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ASYMPTOTIC DISTRIBUTION OF RANK STATISTICS USED FOR
MULTIVARIATE TESTING SYMMETRY

(Preliminary communication)

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This preliminary communication contains assertions on asymptotic distributions of statistics used for the nonparametric multivariate testing symmetry. The results are proved under the hypothesis of symmetry, a near alternative and a general alternative. The proofs are based on the corresponding theorems for univariate case and the theorem on convergence in distribution for vectors (see Theorem V.2.1 in [5]).

Let $X_j = (X_{j1}, \dots, X_{jn})$, $1 \leq j \leq N$, be independent n -dimensional random variables and let R_{ji}^+ be the rank of $|X_{ji}|$ in the sequence of absolute values $|X_{1i}|, \dots, |X_{Ni}|$. Put

$$S_c = (S_{1c}, \dots, S_{nc})^1,$$

$$S_{ic} = \sum_{j=1}^N c_{ji} a_{Ni}(R_{ji}^+) \operatorname{sgn} X_{ji}, \quad 1 \leq i \leq n,$$

with c_{ji} being regression constants, $a_{Ni}(j)$ scores and

$$\operatorname{sgn} x = \begin{cases} 1 & \text{if } x \geq 0, \\ -1 & \text{if } x < 0. \end{cases}$$

By Σ_{rc} we denote the conditional matrix of S_c given $|X_{ji}| \frac{\text{sgn } X_{ji}}{\text{sgn } X_{j1}}$, $1 \leq j \leq N$, $1 \leq i \leq r$, under (1) given below and by Σ_{rc}^- the generalized inverse of Σ_{rc} (see [4]).

We are interested in an investigation of the asymptotic distribution of the statistics

$$Q_c = S_c' \Sigma_{rc}^- S_c$$

under various systems of conditions.

The problem was solved for example in the papers of Puri and Sen [3], Patel [2] and Adichie [1]. The attention has been devoted to the case $c_{ji} = 1$ or $X_{j1} = X_{j2} = \dots = X_{jn}$.

At first let us consider the following system of conditions for the distribution of X_1, \dots, X_N :

- (1) {
- a) X_1, \dots, X_N are independent;
 - b) $F_{1ik} = \dots = F_{Nik}$, $i \neq k$, $1 \leq i, k \leq r$;
 - c) $F_i(x) = 1 - F_i(-x)$, $1 \leq i \leq r$;
 - d) F_i are continuous;
 - e) $P(\text{sgn } X_{j1} = v_1, \dots, \text{sgn } X_{jn} = v_n) =$
 $= P(\text{sgn } X_{j1} = -v_1, \dots, \text{sgn } X_{jn} = -v_n)$,
 $1 \leq j \leq N$;
- where F_{jik} and F_i , $1 \leq j \leq N$, $1 \leq i \leq r$, are the distribution functions of (X_{ji}, X_{jk}) and X_{ji} , respectively.

These conditions are fulfilled when X_1, \dots, X_N satisfy the multivariate hypothesis of symmetry (definition see [3]).

Let us denote by $D_c = (d_{ik})_{i,k=1,\dots,r}$ the diagonal matrix with

$$d_{ii} = \left(\sum_{j=1}^N c_{ji}^2 \int_0^1 \varphi_i^2(u) du \right)^{-1/2} .$$

Further we shall suppose the covariance matrix Σ_c of S_c under (1) satisfies:

- If $\{D_{c_n}, \Sigma_{c_n}, D_{c_n}\}_{n=1}^{\infty}$ has a limit Σ for (4)
- (2) given below with $c_{ji} = c_{ji}$ then Σ is regular.

On the asymptotic distribution of Q_c under (1) we can state:

Theorem 1. Let (1), (2) and

$$(3) \int_0^1 (a_{Ni} (1 + [uN]) - \varphi_i(u))^2 du \rightarrow 0, \quad 1 \leq i \leq r,$$

where φ_i is squared integrable and $[uN]$ is the largest integer not exceeding uN . Then the statistics Q_c are for

$$(4) \frac{\max_{1 \leq j \leq N} c_{ji}^2}{\sum_{j=1}^N c_{ji}^2} \rightarrow 0, \quad 1 \leq i \leq r,$$

asymptotically χ^2 -distributed with r -degrees of freedom.

Now we turn to another case. Under (6) given below the following conditions ensure that X_1, \dots, X_N "nearly" satisfy the hypothesis of symmetry:

- (5) $\left\{ \begin{array}{l} \text{a) } X_1, \dots, X_N \text{ are independent;} \\ \text{b) } X_{ji} \text{ has a density } f_i(x, \theta_{ji}) \text{ where } \\ \quad \theta_{ji} \text{ is an unknown parameter;} \\ \text{c) } f_i(x, \theta) \text{ is absolutely continuous at} \end{array} \right.$

θ for almost all x , $1 \leq i \leq r$;

$$d) \lim_{\theta \rightarrow 0} \int \frac{\dot{f}_i(x, \theta)^2}{f_i(x, \theta)} dx = \int \frac{(\dot{f}_i(x, 0))^2}{f_i(x, 0)} dx = I(f_i),$$

$$\text{where } \dot{f}_i(x, \theta) = \frac{\partial f_i(x, \theta)}{\partial \theta}, \quad 1 \leq i \leq r;$$

$$e) \lim_{\theta \rightarrow 0} \frac{1}{\theta} (f_i(x, \theta) - f_i(x, 0)) = \dot{f}_i(x, 0) \quad \text{for almost all } x,$$

f) $f_i(x, 0)$ are symmetric about 0, $1 \leq i \leq r$;

g) $F_{ik}(x, y, \theta_{ji}, \theta_{jk})$ is continuous at $\theta_{ji} = \theta_{jk} = 0$ for all x, y , $1 \leq i, k \leq r$, with $F_{ik}(x, y, \theta_{ji}, \theta_{jk})$ being the distribution function of (X_{ji}, X_{jk}) respectively.

