

## ON SOME TEST CRITERIA FOR COVARIANCE MATRIX

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Some new test criteria are proposed for testing various hypotheses concerning covariance matrices. Asymptotic expansions of their null distributions are derived in terms of the  $\chi^2$  distribution.

**1. Introduction and summary.** This paper is concerned with problems of testing hypotheses (i) for the equality of covariance matrix to a given matrix (ii) for sphericity (iii) for independence and (iv) for the equality of covariance matrices. For testing these hypotheses when we have a multivariate normal population, the usual test criteria are the likelihood ratio (=LR) tests. When the alternative hypotheses are true, the LR tests of these hypotheses are asymptotically normal. Noting that these limiting distributions are singular at the null hypotheses since the variances vanish, we propose new test statistics which are obtained from these variances by replacing the covariance matrices by their unbiased estimators. Furthermore, we derive asymptotic expansions of the distributions of their test criteria, under the null hypotheses, in terms of  $\chi^2$  distributions.

**2. Preliminaries.** In this section we give some lemmas used in deriving the asymptotic expansions of the distributions under the null hypotheses. First we define the logarithm of a matrix. We denote by  $\mathfrak{A}$  the set of all real symmetric  $p \times p$  matrices and by  $\mathfrak{B}$  the set of all real symmetric positive definite  $p \times p$  matrices. Now we consider a mapping  $f$  from  $A (\in \mathfrak{A})$  to  $e^A$ . This matrix  $e^A$  can be defined as follows:

$$(2.1) \quad e^A = I + A + \frac{A^2}{2!} + \cdots + \frac{A^k}{k!} + \cdots$$

Then we have the following lemma. (See Chevalley [1] page 14.)

**LEMMA 2.1.** *This function  $f$  is one to one mapping and  $e^{\mathfrak{A}} = \mathfrak{B}$ , where  $e^{\mathfrak{A}}$  means the set of all matrices  $e^A$  for  $A \in \mathfrak{A}$ .*

By this lemma, we can define the logarithm for a real symmetric positive definite matrix.

Next we consider the distribution of the matrix  $Y = (n/2)^{\frac{1}{2}} \log S/n$ , where  $S(p \times p)$  has the Wishart distribution  $W(I, n)$ . Since  $S = ne^{(2/n)^{\frac{1}{2}}Y}$ , expressing the characteristic roots of the matrix  $Y$  as  $ch_i(Y)$ , as in Jack ([6] Lemma 8) the Jacobian is given by

$$(2.2) \quad \left| \frac{\partial S}{\partial Y} \right| = (2n)^{p(p+1)/4} \text{etr}[(2/n)^{\frac{1}{2}}Y] \prod_{i>j} \frac{f(\lambda_i) - f(\lambda_j)}{\lambda_i - \lambda_j},$$

where  $f(\lambda_i) = e^{\lambda_i}$  with  $\lambda_i = (2/n)^{\frac{1}{2}}ch_i(Y)$ .

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Since we are interested in obtaining asymptotic expansions, we note that the last term  $\prod_{i>j}^p [f(\lambda_i) - f(\lambda_j)]/[\lambda_i - \lambda_j]$  in the Jacobian (2.2) can be expanded for large  $n$  as

$$(2.3) \quad \prod_{i>j}^p \frac{f(\lambda_i) - f(\lambda_j)}{\lambda_i - \lambda_j} = 1 + \frac{1}{2}(p - 1)(2/n)^{\frac{1}{2}} \text{tr } Y + \frac{1}{12n} \{(3p^2 - 6p + 2)(\text{tr } Y)^2 + p \text{tr } Y^2\} + O(n^{-\frac{3}{2}}).$$

Since  $|e^A| = \text{etr } A$  for any square matrix  $A$ , the ‘‘asymptotic’’ distribution of  $Y$  is given by

$$(2.4) \quad c^* \cdot \text{etr} \left[ \frac{1}{2}(n - p + 1)(2/n)^{\frac{1}{2}} Y - (n/2)e^{(2/n)^{\frac{1}{2}} Y} \right] \times \left\{ 1 + \frac{1}{2}(p - 1)(2/n)^{\frac{1}{2}} \text{tr } Y + \frac{1}{12n} \{(3p^2 - 6p + 2)(\text{tr } Y)^2 + p \text{tr } Y^2\} + O(n^{-\frac{3}{2}}) \right\},$$

where

$$(2.5) \quad c^* = \left\{ \prod_{\alpha=1}^p \Gamma \left[ \frac{1}{2}(n + 1 - \alpha) \right] \right\}^{-1} (n/2)^{p(2n-p-1)/4} \pi^{-p(p-1)/4}.$$

Finally we state three well-known lemmas.

