

ON TESTING THE HYPOTHESIS THAT SUBMATRICES OF THE MULTIVARIATE REGRESSION MATRICES OF k POPULATIONS ARE EQUAL

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(Received Sept. 28, 1963; revised July 29, 1964)

1. Introduction and summary

Let $\Pi_1, \Pi_2, \dots, \Pi_k$ be k normal p variate populations which are independently distributed. Let

$$\begin{bmatrix} X_1^{(g)} \\ Z_1^{(g)} \end{bmatrix}, \begin{bmatrix} X_2^{(g)} \\ Z_2^{(g)} \end{bmatrix}, \dots, \begin{bmatrix} X_{N_g}^{(g)} \\ Z_{N_g}^{(g)} \end{bmatrix}$$

denote N_g independent samples from the g th population Π_g , where $X_\alpha^{(g)}$ and $Z_\alpha^{(g)}$ are vectors of observed and fixed variates, respectively. The relation between $E(X_\alpha^{(g)})$ and $Z_\alpha^{(g)}$ for the α th sample of the g th population is given by

$$E(X_\alpha^{(g)}) = B_g Z_\alpha^{(g)} \quad (\alpha=1, 2, \dots, N_g, g=1, 2, \dots, k),$$

where B_g is a $p \times q$ matrix. Partitioning B_g into two parts

$$B_g = (B_{g1} B_{g2}), \quad g=1, 2, \dots, k,$$

where B_{g1} is a $p \times q_1$ matrix and B_{g2} is a $p \times q_2$ matrix ($q=q_1+q_2$), we consider the hypothesis

$$H: B_{11} = B_{21} = \dots = B_{k1} (=B_1^0)$$

and derive the likelihood ratio criterion and its distribution.

It is known that the likelihood ratio criterion is distributed as a ratio of central Wishart distributions with $N-kq$ and $N-kq_2$ degrees of freedom in the case where B_1^0 is a known matrix. In this paper we consider the case where B_1^0 is an unknown matrix. We assume that all the k populations have the same covariance matrix Σ . First we introduce the following matrix notations:

^{*)} *Editorial remark:* Kabe and Hayakawa obtained these results independently and due to editorial convenience the results were combined and published as a single paper.

$$\begin{aligned}
 X^{(g)} &= [X_1^{(g)}, X_2^{(g)}, \dots, X_{N_g}^{(g)}] & (p \times N_g), & \quad (g=1, 2, \dots, k) \\
 X &= [X^{(1)}, X^{(2)}, \dots, X^{(k)}] & (p \times N), & \quad N = \sum_{g=1}^k N_g, \\
 Z^{(g)} &= [Z_1^{(g)}, Z_2^{(g)}, \dots, Z_{N_g}^{(g)}] & (q \times N_g), \\
 &= \begin{bmatrix} Z_1^{(g)} \\ Z_2^{(g)} \end{bmatrix} = \begin{bmatrix} Z_{11}^{(g)}, Z_{21}^{(g)}, \dots, Z_{N_{g1}}^{(g)} \\ Z_{12}^{(g)}, Z_{22}^{(g)}, \dots, Z_{N_{g2}}^{(g)} \end{bmatrix} & \begin{matrix} (q_1 \times N_g) \\ (q_2 \times N_g), \quad q = q_1 + q_2, \end{matrix}
 \end{aligned}$$

where $Z_1^{(g)}$, $Z_2^{(g)}$, and $Z^{(g)}$ have ranks q_1 , q_2 and q , respectively,

$$\begin{aligned}
 Z &= [Z^{(1)}, Z^{(2)}, \dots, Z^{(k)}] & (q \times N) \\
 &= \begin{bmatrix} Z_1 \\ Z_2 \end{bmatrix} = \begin{bmatrix} Z_1^{(1)}, Z_1^{(2)}, \dots, Z_1^{(k)} \\ Z_2^{(1)}, Z_2^{(2)}, \dots, Z_2^{(k)} \end{bmatrix} & \begin{matrix} (q_1 \times N) \\ (q_2 \times N), \quad q = q_1 + q_2 \end{matrix}
 \end{aligned}$$

where Z_1 , Z_2 and Z have ranks q_1 , q_2 and q , respectively. We shall assume that $N_g \geq p + q$ for all g .

2. The maximum likelihood estimates

Suppose that $\{X^{(g)}\}$ are k independent observation matrices and $\{Z^{(g)}\}$ are k known matrices. Since $\{X_a^{(g)}\}$ are independently distributed according to $N[B_g Z_a^{(g)}, \Sigma]$, the likelihood function of B_1, B_2, \dots, B_k and Σ is

$$\begin{aligned}
 L(X) &= \prod_{g=1}^k L(X^{(g)}) \\
 &= \frac{1}{(2\pi)^{(1/2)PN} |\Sigma|^{(1/2)N}} \exp \left\{ -\frac{1}{2} \operatorname{tr} \Sigma^{-1} [X - (B_1 Z^{(1)}, B_2 Z^{(2)}, \dots, B_k Z^{(k)})] \right. \\
 &\quad \left. \cdot [X - (B_1 Z^{(1)}, B_2 Z^{(2)}, \dots, B_k Z^{(k)})]' \right\}.
 \end{aligned}$$

