

## ON A CONSTRUCTION OF CERTAIN OPTIMUM RANK TESTS

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# ON A CONSTRUCTION OF CERTAIN OPTIMUM RANK TESTS

By

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§ 1. **Introduction.** We have in the previous paper [6] derived a modification of certain well-known rank sum tests in the sense of raising their asymptotic efficiency. For this purpose, the following forms have been proposed as the test statistics

$$(1) \quad mT^k = \sum_{i=1}^N E_{N,i}^k Z_i, \quad k > 0$$

where  $Z_i$  is 1 or 0 if the  $i$ -th smallest in the combined sample is  $X$  or  $Y$  and

$$(2) \quad E_{N,i}^k = \begin{cases} (i/N)^k & \text{for locations problems} \\ [(i/N)^k + ((N+1-i)/N)^k - |(i/N)^k - ((N+1-i)/N)^k|]/2 & \text{for scale problems.} \end{cases}$$

We shall also notice that all notations in this paper are followed by [6]. In [6], the asymptotic normality of the statistics  $T_k$  has been proved and moreover the asymptotic efficiency has been calculated for some  $k > 0$  and some alternatives. We shall deal in this paper with more general statistics among the form (1), i. e.  $E_{N,i}^k$  is some function of rank, and intend to construct the tests with some optimum properties.

§ 2. **Wilcoxon's type.** Our purpose is to test the hypothesis  $\theta = 0$  against the alternative  $\theta > 0$  based on the two samples  $X_1, \dots, X_m$  and  $Y_1, \dots, Y_n$  from the distributions  $F(x)$  and  $F(x + \theta)$ . Then we define the statistics  $U[h]$  as an extension of Wilcoxon's  $U$

$$(3) \quad m\tilde{U}[h] = \sum_{i=1}^N h(i/1+N)Z_i, \quad N = m + n, \quad \lambda_N = m/N$$

, where we assume  $h(t)$  to satisfy the following conditions

$$(i) \quad h(1) = 0(\sqrt{N})$$

$$(4) \quad (ii) \quad |h^{(i)}(t)| \leq K\{t(1-t)\}^{-i-1/2+\delta} \text{ for } i=0, 1, 2 \text{ and some } \delta > 0$$

$$(iii) \quad \int_0^1 h(t) dt = 0, \quad \int_0^1 h^2(t) dt = 1.$$

Assumptions (i) and (ii) are needed for the asymptotic normality of  $\tilde{U}[h]$  and (iii) is only a normalized condition. Following [2], we may prove the asymptotic normality of  $\tilde{U}[h]$  by the same technique as in [6]. Moreover applying the theory of Chernoff-Savage [2], we may get the efficacy  $E^2(\tilde{U})$  of our  $\tilde{U}[h]$  tests if the mean value of  $\tilde{U}$  has a finite derivative under the hypothesis

$$(5) \quad E^2(\tilde{U}) = \lambda(1-\lambda) \left[ \int_{-\infty}^{\infty} f(x) h'[F(x)] dF(x) \right]^2$$

where  $h'[F(x)]$  means the derivative with regard to  $F(x)$ .

If there exists some suitable function  $h(t)$  maximizing the value of  $E^2(\tilde{U})$ , it will give an optimum statistic within the class of our form (3). Then it is sufficient to maximize the integral

$$(6) \quad \begin{aligned} I[h] &= \int_{-\infty}^{\infty} f(x) h'[F(x)] dF(x) \\ &= \int_0^1 f[F^{-1}(t)] h'(t) dt. \end{aligned}$$

(a) Normal case. We may use the method of undetermined multipliers in the calculus of variations to maximize  $I[h]$  subject to the assumption (iii). Thus from

$$\int_0^1 \{h'(t) \varphi[\Phi^{-1}(t)] - \alpha h^2(t)\} dt = \max$$

where  $\Phi$  is the standard normal distribution function with density  $\varphi$  and  $\alpha$  is an undetermined multiplier, we may obtain the following solution of Euler equation for variations,

$$(7) \quad h(t) = \Phi^{-1}(t).$$

The statistic  $\tilde{U}[h]$  with the above  $h(t)$  derives what is called as the Normal scores test that has been shown to be locally most powerful rank test for

the normal alternative by Dwass [3] and others. Thus it has been shown that an optimum test also results from the point of view of our generalization.