LEMMA 2.2. *Let a  $p(p + 1)/2 \times 1$  vector  $(y_{11}, y_{22}, \dots, y_{pp}, y_{12}, y_{13}, \dots, y_{p-1,p})'$  have a normal distribution with mean zero and covariance matrix  $\Sigma^* = (\sigma_{i,j;k,l})$ . Put the subscript pairs  $(i, j) = a, (k, l) = b, (m, n) = c, (q, r) = d, (s, t) = e$  and  $(u, v) = f$ , then we have*

$$(2.6) \quad E y_a y_b y_c y_d = \sigma_{a;b} \sigma_{c;d} + \sigma_{a;c} \sigma_{b;d} + \sigma_{a;d} \sigma_{b;c},$$

and

$$(2.7) \quad E y_a y_b y_c y_d y_e y_f = \sigma_{a;b} [\sigma_{c;d} \sigma_{e;f} + \sigma_{c;e} \sigma_{d;f} + \sigma_{c;f} \sigma_{d;e}] + \sigma_{a;c} [\sigma_{b;d} \sigma_{e;f} + \sigma_{b;e} \sigma_{d;f} + \sigma_{b;f} \sigma_{d;e}] + \sigma_{a;d} [\sigma_{b;c} \sigma_{e;f} + \sigma_{b;e} \sigma_{c;f} + \sigma_{b;f} \sigma_{c;e}] + \sigma_{a;e} [\sigma_{b;c} \sigma_{d;f} + \sigma_{b;d} \sigma_{c;f} + \sigma_{b;f} \sigma_{c;d}] + \sigma_{a;f} [\sigma_{b;c} \sigma_{d;e} + \sigma_{b;d} \sigma_{c;e} + \sigma_{b;e} \sigma_{c;d}].$$

LEMMA 2.3. *Let  $A$  and  $C$  be  $p \times p$  matrices and  $B$  and  $D$  be  $q \times q$  matrices. Then for the Kronecker product of matrices, we have*

$$(2.8) \quad (A + C) \otimes B = A \otimes B + C \otimes B,$$

$$(2.9) \quad (A \otimes B)(C \otimes D) = AC \otimes BD,$$

and

$$(2.10) \quad |A \otimes B| = |A|^q |B|^p.$$

LEMMA 2.4. *Let  $P_1$  and  $P_2$  be symmetric and idempotent matrices such that  $P_1 + P_2 = I$  and  $P_1 P_2 = 0$ . Then for any numbers  $c_1$  and  $c_2$  ( $c_1 c_2 \neq 0$ ),*

$$(2.11) \quad (c_1 P_1 + c_2 P_2)^{-1} = c_1^{-1} P_1 + c_2^{-1} P_2.$$

**3. Some test statistics for covariance.** Let the  $p \times 1$  vectors  $X_1, X_2, \dots, X_N$  be a random sample from a normal distribution with mean vector  $\mu$  and covariance matrix  $\Sigma$ . The modified LR criterion for testing the hypothesis  $H_1: \Sigma = \Sigma_0$  against the alternatives  $K_1: \Sigma \neq \Sigma_0$  for some given positive definite matrix  $\Sigma_0$ , is given by

$$(3.1) \quad \lambda_1 = (e/n)^{(np/2)} |S\Sigma_0^{-1}|^{n/2} \text{etr} \left[ -\frac{1}{2} \Sigma_0^{-1} S \right],$$

where  $S = \sum_{\alpha=1}^N (X_\alpha - \bar{X})(X_\alpha - \bar{X})'$ ,  $\bar{X} = N^{-1} \sum_{\alpha=1}^N X_\alpha$  and  $n = N - 1$ . The unbiasedness of this test is shown by Sugiura and Nagao [14] and the monotonicity of the power function with respect to  $p$  characteristic roots of  $\Sigma\Sigma_0^{-1}$  is established by Nagao [10] and Das Gupta [2]. Also Sugiura [15] gave the asymptotic expansion of the statistic  $-2 \log \lambda_1$  under the alternative, the first term of which is a normal distribution. Our concern is the variance of the limiting distribution. Without a constant multiplier, this variance is given by

$$(3.2) \quad \tau_1^2 = \text{tr} (\Sigma\Sigma_0^{-1} - I)^2.$$

Thus  $\tau_1^2$  may be regarded as a measure of departure from the null hypothesis. So replacing  $\Sigma$  by the unbiased estimator  $S/n$ , we propose a test criterion

$$(3.3) \quad T_1 = \frac{n}{2} \text{tr} (S\Sigma_0^{-1}/n - I)^2.$$

Thus the  $T_1$  test rejects the hypothesis  $H_1$ , when the observed value of  $T_1$  is larger than a preassigned constant.

Similarly, using the same notation as above, the modified LR statistic for testing the sphericity hypothesis  $H_2: \Sigma = \sigma^2 I$  against the alternatives  $K_2: \Sigma \neq \sigma^2 I$ , where  $\sigma^2$  is unspecified, is given by

$$(3.4) \quad \lambda_2 = |S|^{n/2} (p^{-1} \text{tr } S)^{-np/2}.$$

The unbiasedness of this test criterion was proved by Gleser [4] and Sugiura and Nagao [14]. Also Sugiura [15] and Nagao [12] derived the asymptotic expansions under the alternatives and the local alternatives, respectively. The variance of the limiting distribution of  $-2 \log \lambda_2$  is proportional to

$$(3.5) \quad \tau_2^2 = \text{tr} \{ \Sigma (\text{tr } \Sigma)^{-1} - p^{-1} I \}^2.$$

Thus for the sphericity test, we propose the following test statistic:

$$(3.6) \quad T_2 = \frac{p^2 n}{2} \text{tr} \{ S(\text{tr } S)^{-1} - p^{-1} I \}^2.$$

From a different standpoint, the above criterion  $T_2$  was shown to be the locally best invariant test by John [7] and Sugiura [16].

