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

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## 1. Introduction and summary

Let  $\prod_1, \prod_2, \dots, \prod_k$  be k normal p variate populations which are independently distributed. Let

$$\left[egin{array}{c} X_1^{(g)} \ Z_1^{(g)} \end{array}
ight], \ \left[egin{array}{c} X_2^{(g)} \ Z_2^{(g)} \end{array}
ight], \ \cdots, \left[egin{array}{c} X_{N_g}^{(g)} \ Z_{N_g}^{(g)} \end{array}
ight]$$

denote  $N_g$  independent samples from the gth population  $\prod_g$ , 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  $\alpha$ th sample of the gth population is given by

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

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

$$B_{q}=(B_{q_1}B_{q_2}), \qquad q=1, 2, \cdots, k,$$

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

$$H: B_{11}=B_{21}=\cdots=B_{k1} \ (=B_1^o)$$

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 degrees of freedom in the case where  $B_1^o$  is a known matrix. In this paper we consider the case where  $B_1^o$  is an unknown matrix. We assume that all the k populations have the same covariance matrix  $\Sigma$ . First we introduce the following matrix notations:

<sup>\*)</sup> Editorial remark: Kabe and Hayakawa obtained these results independently and due to editorial convenience the results were combined and published as a single paper.

$$egin{aligned} X^{(g)} = & [X_1^{(g)}, X_2^{(g)}, \, \cdots, \, X_{N_g}^{(g)}] & (p imes N_g), & (g = 1, \, 2, \, \cdots, \, k) \ X = & [X^{(1)}, \, X^{(2)}, \, \cdots, \, X^{(k)}] & (p imes N), & N = \sum\limits_{g = 1}^k N_g, \ Z^{(g)} = & [Z_1^{(g)}, \, Z_2^{(g)}, \, \cdots, \, Z_{N_g}^{(g)}] & (q imes N_g), \ & = & \left[ egin{aligned} Z_1^{(g)}, \, Z_2^{(g)}, \, \cdots, \, Z_{N_g}^{(g)}, \, Z_{N_g}^{(g)}, \, \cdots, \, Z_{N_g}^{(g)}, \, Z_{N_g$$

where  $Z_1^{(g)}$ ,  $Z_2^{(g)}$ , and  $Z^{(g)}$  have ranks  $q_1$ ,  $q_2$  and  $q_3$ , respectively,

$$egin{aligned} m{Z} = & [m{Z}^{(1)}, m{Z}^{(2)}, \, \cdots, \, m{Z}^{(k)}] & (q imes N) \ = & \left[ m{Z}_1 \ m{Z}_2 \end{array} 
ight] = & \left[ m{Z}_1^{(1)}, \, m{Z}_1^{(2)}, \, \cdots, \, m{Z}_1^{(k)} \ m{Z}_2^{(2)}, \, \cdots, \, m{Z}_2^{(k)} \end{array} 
ight] & (q_1 imes N) \ (q_2 imes N), \quad q = q_1 + q_2 \end{aligned}$$

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

## 2. The maximum likelihood estimates

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

$$L(X) = \prod_{g=1}^{k} L(X^{(g)})$$

$$= \frac{1}{(2\pi)^{(1/2)PN} |\Sigma|^{(1/2)N}} \exp\left\{-\frac{1}{2} \operatorname{tr} \Sigma^{-1} [X - (B_1 Z^{(1)}, B_2 Z^{(2)}, \cdots, B_k Z^{(k)})] \right\}$$

$$\cdot [X - (B_1 Z^{(1)}, B_2 Z^{(2)}, \cdots, B_k Z^{(k)})]' \right\}.$$

Let  $\widehat{B}_{q}^{\rho}$   $(g=1,\dots,k)$  and  $\widehat{\Sigma}_{\rho}$  be the maximum likelihood estimates of  $B_{q}$   $(g=1,2,\dots,k)$  and  $\Sigma$  over the entire parameter space. Then we have

$$\hat{B}_{g}^{o} = X^{(g)} Z^{(g)'} [Z^{(g)} Z^{(g)'}]^{-1} \qquad (g=1, 2, \dots, k),$$

$$N \hat{\Sigma}_{o} = \sum_{g=1}^{k} X^{(g)} [I_{N_{g}} - Z^{(g)'} (Z^{(g)} Z^{(g)'})^{-1} Z^{(g)}] X^{(g)'},$$

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

Then, we can represent  $N\hat{\Sigma}_{\varrho}$  as

$$N \hat{\Sigma}_{\rho} = X(I_N - A)X'$$

where  $I_N$  is the  $N \times N$  identity matrix. Concerning this A, we have  $A^2 = A$ , that is, A is an idempotent matrix, and rank  $A = \operatorname{tr} A = kq$ . Further, it is known that X(I-A)X' is distributed according to the central Wishart distribution with N-kq degrees of freedom, [1]. Now we estimate  $B_q$  and  $\Sigma$  under the hypothesis H. The likelihood function is written as

$$L(X) = \frac{1}{(2\pi)^{(1/2)PN} |\Sigma|^{(1/2)N}} \cdot \exp\left\{-\frac{1}{2} \operatorname{tr} \Sigma^{-1} [X - B_1^{\circ} Z_1 - (B_{12} Z_2^{(1)}, B_{22} Z_2^{(2)}, \cdots, B_{k2} Z_2^{(k)})] \cdot [X - B_1^{\circ} Z_1 - (B_{12} Z_2^{(1)}, B_{22} Z_2^{(2)}, \cdots, B_{k2} Z_2^{(k)})]'\right\}.$$

