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A CLASS OF NONPARAMETRIC TESTS FOR LOCATION AND SCALE PARAMETERS^{*)}

by

YVES LEPAGE**)

ABSTRACT

A class of nonparametric two-sample tests for testing identity of distributions versus alternatives containing both location and scale parameters is proposed. The asymptotic distribution of the statistics of the class is given under both the hypothesis and a contiguous sequence of alternatives. Some asymptotic optimality properties are deduced for particular tests of the class and finally, the asymptotic efficiency is found.

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1. INTRODUCTION

The purpose of this paper is to propose and study a class of nonparametric two-sample tests for testing the null hypothesis of randomness versus alternatives containing both location and scale parameters. In section 2, the statistics of the class are defined by a quadratic combination of two linear rank statistics.

Section 3 contains the asymptotic distribution of the statistics of the class under both the null hypothesis and a contiguous sequence of alternatives. In section 4, it is proved that the asymptotically maximin most powerful test belongs to the proposed class. Finally, the asymptotic efficiency of the tests of the class relative to the asymptotically maximin most powerful test is given in section 5.

2. THE CLASS OF TESTS

Let $(m_v, m_v), v = 1, 2, ..., be a sequence of pairs of positive integers$ $such that <math>N_v = m_v + m_v \rightarrow \infty$ when $v \rightarrow \infty$. For each v, consider a sequence of random variables $X_{v1}, ..., X_{vN_v}$ and denote by R_{v1} the rank of X_{v1} among $X_{v1}, ..., X_{vN_v}$.

Suppose that under H_v, the random variables X_{v1}, \ldots, X_{vN_v} are independently and identically distributed according to a continuous distribution and that under the alternatives, the joint density of X_{v1}, \ldots, X_{vN_v} is given by

(2.1)
$$q_{v} = \prod_{i=1}^{m_{v}} e^{-c_{vi}} f(e^{vi}x_{i}-d_{vi}) \prod_{i=m_{v}+1}^{N_{v}} f(x_{i})$$

with $c_{\vee i} = \Delta_1 (m_{\vee} n_{\vee} / N_{\vee})^{-\frac{1}{2}}$ and $d_{\vee i} = \Delta_2 (m_{\vee} n_{\vee} / N_{\vee})^{-\frac{1}{2}}$, $i=1,\ldots,m_{\vee}$, where $(\Delta_1, \Delta_2) \in \mathbb{R}^2$ and a known density f in the class C of absolutely continuous density functions on \mathbb{R} such that

(2.2)
$$I(f) = \int_{0}^{1} \phi^{2}(u,f) du < \infty$$
 and $I_{1}(f) = \int_{0}^{1} \phi_{1}^{2}(u,f) du < \infty$,

where if F(x) is the distribution function corresponding to f(x),

1

(2.3)
$$\phi(u,f) = -\frac{f'(F^{-1}(u))}{f(F^{-1}(u))}$$
 and $\phi_1(u,f) = -1 - F^{-1}(u) \frac{f'(F^{-1}(u))}{f(F^{-1}(u))}$, $0 < u < 1$.

Define for k = 1, 2, the linear rank statistics

(2.4)
$$S_{\nu k} = \sum_{i=1}^{m} a_{\nu k}(R_{\nu i}),$$

where $a_{\nu k}(1), \ldots, a_{\nu k}(N_{\nu})$ are the values of a score function $a_{\nu k}(\cdot)$. Further, let

(2.5)
$$\mu_{\nu k} = E(S_{\nu k}|H_{\nu}) = \frac{m_{\nu}}{N_{\nu}} \sum_{i=1}^{N_{\nu}} a_{\nu k}(i) = m_{\nu} \overline{a}_{\nu k}$$

and

(2.6)
$$\sigma_{\nu k}^2 = \operatorname{Var}(S_{\nu k}|H_{\nu}) = \frac{m_{\nu}n_{\nu}}{N_{\nu}(N_{\nu}-1)} \sum_{i=1}^{N_{\nu}} (a_{\nu k}(i)-\bar{a}_{\nu k})^2.$$

Finally, the class of statistics that we will consider is of the form

(2.7)
$$U_{\nu} = \sum_{k=1}^{2} \left(\frac{S_{\nu k}^{-\mu} \nu k}{\sigma_{\nu k}} \right)^{2}.$$

It should be noted that if $a_{v1}(i) = i$ and $a_{v2}(i) = (N_v+1)/2 - |i-(N_v+1)/2|$, i=1,...,N_v, the statistics U_v reduce to the combined Wilcoxon and Ansari-Bradley statistics studied by LEPAGE (1971).

