

ON THE ROBUST-EFFICIENCY OF THE COMBINATION OF INDEPENDENT NONPARAMETRIC TESTS*

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1. Summary

Optimum combination of independent one-sample and two-sample nonparametric tests of a general class has been studied by Puri [3], [4], and this unifies the basic work by Elteren [2] and others. The object of the present investigation is to show that the combined tests considered by Puri and Elteren are asymptotically more robust-efficient when the different sets have heteroscedastic distributions.

2. The main results

Let X_{i1}, \dots, X_{im_i} be m_i independent and identically distributed random variables (i.i.d.r.v.) having a continuous cumulative distribution function (c.d.f.) $F_i(x)$, and let Y_{i1}, \dots, Y_{in_i} be a second set of i.i.d.r.v.'s having a continuous c.d.f. $G_i(x)$, for $i=1, \dots, c (\geq 2)$; all the $2c$ samples are assumed to be mutually independent. It is desired to have a combined nonparametric test for the null hypothesis $H_0: G_i(x)=F_i(x)$ for all $i=1, \dots, c$, against $G_i(x)=F_i(x-\theta)$, $\theta \neq 0$. For the i th set, let $N_i=m_i+n_i$ and define the usual Chernoff-Savage [1] type of statistic as

$$(2.1) \quad T_{N,i} = (1/m_i) \sum_{r=1}^{N_i} E_{N_i,r} Z_{N_i,r},$$

where $E_{N,r} = J_N(r/N)$, $1 \leq r \leq N$ satisfies the conditions of Theorem 1 of Chernoff and Savage [1] and $Z_{N_i,r}$ is one or zero according as the r th smallest observation in the combined sample is from an X -observation or not, $r=1, \dots, N_i$, $i=1, \dots, c$. The combined tests are based on the statistics of the type

$$(2.2) \quad T_N = \sum_{i=1}^c h_i T_{N,i}, \quad (N = N_1 + \dots + N_c),$$

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where h_1, \dots, h_c are suitable compounding coefficients. Puri [3] has obtained the locally best test statistic when $F_1 = \dots = F_c = F$, and has shown it to be

$$(2.3) \quad T_N^* = \sum_{i=1}^c m_i T_{N,i}.$$

We shall study the robust-efficiency of T_N^* when F_1, \dots, F_c are not necessarily all identical. For this purpose, let $J(u) = \lim_{N \rightarrow \infty} J_N(u)$, $0 < u < 1$, and assume that it satisfies condition 3 of Theorem 1 of Chernoff and Savage [1]. Define

$$(2.4) \quad A^2 = \int_0^1 J^2(u) du - \left(\int_0^1 J(u) du \right)^2, \quad B(F) = \int_{-\infty}^{\infty} (d/dx) J(F(x)) dF(x).$$

Also, let

$$(2.5) \quad n_i^0 = m_i n_i / N_i, \quad i = 1, \dots, c, \quad N^0 = \sum_{i=1}^c n_i^0 \quad \text{and} \\ \rho_i = n_i^0 / N^0, \quad i = 1, \dots, c.$$

Assume that ρ_1, \dots, ρ_c are all bounded away from zero and one as N^0 (or N) $\rightarrow \infty$.

Then, proceeding precisely on the same line as in Sections 5, 6 and 7 of Puri [3] we arrive at the following theorem.

THEOREM 2.1. *Under the sequence of alternatives $\{H_N: \theta = \theta_N = \lambda / (N^0)^{1/2}, \lambda \text{ real and finite}\}$, $(N^0)^{-1/2} (T_N^* - M \int_0^1 (Ju) du)$ (where $M = m_1 + \dots + m_c$) has asymptotically a normal distribution with mean $\sum_{i=1}^c \rho_i B(F_i)$ and variance A^2 , provided the conditions of Lemma 7.1 of Puri [3] hold for all F_1, \dots, F_c .*

The asymptotically locally optimum parametric test (when $F_1 = \dots = F_c$) is based on the statistic $t^* = \sum_{i=1}^c n_i^0 (\bar{X}_i - \bar{Y}_i)$, where \bar{X}_i and \bar{Y}_i are respectively the sample averages of the X 's and Y 's in the i th set, $i = 1, \dots, c$ (cf. Puri [3]). Under $\{H_N\}$, it is also seen (by some standard computations) that $(N^0)^{-1/2} t^*$ has asymptotically a normal distribution with mean λ and variance $\sum_{i=1}^c \rho_i \sigma_i^2$, where σ_i^2 is the variance of the cdf F_i , $i = 1, \dots, c$. Hence, for $\{H_N\}$, the asymptotic relative efficiency (A. R. E.) of T_N^* with respect to t^* is equal to

$$(2.6) \quad e(T_N^*/t^*) = \left(\sum_{i=1}^c \rho_i B(F_i) \right)^2 / \left(\sum_{i=1}^c \rho_i \sigma_i^2 \right) / A^2.$$

Recalling that for the cdf F_i , the A. R. E. of $T_{N,i}$ with respect to $(\bar{X}_i - \bar{Y}_i)$ is equal to $e_i = B^2(F_i) \sigma_i^2 / A^2$, $i = 1, \dots, c$, (2.6) may be rewritten as

$$(2.7) \quad e(T_N^*/t^*) = \left(\sum_{i=1}^c \rho_i \sqrt{e_i} / \sigma_i \right)^2 \left(\sum_{i=1}^c \rho_i \sigma_i^2 \right),$$

which is a general expression that may be simplified under certain conditions. Before considering these, we present side by side the one sample case.