Next we consider the test of independence. Let the  $1 \times p$  vector  $X' = (X^{(1)'}, X^{(2)'}, \dots, X^{(q)'})$ , having a normal distribution with mean vector  $\mu' = (\mu^{(1)'}, \dots, \mu^{(q)'})$  and covariance matrix  $\Sigma = (\Sigma_{ij})$ , ( $i, j = 1, 2, \dots, q$ ), be partitioned into  $q$  sub-vectors with components  $p_1, p_2, \dots, p_q$ , respectively. Given a sample  $X_1, X_2, \dots, X_N$  of  $N$  observations on  $X$ , we wish to test the hypothesis  $H_3: \Sigma_{ij} = 0$

( $i \neq j$ ) against the alternatives  $K_3: \Sigma_{ij} \neq 0$  for some  $i, j$  ( $i \neq j$ ). Then the LR criterion for this hypothesis is given by

$$(3.7) \quad \lambda_3 = \frac{|S|^{N/2}}{\prod_{\alpha=1}^q |S_{\alpha\alpha}|^{N/2}},$$

where  $S$  is partitioned in the same manner as  $\Sigma$ . The corresponding matrix  $S_{\alpha\alpha}$  is defined. The unbiasedness of this test is shown by Narain [9]. Also Nagao [13] gave the asymptotic expansion of the distribution under the fixed alternative. The asymptotic variance of  $-2 \log \lambda_3$  is proportional to

$$(3.8) \quad \tau_3^2 = \text{tr}(\Sigma \Sigma_D^{-1} - I)^2,$$

where the above  $\Sigma_D$  is a matrix obtained by replacing all submatrices  $\Sigma_{ij} = 0$  ( $i \neq j$ ) in the matrix  $\Sigma$ . Thus for the test of independence, we propose the following test statistic:

$$(3.9) \quad T_3 = \frac{n}{2} \text{tr}(S S_D^{-1} - I)^2,$$

where  $S_D$  is a matrix which is obtained by replacing  $\Sigma_{\alpha\alpha}$  by  $S_{\alpha\alpha}$  in  $\Sigma_D$ . In the case  $q = 2$ , (3.9) reduces to  $T_3 = n \text{tr} S_{12} S_{22}^{-1} S_{21} S_{11}^{-1}$ , which is Pillai's statistic.

Finally by the same consideration, for testing for the homogeneity of covariance matrices we propose a test criterion. Let the  $p \times 1$  vectors  $X_{i1}, X_{i2}, \dots, X_{iN_i}$  be a random sample from a normal distribution with mean  $\mu_i$  and covariance matrix  $\Sigma_i$  ( $i = 1, 2, \dots, k$ ). The modified LR statistic for testing the hypothesis  $H_4: \Sigma_1 = \Sigma_2 = \dots = \Sigma_k$  against the alternatives  $K_4: \Sigma_i \neq \Sigma_j$  for some  $i, j$  ( $i \neq j$ ), is given by

$$(3.10) \quad \lambda_4 = \{ \prod_{\alpha=1}^k |S_{\alpha}/n_{\alpha}|^{n_{\alpha}/2} \} / |S/n|^{n/2},$$

where  $S_{\alpha} = \sum_{\beta=1}^{N_{\alpha}} (X_{\alpha\beta} - \bar{X}_{\alpha})(X_{\alpha\beta} - \bar{X}_{\alpha})'$ ,  $\bar{X}_{\alpha} = N_{\alpha}^{-1} \sum_{\beta=1}^{N_{\alpha}} X_{\alpha\beta}$ ,  $S = \sum_{\alpha=1}^k S_{\alpha}$ ,  $n_{\alpha} = N_{\alpha} - 1$  and  $n = \sum_{\alpha=1}^k n_{\alpha}$ . Under the alternative, the asymptotic expansion of the distribution of the statistic  $-2 \log \lambda_4$  was given by the previous paper [11]. Also in the two-sample case this test is shown to be unbiased by Sugiura and Nagao [14]. The variance of the limiting distribution without a constant multiplier is expressed as

$$(3.11) \quad \tau_4^2 = \sum_{\alpha=1}^k \rho_{\alpha} \text{tr}(\Sigma_{\alpha} \tilde{\Sigma}^{-1} - I)^2,$$

where  $\rho_{\alpha} = n_{\alpha}/n$  and  $\tilde{\Sigma} = \sum_{\alpha=1}^k \rho_{\alpha} \Sigma_{\alpha}$ .