3. ASYMPTOTIC DISTRIBUTION

A sequence of score functions $a_{v}(\cdot), v = 1, 2, ...$, is said to be generated by a real valued function $\phi(u)$, 0 < u < 1, if

(i)
$$\int_{0}^{1} \phi^{2}(u) \, du < \infty$$
 and $\int_{0}^{1} (\phi(u) - \overline{\phi})^{2} \, du > 0$ where $\overline{\phi} = \int_{0}^{1} \phi(u) \, du$

and,

(ii)
$$\lim_{v \to \infty} \int_{0}^{1} (a_{v}(1+[uN_{v}]) - \phi(u))^{2} du = 0$$
 with $[uN_{v}]$ denoting the largest

integer not exceeding uN...

This condition will be needed to obtain the asymptotic distribution under H_{ij} and q_{ij} of the statistics U_{ij} given by (2.7).

<u>Theorem 3.1</u>. Suppose that for k=1,2, the sequence of score functions $a_{\nu k}(\cdot), \nu = 1, 2, \ldots$, is generated by $\phi_k(u), 0 < u < 1$, with

(3.1)
$$\int_{0}^{0} (\phi_2(u) - \overline{\phi}_2) (\phi_2(u) - \overline{\phi}_2) du = 0$$

Then, under H_v , the statistics U_v are, for $\min(m_v, n_v) \rightarrow \infty$ when $v \rightarrow \infty$, asymptotically χ^2 - distributed with 2 degrees of freedom.

<u>Proof.</u> Let $(\lambda_1, \lambda_2) \in \mathbb{R}^2 - \{(0,0)\}$ and define $T_{\nu} = \sum_{i=1}^{m_{\nu}} a(R_{\nu i})$ where for $i = 1, \dots, N_{\nu}, a_{\nu}(i) = \lambda_{\nu 1} a_{\nu 1}(i) + \lambda_{\nu 2} a_{\nu 2}(i)$ with

(3.2)
$$\lambda_{vk} = \lambda_k (\frac{1}{N_v} \sum_{i=1}^{N_v} (a_{vk}(i) - \bar{a}_{vk})^2)^{-\frac{1}{2}},$$

k=1,2. From HÁJEK & ŠIDÁK (1967), p.163, it follows that under H_v, the statistics T_v are asymptotically normal $(0,\lambda_1^2+\lambda_2^2)$ and consequently, the proof is complete. \Box

<u>Theorem 3.2</u>. Suppose that for k=1,2, the sequence of score functions $a_{\nu k}(\cdot), \nu = 1, 2, ..., is generated by <math>\phi_k(u), 0 < u < 1$, with relation (3.1) satisfied. Then, under q_{ν} , the statistics U_{ν} are, for $\Delta_1 \neq 0$ and $\min(m_{\nu}, n_{\nu}) \rightarrow \infty$ when $\nu \rightarrow \infty$ asymptotically noncentral χ^2 - distributed with 2 degrees of freedom and non-centrality parameter

(3.3)
$$\delta^{2} = \sum_{k=1}^{2} \left[\left(\int_{0}^{1} \phi_{k}(u) (\Delta_{2} \phi(u, f) + \Delta_{1} \phi_{1}(u, f)) du \right)^{2} / \int_{0}^{1} (\phi_{k}(u) - \overline{\phi_{k}})^{2} du \right].$$

<u>Proof</u>. Consider the statistics T_{ν} defined in the proof of theorem 3.1. From theorem 5.1 of LEPAGE (1973b), one can deduce that under q_{ν} , the statistics T_{ν} are asymptotically normal $(\mu_{\nu}, \lambda_1^2 + \lambda_2^2)$ where

(3.4)
$$\mu_{v} = \sum_{k=1}^{2} \left[\lambda_{k} \left(\int_{0}^{1} \phi_{k}(\mathbf{u}) \left(\Delta_{2}\phi(\mathbf{u}, \mathbf{f}) + \Delta_{1}\phi_{1}(\mathbf{u}, \mathbf{f}) \right) d\mathbf{u} \right)^{2} / \int_{0}^{1} \left(\phi_{k}(\mathbf{u}) - \overline{\phi}_{k} \right)^{2} d\mathbf{u} \right].$$

4

Consequently, the proof follows. \Box

4. ASYMPTOTIC OPTIMALITY

Let C₁ be the class of density functions of C such that
(4.1)
$$\int_{0}^{1} \phi(u,f)\phi_{1}(u,f)du = 0.$$

It should be observed that if $f \in C$ is symmetric with respect to the origin, then $f \in C_1$. Further, define the alternatives $K_v(b)$, b > 0, by the joint density q_v given by (2.1) with $f \in C_1$, $\Delta_1 \neq 0$ and

(4.2)
$$\Delta_2^2 I(f) + \Delta_1^2 I_1(f) = b^2$$
.

