region given by

(2.2)
$$\boldsymbol{\omega} = \left\{ \mathbf{S} \mid \mathbf{S} \text{ is p.d. and } \mid \mathbf{S} \boldsymbol{\Sigma}_{0}^{-1} \mid^{\frac{n}{2}} \operatorname{etr} \left[-\frac{1}{2} \boldsymbol{\Sigma}_{0}^{-1} \mathbf{S} \right] \leq c_{\alpha} \right\}$$

has the monotonicity property with respect to each of the *p*-characteristic roots of $\sum \sum_{0}^{-1}$, that is, $\operatorname{ch}(\sum \sum_{0}^{-1}) = (\delta_1^2, ..., \delta_p^2)$, where $S = \sum_{\alpha=1}^{N} (X_{\alpha} - \overline{X}) (X_{\alpha} - \overline{X})'$ and n = N-1. More precisely, the power function increases (decreases) with respect

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to each of any δ_i^2 when $\delta_i^2 \ge 1$ ($\delta_i^2 \le 1$) for fixed $\delta_1^2, \dots, \delta_{i-1}^2, \delta_{i+1}^2, \dots, \delta_p^2$.

PROOF. The statistic S is distributed according to the Wishart distribution $W(\Sigma, n)$, so the power function of the test ω is given by

(2.3)
$$P_{K}(\boldsymbol{\omega} \mid \boldsymbol{\Sigma}) = c_{p,n} \int_{\boldsymbol{S} \in \boldsymbol{\omega}} |\boldsymbol{S}|^{\frac{1}{2}(n-p-1)} |\boldsymbol{\Sigma}^{-1}|^{\frac{n}{2}} \operatorname{etr} \left[-\frac{1}{2} \boldsymbol{\Sigma}^{-1} \boldsymbol{S} \right] d\boldsymbol{S}$$

where $c_{p,n}^{-1} = \pi^{p(p-1)/4} 2^{np/2} \boldsymbol{\Pi}_{i=1}^{p} \boldsymbol{\Gamma}[(n-i+1)/2]$. Put $\boldsymbol{A} = \sum_{0}^{-1/2} \boldsymbol{S} \underline{\boldsymbol{\Sigma}}_{0}^{-1/2}$, then the matrix \boldsymbol{A} is also p.d. and the Jacobian is given by $|\partial \boldsymbol{A}/\partial \boldsymbol{S}| = |\boldsymbol{\Sigma}_{0}^{-1}|^{(p+1)/2}$. Therefore we have

(2.4)
$$P_{K}(\boldsymbol{\alpha}_{1} | \boldsymbol{\Sigma}) = c_{p,n} \int_{\boldsymbol{A} \in \boldsymbol{\omega}_{1}} |\boldsymbol{A}|^{\frac{1}{2}(n-p-1)} |\boldsymbol{\Sigma}_{0}^{-1}\boldsymbol{\Sigma}|^{-\frac{n}{2}} \operatorname{etr}\left[-\frac{1}{2}\boldsymbol{\Sigma}^{-1}\boldsymbol{\Sigma}_{0}^{\frac{1}{2}}\boldsymbol{A}\boldsymbol{\Sigma}_{0}^{\frac{1}{2}}\right] d\boldsymbol{A}$$

where $\omega_1 = \left\{ A \mid A \text{ is p.d. and } \mid A \mid^{n/2} \operatorname{etr} \left[-\frac{1}{2}A \right] \leq c_{\alpha} \right\}$. Let T be an orthogonal matrix such that $T' \sum_{0}^{-1/2} \sum \sum_{0}^{-1/2} T = A$, where $A = \operatorname{diag}(\delta_1^2, \dots, \delta_p^2)$. Put B = T'AT, then B is also p.d. and the Jacobian is given by $|\partial B/\partial A| = 1$. On the other hand, $T'AT \in \omega_1$ is equivalent to $A \in \omega_1$ for any orthogonal matrix T. Thus we have

(2.5)
$$P_{K}(\boldsymbol{\omega}_{1} | \boldsymbol{\Sigma}) = c_{p,n} \int_{\boldsymbol{B} \in \boldsymbol{\omega}_{1}} |\boldsymbol{B}|^{\frac{1}{2}(n-p-1)} |\boldsymbol{A}|^{-\frac{n}{2}} \operatorname{etr} \left[-\frac{1}{2} \boldsymbol{\Lambda}^{-1} \boldsymbol{B} \right] d\boldsymbol{B}.$$