(b) Uniform case. From the form (6),

$$I[h] = \lim_{t \rightarrow 1} h(t) - \lim_{t \rightarrow 0} h(t).$$

Thus we must give more weight (positive or negative) at extreme ranks. Now we assume  $h(t)$  to be  $\Phi^{-1}(t)$  for a time, then the expectation  $\mu(\theta)$  has not a finite derivative  $\mu'(\theta)$  at  $\theta=0$ . On the other hand, the fact that the efficiency of  $\tilde{U}_k$  tests approach to infinity when  $k \rightarrow 0$  as has been shown in [6] leads us to the following consideration. Construct the normalized function  $h_0(t)$  corresponding to the case  $k \rightarrow 0$  in [6] as follows,

$$(8) \quad h_0(t) = \lim_{k \rightarrow 0} \frac{(k+1)\sqrt{2k+1}}{k} \left( t^k - \frac{1}{k+1} \right) \\ = \log t + 1.$$

However our test statistic

$$(9) \quad m\tilde{U}[h_0] = \sum_{i=1}^N \left( \log \frac{i}{N+1} + 1 \right) Z_i$$

has not also a mean value with a finite derivative at  $\theta=0$ . For these cases, we may apply the theory of Hodges-Lehmann [4] that is a slight generalization of Pitman's efficiency.

Consider the sequences of test statistics  $\{S_N\}$ ,  $\{T_N\}$  to test the hypothesis  $\theta=\theta_0$  whose expectations

$$\mu_N(\theta) = E_\theta(S_N), \quad \nu_N(\theta) = E_\theta(T_N)$$

and suppose that

$$[S_N - \mu_N(\theta)]/b_N \quad \text{and} \quad [T_N - \nu_N(\theta)]/c_N$$

tend in law to  $N(0, 1)$  whenever  $\theta \rightarrow \theta_0$ . Let the sample sizes necessary to achieve the same power  $\beta$  against the same alternative at the same significance level  $\alpha$  be respectively  $r_N$  and  $N$ . Then it has been proved in [4] that

$$(10) \quad \lim_{N \rightarrow \infty} \frac{\nu_N(\theta) - \nu_N(\theta_0)}{\mu_N(\theta) - \mu_N(\theta_0)} \cdot \frac{b_{r_N}}{c_N} = 1$$

Now let

$$mS_N = m\tilde{U}[h_0] = \sum_{i=1}^N \left( \log \frac{i}{N+1} + 1 \right) Z_i$$

$$mT_N = \bar{\Phi}^{-1} \left( \frac{i}{N+1} \right) Z_i.$$

We may get from (10) the asymptotic efficiency  $e_{L,N}$  of  $\tilde{U}[h_0]$  test relative to the Normal scores test

$$(11) \quad e_{L,N} = \lim_{N \rightarrow \infty} \frac{N}{r_N} = \left( \lim_{N \rightarrow \infty} \frac{[\Phi^{-1}\{\lambda_N F(x) + (1-\lambda_N)F(x+\theta)\} - \Phi^{-1}\{F(x)\}]dF(x)}{[\log\{\lambda_N F(x) + (1-\lambda_N)F(x+\theta)\} - \log F(x)]dF(x)} \right)^2.$$

After some easy calculations of integration and  $L'$  Hospital rule, we obtain

$$(12) \quad e_{L,N} = \left( \lim_{N \rightarrow \infty} \left[ \frac{1}{\theta} + \frac{\lambda_N}{1-\lambda_N} \right] / \left[ \frac{1-\lambda_N}{\varphi[\Phi^{-1}\{(1-\lambda_N)\theta\}]} + \frac{\lambda_N}{\varphi[\Phi^{-1}(1-\lambda_N\theta)]} \right] \right)^2 = \infty$$

Thus it follows that for the uniform alternative (and also exponential case, see (c).) our test with  $h_0(t)$  is very much more efficient than the Normal scores test. The similar considerations lead us that the Normal scores test has infinite efficiency with regard to  $U_k$  tests with non-zero fixed  $k$  (Hodges-Lehmann have dealt with only  $k=1$ ).