Let  $\hat{B}_{1}^{o\omega}$ ,  $\hat{B}_{g2}^{\omega}$   $(g=1, 2, \dots, k)$  and  $\hat{\Sigma}_{\omega}$  be the maximum likelihood estimates of  $B_{1}^{o}$ ,  $B_{g2}$   $(g=1, 2, \dots, k)$  and  $\Sigma$  in the parameter space restricted by the null hypothesis H, respectively. Then we have

$$\hat{B}_{1}^{o\omega} = \sum_{g=1}^{k} X^{(g)} (Z_{1}^{(g)} P^{(g)})' A_{11.2}^{-1}$$
 ,

where

$$egin{aligned} &m{P}^{(g)} \!=\! m{I}_{N_g} \!-\! m{Z}_2^{(g)\prime} \!(m{Z}_2^{(g)} m{Z}_2^{(g)\prime})^{-1} m{Z}_2^{(g)}, \qquad (g \!=\! 1, 2, \cdots, k), \ &A_{11.2} \!=\! \sum\limits_{g=1}^k m{Z}_1^{(g)\prime} \!-\! (m{Z}_1^{(g)\prime} \!-\! (m{Z}_1^{(g)} m{Z}_2^{(g)\prime}) (m{Z}_2^{(g)} m{Z}_2^{(g)\prime})^{-1} (m{Z}_2^{(g)} m{Z}_1^{(g)\prime}) m{Z}_1^{(g)\prime}, \end{aligned}$$

and

$$\hat{B}_{g^2}^{\omega} = X^{(g)} Z_2^{(g)\prime} (Z_2^{(g)} Z_2^{(g)\prime})^{-1} - \hat{B}_1^{\sigma\omega} (Z_1^{(g)} Z_2^{(g)\prime}) (Z_2^{(g)} Z_2^{(g)\prime})^{-1}, \qquad (g=1, 2, \cdots, k),$$

and

$$\begin{split} N \, \widehat{\Sigma}_{w} = & X X' - [\sum_{g=1}^{k} X^{(g)} (Z_{1}^{(g)} P^{(g)})'] A_{11,2}^{-1} [\sum_{g=1}^{k} X^{(g)} (Z_{1}^{(g)} P^{(g)})']' \\ & - \sum_{g=1}^{k} X^{(g)} Z_{2}^{(g)} / (Z_{2}^{(g)} Z_{2}^{(g)})')^{-1} Z_{2}^{(g)} X^{(g)}. \end{split}$$

Here we assume that rank  $A_{11.2}=q_1$ . It can easily be seen that  $\hat{B}_1^{o}$  and  $\hat{B}_{g2}^{o}$  are unbiased estimates of  $B_1^{o}$  and  $B_{g2}$ , respectively. Let

$$B = [Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \cdots, Z_1^{(k)}P^{(k)}]'A_{11,2}^{-1}[Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \cdots, Z_1^{(k)}P^{(k)}]$$

$$(N \times N),$$

and

$$C = \begin{bmatrix} Z_{2}^{(1)\prime} (Z_{2}^{(1)} Z_{2}^{(1)\prime})^{-1} Z_{2}^{(1)} \\ Z_{2}^{(2)\prime} (Z_{2}^{(2)} Z_{2}^{(2)\prime})^{-1} Z_{2}^{(2)}, & 0 \\ \vdots & \vdots & \vdots \\ 0 & \ddots & \vdots \\ Z_{2}^{(k)\prime} (Z_{2}^{(k)} Z_{2}^{(k)\prime})^{-1} Z_{2}^{(k)} \end{bmatrix} (N \times N)$$

Then we see that  $B^2 = B$ ,  $C^2 = C$  and rank  $B = \text{tr } B = q_1$ , rank  $C = \text{tr } C = kq_2$ . We can write  $N\hat{\Sigma}_{\bullet}$  as

$$N\widehat{\Sigma}_{\alpha} = X(I_N - B - C)X'$$
.

It is easily seen that  $N\hat{\Sigma}_{\omega}$  is invariant for change of location

$$Y=X-E(X)=X-B_1^oZ_1-(B_{12}Z_2^{(1)},B_{22}Z_2^{(2)},\cdots,B_{k2}Z_2^{(k)}).$$

 $Y(p \times N)$  is distributed according to

$$\frac{1}{(2\pi)^{(1/2)PN}|\Sigma|^{(1/2)N}}\cdot\exp\left\{-\frac{1}{2}\operatorname{tr}\Sigma^{-1}YY'\right\},$$

and we have

$$N \hat{\Sigma}_{\alpha} = Y(I_N - B - C) Y'.$$

## 3. The distribution of $N \, \widehat{\Sigma}_{\omega}$

We now prove that  $N\hat{\Sigma}_{\omega}$  is distributed according to the central Wishart distribution with  $N-kq_1-q_1$  degrees of freedom.