Thus for homogeneity of covariances, we propose the following test criterion:

$$(3.12) \quad T_4 = \frac{1}{2} \sum_{\alpha=1}^k n_{\alpha} \text{tr} \left\{ \frac{S_{\alpha}}{n_{\alpha}} \left( \frac{S}{n} \right)^{-1} - I \right\}^2.$$

**4. Expansion of the distribution of the criterion for  $\Sigma = \Sigma_0$ .** We shall give the asymptotic expansion of the null distribution of the test criterion  $T_1$ . Without loss of generality we may assume that  $S$  has the Wishart distribution  $W(I, n)$ . Then the characteristic function of the statistic  $T_1 = (n/2) \text{tr}(S/n - I)^2$  is given

by

$$(4.1) \quad C_1(t) = c_{p,n} \int \exp[itT_1] |S|^{\frac{1}{2}(n-p-1)} \text{etr}[-\frac{1}{2}S] dS,$$

where

$$(4.2) \quad c_{p,n}^{-1} = 2^{np/2} \pi^{p(p-1)/4} \prod_{\alpha=1}^p \Gamma[\frac{1}{2}(n+1-\alpha)].$$

By expressing  $T_1$  with  $Y = (n/2)^{\frac{1}{2}} \log S/n$ , we have

$$(4.3) \quad T_1 = \text{tr } Y^2 + (2/n)^{\frac{1}{2}} \text{tr } Y^3 + \frac{7}{6n} \text{tr } Y^4 + O(n^{-\frac{3}{2}}).$$

Therefore using the formula (2.4) we can rewrite the characteristic function  $C_1(t)$  as

$$(4.4) \quad \begin{aligned} C_1(t) = c^* \int \exp & \left[ (it) \text{tr } Y^2 + (2/n)^{\frac{1}{2}} (it) \text{tr } Y^3 \right. \\ & \left. + \frac{7}{6n} (it) \text{tr } Y^4 + \frac{1}{2}(n-p+1)(2/n)^{\frac{1}{2}} \text{tr } Y - \frac{n}{2} \text{tr } e^{(2/n)^{\frac{1}{2}} Y} \right] \\ & \times \left\{ 1 + \frac{1}{2}(p-1)(2/n)^{\frac{1}{2}} \text{tr } Y \right. \\ & \left. + \frac{1}{12n} \{ (3p^2 - 6p + 2)(\text{tr } Y)^2 + p \text{tr } Y^2 \} + O(n^{-\frac{3}{2}}) \right\} dY, \end{aligned}$$

where the region of the integral is the set of all real symmetric  $p \times p$  matrices. Since we have

$$(4.5) \quad e^{(2/n)^{\frac{1}{2}} Y} = I + (2/n)^{\frac{1}{2}} Y + \frac{1}{n} Y^2 + \frac{2^{\frac{1}{2}}}{3nn^{\frac{1}{2}}} Y^3 + \frac{1}{6n^2} Y^4 + O(n^{-\frac{3}{2}}),$$

after some calculation, the function  $C_1(t)$  is given by

$$(4.6) \quad \begin{aligned} C_1(t) = c^* \cdot \exp \left[ -\frac{np}{2} \right] \int \exp & \left[ -\frac{1}{2}(1 - 2it) \text{tr } Y^2 \right] \left[ 1 + (2/n)^{\frac{1}{2}} (it - \frac{1}{8}) \text{tr } Y^3 \right. \\ & \left. + \frac{1}{n} \left\{ -\frac{1}{2}(\text{tr } Y)^2 + \frac{p}{12} \text{tr } Y^2 + \frac{1}{2}(14it - 1) \text{tr } Y^4 \right. \right. \\ & \left. \left. + (it - \frac{1}{8})^2 (\text{tr } Y^3)^2 \right\} + O(n^{-\frac{3}{2}}) \right] dY. \end{aligned}$$

Thus we may regard the variable  $Y$  as having a  $p(p+1)/2$  dimensional normal distribution, with mean zero and covariance matrix  $(\sigma_{i,j;k,l})$  with

$$\sigma_{i,j;k,l} = (1 - 2it)^{-1} (\delta_{ik} \delta_{jl} + \delta_{il} \delta_{jk}) / 2.$$

Since  $|(\sigma_{i,j;k,l})| = (1 - 2it)^{-f} 2^{-p(p-1)/2}$ , we have

$$(4.7) \quad \begin{aligned} C_1(t) = c \cdot (1 - 2it)^{-f/2} E & \left[ 1 + (2/n)^{\frac{1}{2}} (it - \frac{1}{8}) \text{tr } Y^3 \right. \\ & \left. + \frac{1}{n} \left\{ -\frac{1}{2}(\text{tr } Y)^2 + \frac{p}{12} \text{tr } Y^2 + \frac{1}{2}(14it - 1) \text{tr } Y^4 \right. \right. \\ & \left. \left. + (it - \frac{1}{8})^2 (\text{tr } Y^3)^2 \right\} + O(n^{-\frac{3}{2}}) \right], \end{aligned}$$

where  $f = \frac{1}{2}p(p + 1)$  and

$$(4.8) \quad c = c^*(2\pi)^{\frac{1}{2}p(p+1)}2^{-p(p-1)/4} \exp[-\frac{1}{2}pn].$$

Using Lemma 2.2, we can obtain the following moments:

$$(4.9) \quad E(\text{tr } Y)^2 = p(t)_1, \quad E \text{tr } Y^2 = \frac{1}{2}p(p + 1)(t)_1,$$

$$E \text{tr } Y^4 = \frac{p}{4} (2p^2 + 5p + 5)(t)_2, \quad E(\text{tr } Y^3)^2 = \frac{3}{4}p(4p^2 + 9p + 7)(t)_3,$$

where  $(t)_\alpha = (1 - 2it)^{-\alpha}$ .