So the power function depends only on $\boldsymbol{\Lambda}$, namely, $P_{K}(\boldsymbol{\omega}_{1} | \boldsymbol{\Sigma}) = P_{K}(\boldsymbol{\omega}_{1} | \boldsymbol{\Lambda})$. Put $\boldsymbol{B} = \boldsymbol{D}^{1/2} \boldsymbol{R} \boldsymbol{D}^{1/2\dagger}$ where $\boldsymbol{D}^{1/2} = \operatorname{diag}(b_{11}^{1/2}, \dots, b_{pp}^{1/2})$ and b_{ii} means *i*-th diagonal element of \boldsymbol{B} , then the Jacobian $|\partial \boldsymbol{B}/\partial(\boldsymbol{R},\boldsymbol{D})| = |\boldsymbol{D}|^{(p-1)/2}$. Put $\boldsymbol{\omega}_{\boldsymbol{R}} = \{\boldsymbol{D} | \boldsymbol{B} = \boldsymbol{D}^{1/2} \boldsymbol{R} \boldsymbol{D}^{1/2} \boldsymbol{\epsilon} \boldsymbol{\omega}_{1}\}$, then we can write

(2.6)
$$P_{K}(\boldsymbol{\omega}_{1}|\boldsymbol{\Lambda}) = c_{p,n} \int_{\boldsymbol{R}>0} |\boldsymbol{R}|^{\frac{1}{2} (n-p-1)} d\boldsymbol{R} \int_{\boldsymbol{\omega}\boldsymbol{R}} |\boldsymbol{D}|^{\frac{n}{2}-1} |\boldsymbol{\Lambda}|^{-\frac{n}{2}} \operatorname{etr} \left[-\frac{1}{2} \boldsymbol{\Lambda}^{-1} \boldsymbol{D} \right] d\boldsymbol{D}$$

$$= c_{p,n} \int_{\boldsymbol{R}>0} |\boldsymbol{R}|^{\frac{1}{2} (n-p-1)} \beta(\delta_{1}^{2}, \dots, \delta_{p}^{2}|\boldsymbol{R}) d\boldsymbol{R},$$

where $\beta(\delta_1^2, \dots, \delta_p^2 | \mathbf{R}) = \int_{\mathbf{W}_{\mathbf{R}}} \mathbf{I}_{i=1}^{b} b_{ii}^{(n/2)-1} (\delta_i^2)^{-n/2} \exp[-b_{ii}/2\delta_i^2] db_{ii}$, and the region $\mathbf{R} > 0$ means the set of all p.d. matrices such that all diagonal elements are one. We can show that if $\delta_i^{*2} \ge \delta_i^2 \ge 1$ or $\delta_i^{*2} \le \delta_i^2 \le 1$, then

$$(2.7) \qquad \beta(\delta_1^2, \ldots, \delta_{i-1}^2, \ \delta_i^{*2}, \delta_{i+1}^2, \ldots, \delta_p^2 | \mathbf{R}) \geq \beta(\delta_1^2, \ldots, \delta_i^2, \ldots, \delta_p^2 | \mathbf{R}).$$

For instance, the range of the integration with respect to variable b_{11} for fixed b_{22}, \ldots, b_{pp} is written as $b_{11}^{n/2} \exp[-b_{11}/2] \leq c_{\alpha} |\mathbf{R}|^{-n/2} \mathbf{I}_{i=2}^{p} b_{ii}^{-n/2} \exp[b_{ii}/2]$. By the following lemma which assures the monotonicity of the power function of the test (2.2) in case of p=1, we have $\beta(\delta_{1}^{*2}, \delta_{2}^{2}, \ldots, \delta_{p}^{2} |\mathbf{R}) \geq \beta(\delta_{1}^{2}, \ldots, \delta_{p}^{2} |\mathbf{R})$

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[†] This transformation was used by Gleser [2] to prove the unbiasedness of the sphericity test.

for $\delta_1^{*2} \ge \delta_1^2 \ge 1$ or $\delta_1^{*2} \le \delta_1^2 \le 1$. By integrating the inequality multiplied by $|\mathbf{R}|^{(n-p-1)/2}$ on both sides of (2.7) with respect to \mathbf{R} , we have the desired conclusion.

Now it is enough to prove Theorem 1 in the univariate case, which was stated by Ramachandran [3]. He, however, did not mention its proof explicitly. So we write the lemma with the sketch of the proof. We use σ^2 and σ_0^2 instead of covariance matrix Σ and Σ_0 .

LEMMA. For testing the hypothesis $H: \sigma = \sigma_0$ against the alternatives $K: \sigma \neq \sigma_0$ for unknown mean μ , the modified LR critical region given by

(2.8)
$$\omega = \left\{ S \mid S > 0 \text{ and } (S\sigma_0^{-2})^{\frac{n}{2}} \exp\left[-\frac{1}{2}S\sigma_0^{-2}\right] \le c_{\alpha} \right\}$$

has the monotonicity property with respect to $\delta^2 = \sigma^2 / \sigma_0^2$, where $S = \sum_{\alpha=1}^{N} (X_{\alpha} - \overline{X})^2$, $\overline{X} = N^{-1} \sum_{\alpha=1}^{N} X_{\alpha}$ and n = N-1.

