

NOTES

DISTRIBUTION OF THE LIKELIHOOD RATIO FOR TESTING MULTIVARIATE LINEAR HYPOTHESES¹

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1. Introduction. Random orthogonal transformations having elements depending on certain random elements have been used by Wijsman [4] to derive the Wishart distribution and the important statistics such as Hotelling's T^2 . The purpose of this paper is to use these transformations in a simple derivation of the result that the likelihood ratio for testing multivariate linear hypotheses is distributed as the product of q independent Beta variables (cf., Anderson [1], Section 8.5.2). Indirect derivations through the use of moments etc., are given in Wilks [5] and Bartlett [2].

2. Notation and results. Let X be a $q \times r$ matrix of $N(0, 1)$ variables and Y a $q \times s$ matrix ($s \geq q$) of $N(0, 1)$ variables, all variables being independent. Let $A_{q,r} = XX'$, $B_{q,s} = YY'$. In terms of the canonical reduction as given by Hsu [3], it can be shown that the likelihood criterion for testing a general linear hypothesis with r constraints ($r < q$) can be written in the form

$$(1) \quad \Lambda = \frac{|B_{q,s}|}{|A_{q,r} + B_{q,s}|}.$$

If $q = 1$, the problem is trivial. In the following, we shall assume $q > 1$. Denote by x_{ij} , y_{ij} the (i, j) th elements and by x_i , y_i the i th rows of the matrices X and Y .

Let c_1 be the column vector $y'_1 / (y_1 y'_1)^{\frac{1}{2}}$, so that $c'_1 c_1 = 1$, and complete c_1 with $s - 1$ additional columns to an orthogonal matrix $\|c_1: \Omega_B\|$. Following Wijsman [4] we make a random orthogonal transformation from Y to Z ,

$$(2) \quad Z = Y \|c_1: \Omega_B\|.$$

In the first row of Z all elements are 0 except z_{11} , which is equal to

$$(3) \quad z_{11} = (y_1 y'_1)^{\frac{1}{2}}.$$

If the first row and column of Z are deleted, there results a $(q - 1) \times (s - 1)$ matrix V , whose elements are $N(0, 1)$ variables, independent of each other and

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of z_{11} [4]. Furthermore [4],

$$(4) \quad |B_{q,s}| = |YY'| = |ZZ'| = z_{11}^2 |VV'| = z_{11} |B_{q-1,s-1}|,$$

where we have set $B_{q-1,s-1} = VV'$.

Let Ω_A be an $r \times (r - 1)$ matrix whose columns are mutually orthogonal, and orthogonal to x'_1 . Define the following column vectors

$$\begin{aligned} c_2 &= x'_1 / (x_1 x'_1 + y_1 y'_1)^{\frac{1}{2}}, & c_3 &= y'_1 / (x_1 x'_1 + y_1 y'_1)^{\frac{1}{2}}, \\ c_4 &= c_2(c'_3 c_3 / c'_2 c_2)^{\frac{1}{2}}, & c_5 &= -c_3(c'_2 c_2 / c'_3 c_3)^{\frac{1}{2}}, \end{aligned}$$

and transform $\|X:Y\|$ to W with the following orthogonal transformation

$$(5) \quad W = \|X:Y\| \begin{vmatrix} c_2 : c_4 : \Omega_A : O \\ \text{---} \text{---} \text{---} \text{---} \\ c_3 : c_5 : O : \Omega_B \end{vmatrix}.$$

Since the vectors c_2 and c_3 are zero with probability zero, that such a random orthogonal transformation can be chosen measurably follows from the arguments of Wijsman [4]. The elements in the first row of W are 0 except w_{11} , which is

$$(6) \quad w_{11} = (x_1 x'_1 + y_1 y'_1)^{\frac{1}{2}}.$$

It can be easily checked from (5) and (2) that the $(q - 1) \times (r + s - 1)$ matrix T , which results after deleting the first row and column of W , can be written as

$$(7) \quad T = \|U:V\|,$$

where U is the $(q - 1) \times r$ matrix

$$(8) \quad U = \left\| \begin{vmatrix} x_2 & \vdots & y_2 \\ \vdots & \vdots & \vdots \\ x_q & \vdots & y_q \end{vmatrix} \begin{vmatrix} c_4 & \Omega_A \\ c_5 & O \end{vmatrix} \right\|,$$

