

Kendall-type Tests in Multivariate Regression

K. Choi

Hong-Ik University, Department of General Study (Mathematics)

Cho-Chi-Won, Yun-Gi-Gun

Chung-Nam, 339-701, Korea

E-Mail : kmchoi@wow.hongik.ac.kr

1. The Model and Test Statistics

Suppose X_i 's and Y_j 's are row vectors of matrices X and Y respectively. Let the paired data be $(X_1, Y_1), \dots, (X_n, Y_n)$, where $X_i \in R^p$ and $Y_i \in R^q$ for $p \geq 1$ and $q \geq 1$. Then we can consider a model

$$Y = XB/\sqrt{n} + U, \quad (1)$$

where X is an $n \times p$ known matrix, Y is an $n \times q$ observed matrix, and B is an unknown $p \times q$ parameter matrix. And U is a matrix of unobserved random disturbances whose rows for given X are uncorrelated with X , each with mean 0 and common covariance matrix Σ . We want to test $H_o : B/\sqrt{n} = 0$ against $H_A : B/\sqrt{n} \neq 0$. As a measure of association between X and Y , we extend Kendall's τ (Gibbon,1985 ; Choi and Marden, 1997) to a multivariate one as follows:

$$T = \frac{1}{C(n, 2)} \sum \sum_{i < j} b(X_i - X_j)b(Y_i - Y_j)'. \quad (2)$$

The statistic of our main interest based on T in (??) is defined as

$$\hat{\tau} \equiv \text{vec}(T). \quad (3)$$

For comparisons we express the linear model (??) in a vector form (Mardia et al.,1979) as follows :

$$\text{vec}(Y) = X^* \text{vec}(B)/\sqrt{n} + \text{vec}(U), \quad (4)$$

where $X^* \equiv X \otimes I_q$. Then projection matrix P^* on X^* is given by $X^*(X^{*'}X^*)^{-1}X^{*'} = P \otimes I_q$, where $P = X(X'X)^{-1}X'$. Under the null hypothesis the conventional test statistic is

$$LS = \text{vec}(Y)'P^* \text{vec}(Y) \sim \chi_{pq}^2. \quad (5)$$

2. Distributions and Asymptotic Relative Efficiencies

Theorem 1 *Suppose that Σ_X and Σ_Y are not zero, and the distribution of X_i 's is not degenerate. Under the null hypothesis,*

$$\sqrt{n}\hat{\tau} \rightarrow N_{pq}(0, 4\Sigma_X \otimes \Sigma_Y), \quad (6)$$

in distribution. As n goes to infinity, with a consistent and equivariant estimator $\widehat{\text{Cov}}(\hat{\tau})$ of $\text{Cov}(\hat{\tau})$,

$$\hat{\tau}'\widehat{\text{Cov}}(\hat{\tau})^{-1}\hat{\tau} \rightarrow \chi_{pq}^2. \quad (7)$$

In order to obtain the asymptotic relative efficiency (ARE) of $\hat{\tau}$ with respect to LS , we use a probability density function (??) of X_i 's or Y_i 's, which has nice spherical feature and flexibility of the tail shape (Randles, 1989).

$$f(z) = k_p \exp(-[z'z/c_p]^\nu), \quad z \in R^p, \quad (8)$$

where

$$c_p = \frac{p\Gamma\left(\frac{p}{2\nu}\right)}{\Gamma\left(\frac{p+2}{2\nu}\right)}, \quad k_p = \frac{\nu\Gamma\left(\frac{p}{2}\right)}{\Gamma\left(\frac{p}{2\nu}\right)[\pi c_p]^{p/2}}. \quad (9)$$

Here if $\nu > 1$, the pdf has heavy tails. If $\nu < 1$, it has light tails. If $\nu = 1$, it represents the normal distribution.

Theorem 2 Both X_1, \dots, X_n and Y_1, \dots, Y_n are iid from spherically symmetric distributions defined in (??) with their tail shape parameters ν_X and ν_Y respectively. Then

$$ARE(\hat{\tau}, LS) = \frac{1}{4\sigma_X\sigma_Y} \frac{(q-1)^2}{p^2q^2} E^2[\|X_1 - X_2\|] E^2\left[\frac{1}{\|Y_1 - Y_2\|}\right], \quad (10)$$

where $\sigma_X \equiv E\left[\frac{(X_1 - X_2)'(X_1 - X_3)'}{\|X_1 - X_2\| \|X_1 - X_3\|}\right] / p$.

For any $p \geq 2$, $E[\|X_1 - X_2\|]$ is reduced to an integral of two-variable integrand. For any odd $q \geq 3$, $E[1/\|Y_1 - Y_2\|]$ is reduced to an integral of three-variable integrand, so that it can be calculated numerically.

REFERENCES

- Choi, K. and Marden, J. (1997). An approach to multivariate rank transform tests in multivariate analysis of variance, *J. Amer. Statist. Ass.*, **92**, 1581-1590.
- Gibbons, J. D. (1985). *Nonparametric Statistical Inference*, Dekker, Inc.
- Mardia, K.V., Kent, J.T. and Bibby, J.M. (1979). *Multivariate Analysis*. Academic Press.
- Randles, R. H. (1989). A distribution-free multivariate sign test based on interdirections, *J. Amer. Statist. Ass.*, **84**, 1045-1050.