Let \hat{B}_g^0 ($g=1, \dots, k$) and $\hat{\Sigma}_0$ be the maximum likelihood estimates of B_g ($g=1, 2, \dots, k$) and Σ over the entire parameter space. Then we have

$$\begin{aligned}
 \hat{B}_g^0 &= X^{(g)} Z^{(g)'} [Z^{(g)} Z^{(g)'}]^{-1} & (g=1, 2, \dots, k), \\
 N \hat{\Sigma}_0 &= \sum_{g=1}^k X^{(g)} [I_{N_g} - Z^{(g)'} (Z^{(g)} Z^{(g)'})^{-1} Z^{(g)}] X^{(g)'},
 \end{aligned}$$

where I_{N_g} is the $N_g \times N_g$ identity matrix. Let

Let

$$E_3^{(g)} = F^{(g)} Z_2^{(g)} \quad (q_2 \times N_g), \quad (g=1, 2, \dots, k).$$

Then

$$E_3^{(g)} E_3^{(g)'} = F^{(g)} Z_2^{(g)} Z_2^{(g)'} F^{(g)'} = I_{q_2}.$$

Let

$$E_{3(kq_2 \times N)} = \begin{bmatrix} E_3^{(1)} & & & \\ & E_3^{(2)} & & 0 \\ & & \ddots & \\ & 0 & & E_3^{(k)} \end{bmatrix}$$

$$= \begin{bmatrix} F^{(1)} & & & \\ & F^{(2)} & & 0 \\ & & \ddots & \\ & 0 & & F^{(k)} \end{bmatrix} \begin{bmatrix} Z_2^{(1)} & & & \\ & Z_2^{(2)} & & 0 \\ & & \ddots & \\ & 0 & & Z_2^{(k)} \end{bmatrix}.$$

Then

$$E_3 E_3' = I_{kq_2}.$$

By using $E_3^{(g)}$'s, we can represent C as

$$C = \begin{bmatrix} E_3^{(1)'} E_3^{(1)} & & & \\ & E_3^{(2)'} E_3^{(2)} & & 0 \\ & & \ddots & \\ & 0 & & E_3^{(k)'} E_3^{(k)} \end{bmatrix} = E_3' E_3.$$

$$2^\circ. \quad YBY' = Y[Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \dots, Z_1^{(k)}P^{(k)}]' A_{11.2}^{-1} [Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \dots, Z_1^{(k)}P^{(k)}].$$

Since $\text{rank } A_{11.2} = q_1$, there exists also non-singular matrix R such that

$$RA_{11.2}R' = I_{q_1}.$$

Let

$$E_2 = R[Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \dots, Z_1^{(k)}P^{(k)}] \quad (q_1 \times N).$$

Then

$$E_2 E_2' = RA_{11.2}R' = I_{q_1}.$$

We also have

$$\begin{aligned}
 V = \lambda^{2/N} &= \frac{|N \hat{\Sigma}_a|}{|N \hat{\Sigma}_b|} = \frac{|Y(I-A)Y'|}{|Y(I-B-C)Y'|} \\
 &= \frac{|Y(I-A)Y'|}{|Y(I-A)Y' + Y(A-B-C)Y'|}.
 \end{aligned}$$

It is easily seen that $A-B-C$ is an idempotent matrix. Now, we apply theorem 1 of Hogg's [2] to

$$Y(I-B-C)Y' = Y(I-A)Y' + Y(A-B-C)Y',$$

where $Y(I-B-C)Y'$ and $Y(I-A)Y'$ are distributed according to the central Wishart distribution with $N-kq_2-q_1$ and $N-kq$ degrees of freedom, respectively, and $A-B-C$ is positive semi-definite. Then we see that $Y(A-B-C)Y'$ is distributed according to the central Wishart distribution with $(k-1)q_1$ degrees of freedom, independently of $Y(I-A)Y'$. Thus $\lambda^{2/N}$ is a U -statistic such that $U_{p, (k-1)q_1, N-kq}$ is distributed as $\lambda^{2/N} = \prod_{i=1}^p X_i$, where X_i has the beta density $\beta(x; \frac{1}{2}(N-kq+1-i), \frac{1}{2}(k-1)q_1)$, and X_1, X_2, \dots, X_p are independent with each other, [1].

Although we have treated the aspect of testing submatrices, the testing of matrices follows on similar lines, and to test the hypothesis $B_1=B_2=B_3=\dots=B_k$, we have the criterion

$$\lambda_1^{2/N} = \frac{|YY' - YAY'|}{|YY' - YAY' + YAY' - YZ'(ZZ')^{-1}ZY'|},$$

whose distribution is that of $U_{p, (k-1)q, N-kq}$.

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REFERENCES

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- [2] Robert V. Hogg, "On the independence of certain Wishart variables," *Ann. Math. Statist.*, 34 (1963), 935-939.