PROOF. Under the alternative K, the statistic S/σ^2 is distributed according to x^2 distribution with n degrees of freedom, so the power function of the test ω is given by

(2.9)
$$P(\omega | \sigma^2) = c_{1,n} \int_{S \in \omega} S^{\frac{n}{2} - 1} (\sigma^2)^{-\frac{n}{2}} \exp[-S/2\sigma^2] dS$$

where $c_{1,n}^{-1} = 2^{\frac{n}{2}} \boldsymbol{\Gamma}[n/2]$. Putting $z = \sigma_0^{-2} S$, we can write the power function (2.9) as

(2.10)
$$P(\omega_1 | \delta^2) = c_{1,n} \int_{z \in \omega_1} z^{\frac{n}{2}-1} (\delta^2)^{-\frac{n}{2}} \exp[-z/2\delta^2] dz,$$

where $\omega_1 = \{z \mid z > 0 \text{ and } z^{\frac{n}{2}} \exp[-z/2] \le c_{\alpha}\}$. Since the equation $z^{\frac{n}{2}} \exp[-z/2] = c_{\alpha}$ has exactly two solutions $z = c_1$ and $c_2(c_1 < c_2)$, we obtain

$$(2.11) \quad \frac{dp(\omega_1|\delta^2)}{d\delta^2} = c_{1,n}(\delta^2)^{-\binom{n}{2}-1} \left\{ c_2^{\frac{n}{2}} \exp\left[-c_2/2\delta^2\right] - c_1^{\frac{n}{2}} \exp\left[-c_1/2\delta^2\right] \right\} \\ = c_{1,n}(\delta^2)^{-\binom{n}{2}-1} c_1^{\frac{n}{2}} \exp\left[-c_2/2\delta^2\right] \left\{ \exp\left[(c_2-c_1)/2\right] - \exp\left[(c_2-c_1)/2\delta^2\right] \right\}.$$

Thus if $\delta^2 > 1$ then $dP(\omega_1 | \delta^2) / d\delta^2 > 0$ and if $\delta^2 < 1$ then $dP(\omega_1 | \delta^2) / d\delta^2 < 0$. Therefore the lemma is proved.

By the analogous argument as in the proof of Theorem 1, we have the following theorem.

THEOREM 2. For testing the hypothesis $H': \sum = \sum_{0}, \mu = \mu_{0}$ against the alternatives $K': \sum \rightleftharpoons \sum_{0}, \mu = \mu_{0}$ where \sum_{0} and μ_{0} are known, the LR critical region given by

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(2.12)
$$\boldsymbol{\omega}^* = \left\{ \mathbf{S}^* | \mathbf{S}^* \text{ is p.d. and } | \mathbf{S}^* \boldsymbol{\Sigma}_0^{-1} |^{\frac{N}{2}} \operatorname{etr} \left[-\frac{1}{2} \boldsymbol{\Sigma}_0^{-1} \mathbf{S}^* \right] \leq c_{\alpha} \right\},$$

where $S^* = \sum_{\alpha=1}^{N} (X_{\alpha} - \mu_0) (X_{\alpha} - \mu_0)'$, has the monotonicity property with respect to each of $ch(\sum \sum_{i=1}^{N} D_i)$.

We can also generalize Theorem 1 to the k sample case. Let $p \times 1$ vectors $X_{i1}, X_{i2}, ..., X_{iN_i} (N_i > p)$ be a random sample from p-variate normal distribution with mean μ_i and covariance matrix $\sum_i (i = 1, ..., k)$. Put $S_j = \sum_{\alpha=1}^{N_j} (X_{j\alpha} - \bar{X}_j) (X_{j\alpha} - \bar{X}_j)'$ and $n_j = N_j - 1$. Then we have the following theorem by the sama argument as in the proof of Theorem 1.

THEOREM 3. For testing the hypothesis $H'': \sum_{j} = \sum_{0j} (j=1, 2, ..., k)$ against the alternatives $K'': \sum_{i} \neq \sum_{0i}$ for some *i*, where the mean μ_{j} is unspecified and \sum_{0j} is a given p.d. matrix (j=1, 2, ..., k), the modified LR critical region given by

(2.13)
$$\boldsymbol{\omega}'' = \left\{ (S_1, S_2, \dots, S_k) | S_j \text{ is p.d. } (j=1, 2, \dots, k) \text{ and} \right. \\ \boldsymbol{\Pi}_{j=1}^k \left[\left. | S_j \sum_{0j}^{-1} |^{\frac{n_j}{2}} \operatorname{etr} \left(-\frac{1}{2} \sum_{0j}^{-1} S_j \right) \right] \le c_a \right\}$$

has the monotonicity property with respect to each of $ch(\sum_{j}\sum_{0}^{-1})$ (j=1, 2, ..., k).

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