In the next theorem, an asymptotically optimal test for H_v versus K_v(b) will be found in the class U_v given by (2.7). The (1- α)-quantile of the χ^2 -distribution with k degrees of freedom will be denoted $\chi^2_{k,\alpha}$ while $F_k(\cdot, \delta^2)$ will represent the distribution function of the non-central χ^2 -distribution with non-centrality parameter δ^2 and k degrees of freedom.

<u>Theorem 4.1</u>. Suppose that $\min(m_{v}, n_{v}) \rightarrow \infty$ and that for k=1,2 the sequence of score functions $\mathbf{a}_{vk}(\cdot), v = 1, 2, \ldots$, are generated by $\phi(\mathbf{u}, \mathbf{f})$ and $\phi_{1}(\mathbf{u}, \mathbf{f})$, $0 < \mathbf{u} < 1$, respectively. Then, the test based on \mathbf{U}_{v} with critical region

$$(4.3) \qquad U_{v} \geq \chi^{2}_{2,\alpha}$$

is an asymptotically maximin most powerful test for H_{y} versus $K_{y}(b)$ at level α . Furthermore, the asymptotic power is given by

(4.4)
$$1 - F_2(\chi^2_{2,\alpha}, b^2).$$

<u>Proof</u>. The fact that the asymptotic power of the U_v-test equals $1 - F_2(\chi^2_{2,\alpha}, b^2)$ follows immediately from theorem 3.2. Consequently, if $\beta(\alpha, H_v, K_v(b))$ denotes the power of the maximin most powerful test (among all tests), it remains to prove that

(4.5)
$$\lim_{v \to \infty} \beta(\alpha, H_v, K_v(b)) = 1 - F_2(\chi^2_{2,\alpha}, b^2).$$

Suppose that (4.5) does not hold. Thus, passing to a subsequence if necessary, we may assume that

(4.6)
$$\lim_{v \to \infty} \beta(\alpha, H_v, K_v(b)) > 1 - F_2(\chi^2_{2,\alpha}, b^2).$$

By introducing the random variables

(4.7)
$$\xi_{v1} = \left(\frac{N_v}{m_v n_v}\right)^{\frac{1}{2}} (I(f))^{-\frac{1}{2}} S_{v1} , \quad \xi_{v2} = \left(\frac{N_v}{m_v n_v}\right)^{\frac{1}{2}} (I_1(f))^{-\frac{1}{2}} S_{v2}$$

and the constants

(4.8)
$$\theta_1 = \Delta_2(I(f))^{\frac{1}{2}}$$
, $\theta_2 = \Delta_1(I_1(f))^{\frac{1}{2}}$

one can, in view of theorem 4.1 of LEPAGE (1973b), proceed in a similar way as HÁJEK & ŠIDÁK (1967), p.256-258, and contradict corollary 1 of LEPAGE (1973a).

<u>Corollary 4.1</u>. Under the hypothesis of theorem 4.1, the test based on U_v with critical region (4.3) is an asymptotically uniformly maximin most powerful test for H_v versus $\bigcup K_v(b)$ at level α . b>0

<u>Proof</u>. In view of relation (4.4) and the definition of uniformly maximin most powerful tests (see HÁJEK & ŠIDÁK (1967),p.29), the result is immediate. \Box

Corollary 4.2. In theorem 4.1 and corollary 4.1, the densities q_v given by (2.1) can be replaced by

(4.9)
$$q'_{\nu} = \prod_{i=1}^{m_{\nu}} \exp(-\Delta_{1}(m_{\nu}n_{\nu}/N_{\nu})^{-\frac{1}{2}})f(\exp(-\Delta_{1}(m_{\nu}n_{\nu}/N_{\nu})^{-\frac{1}{2}})(x_{i}-\Delta_{2}(m_{\nu}n_{\nu}/N_{\nu})^{-\frac{1}{2}}))$$



<u>Proof</u>. In view of corollary 5.1 of LEPAGE (1973b), the proof of theorem 4.1 still holds and consequently, corollary 4.1 also. \Box

<u>Corollary 4.3</u>. In theorem 4.1 and corollary 4.1, the densities q_v given by (2.1) can be replaced by

(4.10)
$$q_{\nu,\omega} = \prod_{i=1}^{m_{\nu}} \exp(-\Delta_1 (m_{\nu} n_{\nu} / N_{\nu})^{-\frac{1}{2}} - \omega) f(\exp(-\Delta_1 (m_{\nu} n_{\nu} / N_{\nu})^{-\frac{1}{2}} - \omega) x_i - \Delta_2 (m_{\nu} n_{\nu} / N_{\nu})^{-\frac{1}{2}})$$

$$\sum_{\substack{\nu \\ i=m_{\nu}+1}}^{N_{\nu}} e^{-\omega} f(e^{-\omega}x_{i})$$

where $\omega \in \mathbb{R}$ is unknown. Furthermore, the test based on U_{ij} with critical region given by (4.3) is then an asymptotically uniformly maximin most powerful α level test for H_{ij} versus $U_{ij} K_{ij}(b)$.