Under (5) it can be stated about θ_c ($F_i(x, \theta_{ji})$) denotes the distribution function of X_{ji}):

Theorem 2. Let (5), (3) with φ_i , $1 \leq i \leq r$ being squared integrable, (2) with Σ_c being the covariance matrix of S_c under (1) with $\theta_{ji} = 0$, $1 \leq i \leq r$, $1 \leq j \leq N$, and

$$(6) \max_{1 \leq j \leq N} \theta_{ji}^2 \rightarrow 0, \quad \sum_{j=1}^N \theta_{ji}^2 I(f_i) \leq b^2, \quad 0 < b^2 < +\infty, \quad 1 \leq i \leq r,$$

hold. Then for (4) it holds.

$$\sup_x |P(\theta_c < x) - G_n(x; \mu_{\theta c}, \Sigma_c(\mu_{\theta c}))| \rightarrow 0,$$

where the components of $\mu_{\theta c} = (\mu_{\theta c 1}, \dots, \mu_{\theta c r})'$ are given by

$$\mu_{\theta c i} = \sum_{j=1}^N \theta_{ji} c_{ji} \int \operatorname{sgn} x \varphi_i (F_i(|x|, 0) - F_i(-|x|, 0)) \dot{f}_i(x, 0) dx$$

and where $G_\nu(x, \sigma)$ is the distribution function of noncentral χ^2 -distribution with ν degrees of freedom and noncentrality parameter σ .

At the end the case of a general alternative will be considered. We shall suppose that X_1, \dots, X_N satisfy only the following:

- (7) a) X_1, \dots, X_N are independent;
 b) the distribution function of X_{ji} is continuous.

Let us denote by Σ_c or $\Sigma_c^0 = (G_{ikc}^0)_{i, k = 1, \dots, r}$ the covariance matrix under (7) or the expectation of Σ_{rc} under (7) respectively. Here we shall need also the following notation

$$D_c^0 = (d_{ik}^0)_{i, k = 1, \dots, r},$$

$$d_{ik}^0 = \begin{cases} d_{ii} & \text{if } \varphi_i \text{ satisfies (12), } i = k, \\ \text{var}^0 S_{ic} & \text{if } \varphi_i \text{ satisfies (13) but not (12),} \\ 0 & \text{if } i \neq k, \end{cases}$$

where var^0 denotes var under (7).

Further we shall suppose that Σ_c^0 satisfies:

- (8) { If there exists a matrix $\Xi = (G_{ikc}^0)_{i, k = 1, \dots, r}$ with the property, for every $\varepsilon > 0$ and $\eta > 0$ there exist an $N_{\varepsilon\eta}$ and $d_\varepsilon^0 > 0$ such that the conditions
- (9) { $N > N_{\varepsilon\eta}$, $\text{var} S_{ic} > N\eta \max_{1 \leq j \leq N} c_{ji}^2$ if φ_i satisfies (13) but not (12),

$$\left. \begin{array}{l} \text{var}^{\circ} S_{ic} > \sigma_{\varepsilon}^{-1} \max_{1 \leq j \leq N} c_{ji}^2 \quad \text{if } \varphi_i \\ \text{satisfies (12)} \\ \text{entail} \\ |d_{ii}^{\circ} d_{kk}^{\circ} \sigma_{ikc}^{\circ} - \sigma_{ik}| < \varepsilon, \\ \text{then } \Sigma \text{ is regular.} \end{array} \right\}$$

The condition (7) is weaker than (1) and (5). On the other side we restrict ourselves to scores of the form either

$$(10) \quad \alpha_{Ni}(j) = E \varphi_i(u_N^{(i)}), \quad 1 \leq j \leq N, \quad 1 \leq i \leq r,$$

or

$$(11) \quad \alpha_{Ni}(j) = \varphi_i\left(\frac{j}{N+1}\right), \quad 1 \leq j \leq N, \quad 1 \leq i \leq r,$$

with $u_N^{(i)}$ denoting the i -th order statistics in a sample of size N from the uniform distribution on and with φ_i defined on $(0, 1)$ that either

$$(12) \quad \text{has a bounded second derivative on } (0, 1)$$

or

$$(13) \quad \text{has a form } \varphi_i = \varphi_{1i} = \varphi_{2i}, \text{ where } \varphi_{ki} \text{ is nondecreasing square integrable and absolutely continuous inside } (0, 1).$$

Theorem 3. Let (7) and (8) be satisfied, let the scores be given by (10) or (11) and φ_i , $1 \leq i \leq r$, defined on $(0, 1)$, satisfy the condition (12) or (13). Then for every $\varepsilon > 0$ and $\eta > 0$ there exist an $N_{\varepsilon\eta}$ and a $\sigma_{\varepsilon}^{\circ} > 0$ such that (9) entails

$$\sup_x |P(Q_c < x) - P(U_c' \Sigma_c^{-1} U_c < x)| < \varepsilon,$$

where $U_c = (u_{1c}, \dots, u_{rc})'$ has the normal distribution $(E S_c, \Sigma_c^{\circ})$.

R e f e r e n c e s

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