Since all product moments of odd degree from a normal population with mean zero are zero, the expectations in (4.7) are given by

$$(4.10) \quad 1 + \frac{1}{n} \left\{ \frac{p}{2} (p + 1)^2(t)_1 - \frac{p}{8} (6p^2 + 13p + 9)(t)_2 \right. \\ \left. + \frac{p}{12} (4p^2 + 9p + 7)(t)_3 \right\} + O(n^{-2}).$$

Applying Stirling's formula to the constant factor in (4.7), we have

$$(4.11) \quad (1 - 2it)^{-(f/2)} \left\{ 1 - \frac{p}{24n} (2p^2 + 3p - 1) \right\} + O(n^{-2}).$$

Thus we obtain the following asymptotic formula:

$$(4.12) \quad C_1(t) = (1 - 2it)^{-(f/2)} \left[ 1 + \frac{1}{n} \left\{ \frac{p}{12} (4p^2 + 9p + 7)(t)_3 \right. \right. \\ \left. \left. - \frac{p}{8} (6p^2 + 13p + 9)(t)_2 + \frac{p}{2} (p + 1)^2(t)_1 \right. \right. \\ \left. \left. - \frac{p}{24} (2p^2 + 3p - 1) \right\} + O(n^{-2}) \right].$$

Inversion of this characteristic function yields the following theorem:

**THEOREM 4.1.** *The null distribution of the  $T_1$  given by (3.3), expanded asymptotically in terms of the  $\chi^2$ -distribution for large  $n$ , is*

$$(4.13) \quad \Pr(T_1 \leq x) = P_f + \frac{1}{n} \left\{ \frac{p}{12} (4p^2 + 9p + 7)P_{f+6} - \frac{p}{8} (6p^2 + 13p + 9)P_{f+4} \right. \\ \left. + \frac{p}{2} (p + 1)^2P_{f+2} - \frac{p}{24} (2p^2 + 3p - 1)P_f \right\} + O(n^{-2}),$$

where  $f = \frac{1}{2}p(p + 1)$  and  $P_f = P(\chi_f^2 \leq x)$ .

**5. Expansion of the distribution of the criterion for sphericity.** By the same consideration as above, the characteristic function of the criterion  $T_2$  is given by

$$(5.1) \quad C_2(t) = c \cdot (1 - 2it)^{-(f/2)} E \left[ 1 + \frac{1}{n} \left\{ \frac{p}{12} \text{tr } Y^2 - \frac{1}{12} (\text{tr } Y)^2 \right. \right. \\ \left. \left. + \frac{1}{12} (14it - 1) \text{tr } Y^4 - \frac{14}{3p} (it) \text{tr } Y \text{tr } Y^3 - \frac{5}{2p} (it)(\text{tr } Y)^2 \right\} \right]$$

$$\begin{aligned}
 & + \frac{12}{p^2} (it)(\text{tr } Y)^2 \text{tr } Y^2 - \frac{6}{p^3} (it)(\text{tr } Y)^4 \\
 & + \left[ (it - \frac{1}{6}) \text{tr } Y^3 - \frac{3}{p} (it) \text{tr } Y \text{tr } Y^2 + \frac{2}{p^2} (it)(\text{tr } Y)^3 \right]^2 \Big\} \\
 & + O(n^{-2}) \Big],
 \end{aligned}$$

where  $f = \frac{1}{2}p(p + 1) - 1$  and the expectation in (5.1) is taken under a normal distribution with mean zero and covariance matrix  $(\sigma_{i,j;k,l})$  with  $\sigma_{i,j;k,l} = (t)_1(\delta_{ik}\delta_{jl} + \delta_{il}\delta_{jk})/2 + p^{-1}\{1 - (t)_1\}\delta_{ij}\delta_{kl}$ . After using Lemma 2.2 to calculate the expectations in (5.1) and simplifying the resulting expression, we obtain

$$\begin{aligned}
 (5.2) \quad C_2(t) & = (1 - 2it)^{-(f/2)} [1 + n^{-1}\{\frac{1}{12}(p^3 + 3p^2 - 8p - 12 - 200p^{-1})(t)_3 \\
 & + \frac{1}{8}(-2p^3 - 5p^2 + 7p + 12 + 420p^{-1})(t)_2 \\
 & + \frac{1}{4}(p^3 + 2p^2 - p - 2 - 216p^{-1})(t)_1 \\
 & + \frac{1}{24}(-2p^3 - 3p^2 + p + 436p^{-1})\}] + O(n^{-2}),
 \end{aligned}$$

where  $f = \frac{1}{2}p(p + 1) - 1$ .