<u>Proof</u>. In view of corollary 5.2 of LEPAGE (1973b), the proof of theorem 4.1 still holds and consequently, the result follows. \Box

5. ASYMPTOTIC EFFICIENCY

Let U_{v}^{0} represent the asymptotically maximin most powerful test of theorem 4.1 and consider the tests based on U_{v} given by (2.7) with critical region $U_{v} \ge \chi_{2,\alpha}^{2}$.

<u>Theorem 5.1</u>. Consider testing H_{v} versus q_{v} given by (2.1) with $f \in C_{1}$ and suppose that for k=1,2, the sequence of score functions $a_{vk}(\cdot), v = 1,2,...,$ is generated by $\phi_{k}(u), 0 < u < 1$, with relation (3.1) satisfied. Then, if $\Delta_{1} \neq 0$ and $\min(m_{v}, n_{v}) \neq \infty$ when $v \neq \infty$, the asymptotic efficiency of U_{v} relative to U_{v}^{0} , denoted $e(U_{v}, U_{v}^{0})$, is given by

(5.1)
$$e(U_{v}, U_{v}^{0}) = \sum_{k=1}^{2} \frac{\left(\int_{0}^{1} \phi_{k}(u) (\Delta_{2}\phi(u, f) + \Delta_{1}\phi_{1}(u, f)) du\right)^{2}}{\left(\int_{0}^{1} (\phi_{k}(u) - \overline{\phi}_{\kappa})^{2} du\right) (\Delta_{2}^{2} I(f) + \Delta_{1}^{2} I_{1}(f))}$$

<u>Proof</u>. It follows from theorem 3.2 that the asymptotic power of the test based on U_v is given by $1 - F_2(\chi^2_{2,\alpha}, \Delta^2_2 I(f) + \Delta^2_1(f))$. Thus, in view of relation (3.3) and ANDREWS (1954), the proof is complete. \Box

It should be noted that if $\phi_1(u) = \phi(u, f_1)$ and $\phi_2(u) = \phi_1(u, f_2)$, 0 < u < 1, with $f_1, f_2 \in C_1$ and

(5.2)
$$\int_{0}^{1} \phi(u, f_{1}) \phi_{1}(u, f) du = \int_{0}^{1} \phi(u, f) \phi_{1}(u, f_{2}) du = 0,$$

the asymptotic efficiency given by (5.1) can be written as

(5.3)
$$e(U_{v}, U_{v}^{0}) = \frac{\Delta_{2}^{2}I(f)e(f_{1}, f) + \Delta_{1}^{2}I_{1}(f)e_{1}(f_{2}, f)}{\Delta_{1}^{2}I(f) + \Delta_{1}^{2}I_{1}(f)},$$

where

(5.4)
$$e(f_{1},f) = \frac{\left(\int_{0}^{1} \phi(u,f_{1})\phi(u,f)du\right)^{2}}{I(f_{1})I(f)} \text{ and } e_{1}(f_{2},f) = \frac{\left(\int_{0}^{1} \phi_{1}(u,f_{2})\phi_{1}(u,f)du\right)^{2}}{I_{1}(f_{2})I_{1}(f)}$$

correspond, respectively to the asymptotic efficiency of the test based on S_{v1} where the sequence of score functions $a_{v1}(\cdot), v = 1, 2, ...$, is generated by $\phi(u, f_1)$, 0 < u < 1, relative to the asymptotically most powerful rank test for contiguous location alternatives for a density f and to the asymptotic efficiency of the test based on S_{v2} where the sequence of score functions $a_{v2}(\cdot), v = 1, 2, ...$, is generated by $\phi_1(u, f_2)$, 0 < u < 1, relative to the asymptotically most powerful rank test for contiguous scale alternatives for a density f (see HÁJEK & ŠIDÁK (1967), p.267-270).

Furthermore, from (5.3), it can be seen that for all $(\Delta_1, \Delta_2) \in \mathbb{R}^2$,

(5.5)
$$\min(e(f_1,f),e_1(f_2,f)) \leq (U_v,U_v^0) \leq \max(e(f_1,f),e_1(f_2f)).$$

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