1°. 
$$YCY' = \sum_{g=1}^{k} Y^{(g)} Z_2^{(g)} (Z_2^{(g)} Z_2^{(g)})^{-1} Z_2^{(g)} Y^{(g)}$$

Since  $Z_2^{(g)}Z_2^{(g)\prime}(q_1\times q_2)$  has rank  $q_2$ , there exists a non-singular matrix  $F^{(g)}$  such that

$$F^{(g)}(Z_2^{(g)}Z_2^{(g)\prime})F^{(g)\prime}=I_{q_2}$$

Let

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

Then

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

Let

Then

$$E_3E_3'=I_{kq_0}$$

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

2°. 
$$YBY' = Y[Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \cdots, Z_1^{(k)}P^{(k)}]'A_{11,2}^{-1}[Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \cdots, Z_1^{(k)}P^{(k)}].$$

Since rank  $A_{\scriptscriptstyle{11.2}}=q_{\scriptscriptstyle{1}}$ , there exists also non-singular matrix R such that

$$RA_{_{11.2}}R'=I_{q_{_{1}}}.$$

Let

$$E_1=R[Z_1^{(1)}P^{(1)}, Z_1^{(2)}P^{(2)}, \cdots, Z_1^{(k)}P^{(k)}] (q_1\times N).$$

Then

$$E_2E_2'=RA_{11.2}R'=I_{q_1}.$$

We also have

$$B=E_2'E_2$$

and

3°. Thus,  $E_{23} = \begin{bmatrix} E_2 \\ E_3 \end{bmatrix}$   $((kq_2+q_1) \times N)$  has  $(kq_2+q_1)$  orthogonal rows such that  $E_{23}E'_{23} = I_{kq_2+q_1}$ . It is possible to find a  $(N-kq_2-q_1) \times N$  matrix  $E_1$  such that  $E = \begin{bmatrix} E_1 \\ E_2 \\ E_3 \end{bmatrix}$  is orthogonal ([1], p. 225, equation (12)).

$$Y=[U_1, U_2, \cdots, U_N]E=UE.$$

Then

Now, let

$$\begin{split} N\widehat{\Sigma}_{\omega} &= UU' - U \begin{bmatrix} E_1 \\ E_2 \\ E_3 \end{bmatrix} E_2' E_2 [E_1' E_2' E_3'] U' \\ &- U \begin{bmatrix} E_1 \\ E_2 \\ E_3 \end{bmatrix} E_3' E_3 [E_1' E_2' E_3'] U' \\ &= UU' - U \begin{bmatrix} 0 & 0 & 0 \\ 0 & I_{q_1} & 0 \\ 0 & 0 & 0 \end{bmatrix} U' - U \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & I_{kq_2} \end{bmatrix} U' \\ &= \sum_{\alpha=1}^{N} U_{\alpha} U_{\alpha}' - \sum_{\alpha=N-kq_2-q_1+1}^{N-kq_2} U_{\alpha} U_{\alpha}' - \sum_{\alpha=N-kq_2+1}^{N} U_{\alpha} U_{\alpha}' \\ &= \sum_{\alpha=1}^{N-kq_2-q_1} U_{\alpha} U_{\alpha}'. \end{split}$$

Here  $U_{\alpha}$  is distributed according to  $N[0, \Sigma]$  and independently of  $U_{\beta}$   $(\beta \neq \alpha)$ . Therefore  $N\widehat{\Sigma}_{\alpha}$  is distributed according to the central Wishart distribution with  $N-kq_2-q_1$  degrees of freedom.

#### 4. The likelihood ratio criterion

By the preceding consideration, under the null hypothesis, the likelihood ratio criterion  $\lambda$  is given by

$$\begin{split} V &= \lambda^{2/N} = \frac{|N \, \widehat{\Sigma}_g|}{|N \, \widehat{\Sigma}_o|} = \frac{|Y(I-A) \, Y'|}{|Y(I-B-C) \, Y'|} \\ &= \frac{|Y(I-A) \, Y'|}{|Y(I-A) \, Y' + Y(A-B-C) \, Y'|} \, . \end{split}$$

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  $U_{p\cdot(k-1)q_1\cdot N-kq}$  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  $B_1=B_2=B_3=\cdots=B_k$ , we have the criterion

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

whose distribution is that of  $U_{p\cdot(k-1)q\cdot N-kq\cdot n}$ 

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### REFERENCES

- T. W. Anderson, An Introduction to Multivariate Statistical Analysis, John Wiley, New York, 1958.
- [2] Robert V. Hogg, "On the independence of certain Wishart variables," Ann. Math. Statist., 34 (1963), 935-939.