Let X_{i1}, \dots, X_{im_i} be m_i i.i.d.r.v.'s having a continuous cdf $F_i(x)$, $i=1, \dots, c (\geq 2)$ all the c samples are assumed to be independent. The null hypothesis states that F_1, \dots, F_c are all symmetric about zero, against their symmetry about some non-zero θ . For the i th source, we define the statistic

$$(2.8) \quad T_{m_i} = (1/m_i) \sum_{r=1}^{m_i} E_{m_i, r} Z_{m_i, r},$$

where $E_{m_i, r}$ are also Chernoff-Savage [1] type of rank-scores and $Z_{m_i, r}$ is one or 0 according as the r th smallest observation among $|X_{i1}|, \dots, |X_{im_i}|$ is from a positive X or not, $r=1, \dots, m_i$, $i=1, \dots, c$. For $F_1 = \dots = F_c = F$, the locally asymptotically best statistic is deduced by Puri [4] as

$$(2.9) \quad T_M^* = \sum_{i=1}^c m_i T_{m_i}, \quad M = m_1 + \dots + m_c.$$

Let $\Psi^*(x)$ be a continuous cdf symmetric about 0, and let $\Psi(x) = 2\Psi^*(x) - 1$, $x \geq 0$. Then, if $E_{m, r}$ is the expected value of the r th smallest observation of a sample of size m from the cdf $\Psi(x)$, the corresponding $J(u)$ is $\Psi^{-1}(u) = \Psi^{*-1}((1+u)/2) = J^*((1+u)/2)$, $0 < u < 1$. We then define

$$(2.10) \quad \mu = \int_0^1 J(u) du, \quad A^2 = \int_0^1 J^2(u) du \quad \text{and} \\ B(F) = \int_{-\infty}^{\infty} (d/dx) J^*(F(x)) dF(x).$$

Then, proceeding precisely on the same line as in Puri [4], we arrive at the following theorem by straight-forward generalizations.

THEOREM 2.2. *Under the sequence of alternatives $\{H_M: \theta = \theta_M = \lambda / M^{1/2}, \lambda \text{ real and finite}\}$, $2M^{-1/2}(T_M^* - \mu/2)/A$ has asymptotically a normal distribution with mean $\lambda \sum_{i=1}^c \rho_i B(F_i)/A$ and unit variance (where $\rho_i = m_i/M$, $i=1, \dots, c$), provided F_1, \dots, F_c satisfy the conditions of Lemma 7.1 of Puri [3].*

Under $\{H_M\}$, the parametrically optimum test-statistic $t^* = \sum_{i=1}^c m_i \bar{X}_i$ leads to the following result by virtue of the well-known central limit theorem: $M^{-1/2}t^*$ has asymptotically a normal distribution with mean λ and variance $\sum_{i=1}^c \rho_i \sigma_i^2$. Thus, the A.R.E. of T_M^* with respect to t^* will have the same expression as in (2.6) and (2.7), with the only difference

that A^2 and $B(F)$ in (2.4) and (2.10) are different. Hence, bounds for (2.7) appear to be equally applicable for both the one sample and two sample situations.

We shall specifically simplify (2.7) when F_1, \dots, F_c have the same form but they may differ possibly in scale factors, viz.,

$$(2.11) \quad F_i(x) = F((x - a_i)/\sigma_i), \quad i = 1, \dots, c;$$

where a_1, \dots, a_c are arbitrary locations and $\sigma_1, \dots, \sigma_c$ are scale parameters. Under

$$(2.11) \quad e_i = B^2(F_i)\sigma_i^2/A^2 = B^2(F)/A^2 \quad \text{for all } i = 1, \dots, c,$$

and hence, (2.7) reduces to

$$(2.12) \quad e(T_N^*/t^*) = (B^2(F)/A^2) \left(\sum_{i=1}^c \rho_i/\sigma_i \right)^{-2} \left(\sum_{i=1}^c \rho_i \sigma_i^2 \right).$$

Now by elementary moment-inequalities, we obtain that

$$(2.13) \quad \left(\sum_{i=1}^c \rho_i/\sigma_i \right)^2 \geq \left(\sum_{i=1}^c \rho_i \sigma_i \right)^{-2} \geq \left(\sum_{i=1}^c \rho_i \sigma_i^2 \right)^{-1},$$

where the equality signs hold iff $\sigma_1 = \dots = \sigma_c$. Thus, from (2.12) and (2.13)

$$(2.14) \quad e(T_N^*/t^*) \geq e = B^2(F)/A^2,$$

where the equality signs holds iff $\sigma_1 = \dots = \sigma_c$. This clearly exhibits the robust-efficiency of the combined test by Puri and Elteren when the parent cdf's are heteroscedastic. Moreover, if $B^2(F)/A^2$ is $\geq e^0$ for all F (viz., the normal scores or Wilcoxon scores statistics for which e^0 are respectively 1 and 0.864), it readily follows from (2.7) and (2.13) that $e(T_N^*/t^*)$ will also be $\geq e^0$, uniformly in F_1, \dots, F_c . Thus, the lower bounds for $e(T_N^*/t^*)$, deduced for $F_1 = \dots = F_c = F$, also remain valid when F_1, \dots, F_c are arbitrarily different from each other.

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