(c) Exponential case. Since we get the same results as the uniform case, we omit them.

**§ 3. Ansari-Bradley's type.** In this section, we concern with the scale problems and generalize Ansari-Bradley's statistic. We first define the statistics  $\tilde{S}[h]$

$$(13) \quad m\tilde{S}[h] = \sum_{i=1}^N J_{N,i} Z_i$$

where  $J_{N,i} = \left[ h\left(\frac{i}{N+1}\right) + h\left(\frac{N+1-i}{N+1}\right) - \left| h\left(\frac{i}{N+1}\right) - h\left(\frac{N+1-i}{N+1}\right) \right| \right] / 2.$

We assume that  $h(t)$  and  $J_{N,i}$  respectively satisfy the assumption (17) and the regularity conditions of Chernoff-Savage. Then its asymptotic normality is established and the efficacy is expressed if the expectation of  $\tilde{S}[h]$  has a finite derivative under the hypothesis

$$(14) \quad E^2(\tilde{S}) = \left[ \int_{-\infty}^0 x f(x) h'[F(x)] dF(x) - \int_0^{\infty} x f(x) h'[1-F(x)] dF(x) \right]^2.$$

When  $F(x)$  is symmetrical, the expression  $I[h]$  in bracket of (14) becomes

$$(15) \quad I[h] = 2 \int_0^{1/2} h'(t) F^{-1}(t) f[F^{-1}(t)] dt.$$

(a) Uniform case. Easy computation shows that

$$I[h] = -2 \lim_{t \rightarrow 0} h(t).$$

Though both  $h(t) = \Phi^{-1}(t)$  and  $h_1(t) = \log 2t + 1$  give very much large weight at extreme ranks, we prefer  $h_1(t)$  to the other  $h(t) = \Phi^{-1}(t)$  as the similar results as in section 2. However it may be possible to exist more favourable  $h(t)$  than  $h_1(t)$ .

(b) Double exponential case. Since the density is

$$f(x) = \frac{1}{2} \exp(-|x|),$$

we may get

$$(16) \quad I[h] = 2 \int_0^{1/2} t h'(t) \log 2t \, dt$$

which must be maximized subject to the normalized restrictions

$$(17) \quad \int_0^{1/2} h(x) dx = 0, \quad \int_0^{1/2} h^2(x) dx = 1/2.$$

Following the method of variations, we may determine the form (18) as the solution of Euler equation,

$$(18) \quad h(t) = \log 2t + 1.$$

This form may be also attained from our  $S_k$  tests in [6] by the similar calculation as (8). Capon [1] has also derived the same form by another point of view and proved to be locally best for the double exponential case.

(c) Normal case.  $I[h]$  may be expressed as follows,

$$(19) \quad I[h] = 2 \int_0^{1/2} h'(t) \Phi^{-1}(t) \varphi[\Phi^{-1}(t)] \, dt.$$

Euler equation under the conditions for  $h(t)$  is given as

$$(20) \quad h(t) - \frac{d}{dt} \Phi^{-1}(t) \varphi[\Phi^{-1}(t)] = 0$$

and it is solved as

$$(21) \quad h(t) = \frac{1}{\sqrt{2}} [\Phi^{-1}(t)^2 - 1].$$

The corresponding statistic  $\tilde{S}[h]$  has thus an optimum property and it has been shown to be locally best rank test for the normal case by Klotz [5].

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