By inverting this characteristic function, we have the following theorem:

**THEOREM 5.1.** *The null distribution of the test  $T_2$  given by (3.6), expanded asymptotically in terms of the  $\chi^2$ -distribution for large  $n$ , is*

$$\begin{aligned}
 (5.3) \quad \Pr(T_2 \leq x) & = P_f + n^{-1}\{\frac{1}{12}(p^3 + 3p^2 - 8p - 12 - 200p^{-1})P_{f+6} \\
 & + \frac{1}{8}(-2p^3 - 5p^2 + 7p + 12 + 420p^{-1})P_{f+4} \\
 & + \frac{1}{4}(p^3 + 2p^2 - p - 2 - 216p^{-1})P_{f+2} \\
 & + \frac{1}{24}(-2p^3 - 3p^2 + p + 436p^{-1})P_f\} + O(n^{-2}),
 \end{aligned}$$

where  $P_f = P(\chi_f^2 \leq x)$  with  $f = \frac{1}{2}p(p + 1) - 1$ .

**6. Expansion of the distribution of the criterion for independence.** The characteristic function of the criterion  $T_3$  is given by

$$\begin{aligned}
 (6.1) \quad C_3(t) & = c \cdot (1 - 2it)^{-(f/2)} E \left[ 1 + \frac{1}{n} \left\{ \frac{p}{12} \text{tr } Y^2 - \frac{1}{12} (\text{tr } Y)^2 \right. \right. \\
 & + \frac{1}{12} (14it - 1) \text{tr } Y^4 - \frac{1}{3} (it) \text{tr } Y_D Y^3 \\
 & - \frac{5}{2} (it) \text{tr } \sum_{\alpha=1}^q (\sum_{\beta=1}^q Y_{\alpha\beta} Y_{\beta\alpha})^2 + 10(it) \text{tr } Y_D^2 Y^2 \\
 & + 2(it) \text{tr } (Y_D Y)^2 - 6(it) \text{tr } Y_D^4 \\
 & \left. \left. + [(it - \frac{1}{6}) \text{tr } Y^3 - 3(it) \text{tr } Y_D Y^2 + 2(it) \text{tr } Y_D^3]^2 \right\} + O(n^{-2}) \right],
 \end{aligned}$$

where  $f = \frac{1}{2}(p^2 - \sum_{\alpha=1}^q p_{\alpha}^2)$  and  $Y_D$  is a matrix corresponding to  $\Sigma_D$ . The expectation of (6.1) is taken under the normal distribution with mean zero and covariance matrix  $(\sigma_{i,j;k,l})$  with  $\sigma_{i,j;k,l} = \frac{1}{2}(\delta_{ik}\delta_{jl} + \delta_{il}\delta_{jk})\{(t)_1 + \{1 - (t)_1\}\delta_{\alpha\beta}\delta_{\alpha\gamma}\delta_{\alpha\delta}\}$  for  $p_1 + \dots + p_{\alpha-1} + 1 \leq i \leq p_1 + \dots + p_{\alpha}$ ,  $p_1 + \dots + p_{\beta-1} + 1 \leq j \leq p_1 + \dots + p_{\beta}$ ,  $p_1 + \dots + p_{\gamma-1} + 1 \leq k \leq p_1 + \dots + p_{\gamma}$  and  $p_1 + \dots + p_{\delta-1} + 1 \leq l \leq p_1 + \dots + p_{\delta}$  with  $p_0 = 0$  ( $\alpha, \beta, \gamma, \delta = 1, 2, \dots, q$ ). Thus by Lemma

2.2 the characteristic function  $C_3(t)$  can be expanded asymptotically as

$$\begin{aligned}
 (6.2) \quad C_3(t) = & (1 - 2it)^{-(f/2)} [1 + n^{-1} \{ \frac{1}{12}(p^3 - 3p\bar{p}_2 + 2\bar{p}_3)(t)_3 \\
 & + \frac{1}{8}(-2p^3 + 4p\bar{p}_2 - 2\bar{p}_3 - p^2 + \bar{p}_2)(t)_2 \\
 & + \frac{1}{4}(p^3 - p\bar{p}_2 + p^2 - \bar{p}_2)(t)_1 + \frac{1}{2^4}(-2p^3 + 2\bar{p}_3 - 3p^2 + 3\bar{p}_2) \} ] \\
 & + O(n^{-2}),
 \end{aligned}$$

where  $\bar{p}_\beta = \sum_{\alpha=1}^q p_\alpha^\beta$ .

Inverting this characteristic function, we have the following theorem:

**THEOREM 6.1.** *The null distribution of the statistic  $T_3$  given by (3.9) is expanded asymptotically for large  $n$  as*

$$\begin{aligned}
 (6.3) \quad \Pr(T_3 \leq x) = & P_f + n^{-1} \{ \frac{1}{12}(p^3 - 3p\bar{p}_2 + 2\bar{p}_3)P_{f+6} \\
 & + \frac{1}{8}(-2p^3 + 4p\bar{p}_2 - 2\bar{p}_3 - p^2 + \bar{p}_2)P_{f+4} \\
 & + \frac{1}{4}(p^3 - p\bar{p}_2 + p^2 - \bar{p}_2)P_{f+2} \\
 & + \frac{1}{2^4}(-2p^3 + 2\bar{p}_3 - 3p^2 + 3\bar{p}_2)P_f \} + O(n^{-2}),
 \end{aligned}$$

where  $\bar{p}_\beta = \sum_{\alpha=1}^q p_\alpha^\beta$  and  $f = \frac{1}{2}(p^2 - \bar{p}_2)$ .