and V is as defined before. Moreover, the elements of U are $N(0, 1)$, independent of each other, of V , of w_{11} and of z_{11} . Setting $UU' = A_{q-1,r}$, we can write, analogously to (4),

$$(9) \quad |A_{q,r} + B_{q,s}| = \|\|X:Y\|\|X:Y\|'\| = |WW'| = w_{11}^2 |TT'|$$

$$= w_{11}^2 |A_{q-1,r} + B_{q-1,s-1}|.$$

Substitution of (4) and (9) into (1) gives

$$(10) \quad \Lambda = \frac{z_{11}^2}{w_{11}^2} \frac{|B_{q-1,s-1}|}{|A_{q-1,r} + B_{q-1,s-1}|}$$

Using (3) and (6), the first factor, z_{11}^2/w_{11}^2 , on the right-hand side in (10), which we will denote by $\beta_{r/2,s/2}$, is a β -variable with degrees of freedom $r/2$ and $s/2$.

Moreover, the second term is independent of the first. By repeated application of the above procedure, we obtain Λ as the product of q independent β -variables $\beta_{r/2,s}, \beta_{r/2,(s-1)/2}, \dots, \beta_{r/2,(s-q+1)/2}$.

If $r = 1$, i.e., if x is a column vector, we have

$$(11) \quad \Lambda^{-1} = 1 + X'(YY')^{-1}X.$$

Since $X'(YY')^{-1}X$ is Hotelling's T^2 times a constant, equation (11) implies that the product of the q independent Beta variables $\beta_{1/2,s/2}, \beta_{1/2,(s-1)/2}, \dots, \beta_{1/2,(s-q+1)/2}$ is distributed as the reciprocal of one plus a constant times an F variable.

If the null hypothesis is not true, let $E(x_{ij}) = \mu_{ij}$. If the matrix (μ_{ij}) is of rank 1, which really means $r = 1$ (since the multivariate hypothesis is assumed to be in canonical form, we transform X and Y to ξ and η respectively through the relations

$$\xi = MX \text{ and } \eta = MY,$$

where M is a $q \times q$ orthogonal matrix with $(\mu_{11}, \mu_{21}, \dots, \mu_{q1})/(\sum_{i=1}^q \mu_{i1}^2)^{1/2}$ for the first row. Obviously $E(\xi_{11}) = (\sum_{i=1}^q \mu_{i1}^2)^{1/2}$, $E(\xi_{i1}) = 0$ for $i \neq 1$ and $E(\eta) = 0$. By treating ξ and η along the same lines as the matrices X and Y in the above discussion, we obtain equation (10) in which, now, $x_{11}^2 = \eta_1 \eta_1'$,

$$w_{11}^2 = \xi_{11}^2 + \eta_1 \eta_1',$$

and the components U and V of $A_{q-1,r}$ and $B_{q-1,s-1}$ are matrices of independent variables, distributed as $N(0, 1)$. Since ξ_{11} has a non zero mean, we refer to z_{11}^2/w_{11}^2 as a noncentral Beta variable and conclude that Λ is distributed as the product of one noncentral and $q - 1$ central independent Beta variables.

REFERENCES

[1] T. W. ANDERSON, *Introduction to Multivariate Statistical Analysis*, John Wiley and Sons, New York, 1957.
 [2] M. S. BARTLETT, "The vector representation of a sample," *Proc. Camb. Phil. Soc. Edinburgh*, Vol. 30 (1934), pp. 327-340.
 [3] P. L. HSU, "Canonical reduction of the general regression problem," *Ann. Eug.*, Vol. 11 (1941-2), pp. 42-46.
 [4] ROBERT A. WIJSMAN, "Random orthogonal transformations and their use in some classical distribution problems in multivariate analysis," *Ann. Math. Stat.*, Vol. 28 (1957), pp. 415-422.
 [5] S. S. WILKS, "Certain generalizations in the analysis of variance," *Biometrika*, Vol. 24 (1932), pp. 471-494.