We remark that this formula (6.3) agrees, in the case  $q = 2$ , with expansions obtained previously by Lee [8] and Fujikoshi [3].

**7. Expansion of the distribution of the criterion for homogeneity of covariance matrices.** By putting  $Y_\alpha = (y_{ij}^{(\alpha)}) = (n_\alpha/2)^{\frac{1}{2}} \log S_\alpha/n_\alpha$  and  $\rho_\alpha = n_\alpha/n$  ( $\alpha = 1, 2, \dots, k$ ) with  $n = \sum_{\alpha=1}^k n_\alpha$ , the characteristic function of the statistic  $T_4$  is given by  $C_4(t) = C_4^*(t)C_4^{**}(t)$ , where

$$\begin{aligned}
 (7.1) \quad C_4^*(t) = & \prod_{\alpha=1}^k \{ [\prod_{j=1}^{p_\alpha} \Gamma[\frac{1}{2}(n_\alpha + 1 - j)]]^{-1} n_\alpha^{\frac{1}{2}p(2n_\alpha - p - 1)} \} \exp[-\frac{1}{2}np] \\
 & \times \pi^{(k/2)p} 2^{-kp(p-1)/4} (1 - 2it)^{-(f/2)},
 \end{aligned}$$

and

$$\begin{aligned}
 (7.2) \quad C_4^{**}(t) = & E \left[ 1 + \frac{1}{n} \left\{ \frac{p}{12} \sum_{\alpha=1}^k \text{tr } Y_\alpha^2 / \rho_\alpha - \frac{1}{12} \sum_{\alpha=1}^k (\text{tr } Y_\alpha)^2 / \rho_\alpha \right. \right. \\
 & - 6(it) \text{tr} (\sum_{\alpha=1}^k \rho_\alpha^{\frac{1}{2}} Y_\alpha)^4 - \frac{5}{2}(it) \text{tr} (\sum_{\alpha=1}^k Y_\alpha^2)^2 \\
 & + 2(it) \text{tr} \sum_{\beta=1}^k [Y_\beta (\sum_{\alpha=1}^k \rho_\alpha^{\frac{1}{2}} Y_\alpha)]^2 + \frac{1}{12}(14it - 1) \text{tr} \sum_{\alpha=1}^k Y_\alpha^4 / \rho_\alpha \\
 & + 10(it) \text{tr} (\sum_{\alpha=1}^k \rho_\alpha^{\frac{1}{2}} Y_\alpha)^2 (\sum_{\alpha=1}^k Y_\alpha^2) \\
 & - \frac{1}{3}(it) \text{tr} (\sum_{\alpha=1}^k Y_\alpha^3 / \rho_\alpha^{\frac{1}{2}}) (\sum_{\alpha=1}^k \rho_\alpha^{\frac{1}{2}} Y_\alpha) \\
 & + [(it - \frac{1}{6}) \text{tr} \sum_{\alpha=1}^k Y_\alpha^3 / \rho_\alpha^{\frac{1}{2}} + 2(it) \text{tr} (\sum_{\alpha=1}^k \rho_\alpha^{\frac{1}{2}} Y_\alpha)^3 \\
 & \left. \left. - 3(it) \text{tr} \sum_{\alpha=1}^k Y_\alpha^2 (\sum_{\alpha=1}^k \rho_\alpha^{\frac{1}{2}} Y_\alpha)^2 \right\} + O(n^{-2}) \right],
 \end{aligned}$$

with  $f = (k - 1)p(p + 1)/2$ . The expectation in (7.2) is taken under a  $kp(p + 1)/2$  dimensional normal distribution with mean zero and covariance matrix  $(\sigma_{i,j;k,l}^{(\alpha,\beta)})$ . This covariance matrix can be calculated as follows: Let  $y = (y_{11}^{(1)}, \dots, y_{pp}^{(1)}, y_{12}^{(1)}, \dots, y_{p-1,p}^{(1)}, \dots, y_{11}^{(k)}, \dots, y_{pp}^{(k)}, y_{12}^{(k)}, \dots, y_{p-1,p}^{(k)})'$ ,  $\gamma = (\rho_1^{\frac{1}{2}}, \rho_2^{\frac{1}{2}}, \dots, \rho_k^{\frac{1}{2}})'$  and  $\tilde{I} = \text{diag}(1, \dots, 1, 2, \dots, 2)$  where the multiplicities of the diagonal elements 1 and 2 are  $p$  and  $p(p - 1)/2$ , respectively. Then by Lemma 2.3 and 2.4 we



have

$$\begin{aligned}
 (1 - 2it) \sum_{\alpha=1}^k \text{tr } Y_{\alpha}^2 + 2(it) \text{tr} (\sum_{\alpha=1}^k \rho_{\alpha}^{\frac{1}{2}} Y_{\alpha})^2 \\
 (7.3) \quad &= y' \{ (1 - 2it) I_k \otimes \tilde{I} + 2(it) \gamma \gamma' \otimes \tilde{I} \} y \\
 &= y' \{ (1 - 2it) (I_k - \gamma \gamma') + \gamma \gamma' \} \otimes \tilde{I} \} y \\
 &= y' \{ (1 - 2it)^{-1} (I_k - \gamma \gamma') + \gamma \gamma' \} \otimes \tilde{I}^{-1} \}^{-1} y.
 \end{aligned}$$

Thus the covariance between  $y_{ij}^{(\alpha)}$  and  $y_{kl}^{(\beta)}$  is given by

$$\sigma_{i,j;k,l}^{(\alpha,\beta)} = \{ (t)_1 (\delta_{\alpha\beta} - (\rho_{\alpha} \rho_{\beta})^{\frac{1}{2}}) + (\rho_{\alpha} \rho_{\beta})^{\frac{1}{2}} \} (\delta_{ik} \delta_{jl} + \delta_{il} \delta_{jk}) / 2.$$

Since we have

$$(7.4) \quad C_4^*(t) = (1 - 2it)^{-(f/2)} \left[ 1 - \frac{1}{24n} \bar{\rho} p (2p^2 + 3p - 1) + O(n^{-2}) \right],$$

by using Lemma 2.2 we can express  $C_4(t)$  in the form

$$\begin{aligned}
 C_4(t) = (1 - 2it)^{-(f/2)} [ 1 + n^{-1} [ \frac{1}{12} \{ \bar{\rho} p (4p^2 + 9p + 7) - 3k^2 p (p + 1)^2 \\
 - (3k - 2) p (p^2 + 3p + 4) \} (t)_3 + \frac{1}{8} \{ -\bar{\rho} p (6p^2 + 13p + 9) \\
 (7.5) \quad + 4k^2 p (p + 1)^2 + (2k - 1) p (2p^2 + 5p + 5) \} (t)_2 \\
 + \frac{1}{4} (2\bar{\rho} - k^2 - k) p (p + 1)^2 (t)_1 + \frac{1}{24} (1 - \bar{\rho}) p (2p^2 + 3p - 1) ] \\
 + O(n^{-2}) ],
 \end{aligned}$$

where  $\bar{\rho} = \sum_{\alpha=1}^k \rho_{\alpha}^{-1}$ .

Inverting (7.5) gives

**THEOREM 7.1.** *The null distribution of the test  $T_4$  given by (3.12), for large  $n = \sum_{\alpha=1}^k n_{\alpha}$ , is*

$$\begin{aligned}
 \Pr(T_4 \leq x) = P_f + n^{-1} [ \frac{1}{12} \{ \bar{\rho} p (4p^2 + 9p + 7) - 3k^2 p (p + 1)^2 \\
 - (3k - 2) p (p^2 + 3p + 4) \} P_{f+6} + \frac{1}{8} \{ -\bar{\rho} p (6p^2 + 13p + 9) \\
 (7.6) \quad + 4k^2 p (p + 1)^2 + (2k - 1) p (2p^2 + 5p + 5) \} P_{f+4} \\
 + \frac{1}{4} (2\bar{\rho} - k^2 - k) p (p + 1)^2 P_{f+2} \\
 + \frac{1}{24} (1 - \bar{\rho}) p (2p^2 + 3p - 1) P_f ] + O(n^{-2}),
 \end{aligned}$$

where  $\bar{\rho} = \sum_{\alpha=1}^k \rho_{\alpha}^{-1}$  and  $P_f = P(\chi_f^2 \leq x)$  with  $f = \frac{1}{2}(k - 1)p(p + 1)$ .

**8. Numerical example.** The null distribution of each test statistic  $T_i$  has the following asymptotic expansion:

$$(8.1) \quad \Pr(T_i \leq x) = P_f + n^{-1} \{ a_6 P_{f+6} + a_4 P_{f+4} + a_2 P_{f+2} + a_0 P_f \} + O(n^{-2}),$$

where  $\sum_{i=0}^3 a_{2i} = 0$  and the different coefficients  $a$ 's and  $f$  degrees of freedom depend on  $T_i$ . Applying the general inverse expansion formula due to Hill and Davis [5], we obtain the following asymptotic formula for the  $100\alpha$  % point of  $T_i$ :

$$\begin{aligned}
 (8.2) \quad u + \frac{1}{n} \left[ \frac{2a_6 u}{f(f+2)(f+4)} \{ u^2 + (f+4)u + (f+2)(f+4) \} \right. \\
 \left. + \frac{2a_4 u}{f(f+2)} (u + f + 2) + \frac{2a_2 u}{f} \right] + O(n^{-2}),
 \end{aligned}$$

where  $\Pr(\chi_f^2 \geq u) = \alpha$ .

We give the approximate 5 % point of each test  $T_i$  ( $i = 1, 2, 3, 4$ ).

test	$T_1$	$T_2$	$T_3$	$T_4$
case	$p = 2,$ $n = 100$	$p = 2,$ $n = 100$	$p_1 = p_2 = p_3 = 1,$ $n = 100$	$p = 2,$ $n_1 = n_2 = 100$
first term	7.815	5.991	7.815	7.815
second term	0.033	0.150	-0.076	-0.149
approx. value	7.848	6.141	7.739	7.666

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