## ON THE CONSISTENCY OF THE TWO-SAMPLE EMPTY CELL TEST

M. Csorgo \* and Irwin Guttman

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1. Introduction. This paper considers the consistency of the two-sample empty cell test suggested by S. S. Wilks [2]. A description of this test is as follows: Let a sample of  $\mathbf{n}_1$  independent observations be taken from a population whose cumulative distribution function  $\mathbf{F}_1(\mathbf{x})$  is continuous, but otherwise unknown. Let  $\mathbf{X}_{(1)} < \mathbf{X}_{(2)} < \ldots < \mathbf{X}_{(n_1)}$  be their order statistics. Let a second sample of  $\mathbf{n}_2$  observations be taken from a population whose cumulative distribution function is  $\mathbf{F}_2(\mathbf{x})$ , assumed continuous, but otherwise unknown.

Define cells 
$$I_1, \ldots, I_{n_1+1}$$
 by

(1.1) 
$$I_{i} = (X_{(i-1)}, X_{(i)}], i = 1, ..., n_{1} + 1,$$

where 
$$X_{(0)} = -\infty$$
 and  $X_{(n_4+1)} = +\infty$ .

Let  $r_1, \ldots, r_{n_1+1}$  be the number of observations of the second sample that lie in  $I_1, \ldots, I_{n_1+1}$  respectively.

Let  $S_0$  be the number of  $I_1$ ,  $i=1,\ldots,n_1+1$  which are

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such that  $r_i = 0$ , that is, the number of empty cells. Under the hypothesis that  $F_1 = F_2$ , Wilks in [2] and [3] gives a somewhat complicated analytic derivation of the probability function of  $S_0$  and obtains the result

(1.2) 
$$P(S_{o} = S_{o}) = \frac{\binom{n_{1} + 1}{s_{o}} \binom{n_{2} - 1}{n_{1} - s_{o}}}{\binom{n_{1} + n_{2}}{s_{o}}} = p(S_{o})$$

where the sample space of  $S_{o}$  is given by

$$\mathcal{S} = [k, k+1, ..., n_1]$$
 and  $k = \max[0, n_1+1 - n_2]$ .

A simplified proof of (1.2) may be found in [4].

Using (1.2), it can be easily shown that

$$E(S_0) = \frac{n_1(n_1 + 1)}{n_1 + n_2}$$

(1.3) 
$$\sigma^{2}(S_{0}) = \frac{n_{1}^{2}(n_{1}^{2} - 1)}{(n_{1}^{+}n_{2}^{-})(n_{1}^{+}n_{2}^{-}1)} + \frac{n_{1}^{2}(n_{1}^{+}1)}{n_{1}^{+}n_{2}^{-}} - \frac{n_{1}^{2}(n_{1}^{+}1)^{2}}{(n_{1}^{+}n_{2}^{-})^{2}}$$

(For these results see Wilks [2] and [3] where the method of factorial moments is used to obtain them.)

If we let  $n_2 = \rho n_1 + O(1)$ ,  $\rho > 0$ , these reduce to

$$E(S_o) = n_1 \left( \frac{1}{1+\rho} \right) + O(\frac{1}{n_1})$$

$$\sigma^2(S_o) = n_1 \left( \frac{\rho^2}{(1+\rho)^3} + O(\frac{1}{n_1}) \right)$$

which in turn imply that

$$E\left(\frac{S_0}{n_1+1}\right) \to \frac{1}{1+\rho}$$

and

$$\sigma^2\left(\frac{S_0}{n_1+1}\right) \to 0$$

as  $n_1$ ,  $n_2 \to \infty$ , and by Tchebychev's inequality, these results imply that  $S_0/n_1+1$  converges in probability to  $\frac{1}{1+\rho}$ , if  $F_1=F_2$ .

We can use these results to make a test of the hypothesis  $F_4 = F_2$  at the approximate  $100\alpha$  % level. This is given by

(1.5) 
$$\begin{cases} \text{Reject if } s \geq b \\ \text{O } \end{cases}$$
 Accept otherwise

where b is such that

$$P(S_o \ge b) = \sum_{\substack{s_o = b \\ s_o = b}} P(s_o) \le \alpha$$

$$(1.6)$$

$$P(S_o \ge b-1) = \sum_{\substack{s_o = b-1 \\ s_o = b-1}} P(s_o) > \alpha.$$

Tables of (1.6) have been tabulated by the authors for  $\alpha = .01$  and .05 and published in Technometrics [4].

2. Consistency. The form of the test (1.5) follows from the following considerations. Let G be the class of pairs of

continuous cumulative density functions  $(F_1(x), F_2(x))$  such that  $F_1(x) = F_2(x)$ . Let  $F_1^{-1}(u)$  be the inverse of the c.d.f.  $F_1(x)$  and let  $G_1$  be the class of pairs of continuous c.d.f.'s  $(F_1(x), F_2(x))$  satisfying:

- (i)  $F_2(F_1^{-1}(u))$  has a derivative, say g(u), for all u on (0, 1) except possibly for a set of probability measure zero.
- (ii) The derivatives of  $F_2(F_1^{-1}(u))$  and  $F_1(F_1^{-1}(u)) = u$  with respect to u on (0, 1) differ over a set of positive probability.

## In [3] Wilks states the following

THEOREM. The test defined by (1.5) and (1.6) is consistent for testing any  $(F_1, F_2) \in G_0$  against any  $(F_1, F_2) \in G_1$  as  $n_1, n_2 \to \infty$  so that  $n_2 = n_1 \rho + O(1)$ , where  $\rho > 0$ .

To prove this theorem it is sufficient to show that if  $(F_1, F_2) \in G_1$ ,  $S_0/(n_1+1)$  converges in probability to a number greater than  $1/(1+\rho)$  as  $n_1$ ,  $n_2 \to \infty$  with  $\frac{n_2}{n_1} \to \rho > 0$ , for it will be recalled from (1.4) that  $1/(1+\rho)$  is the quantity to which  $S_0/(n_1+1)$  converges in probability if  $(F_1, F_2) \in G_0$ .

We recall that  $r_1, \ldots, r_{n_1+1}$  denote the number of observations of the second sample that lie in the  $n_1+1$  cells  $I_1, \ldots, I_{n_1+1}$  respectively. For each non-negative integer r, let  $Q_1(r)$  be the proportion of values among  $r_1, \ldots, r_{n_1+1}$  which are equal to r. Then, in particular, we have  $Q_{n_1}(0) = \frac{S}{n_1+1}, \text{ the proportion of empty cells.}$ 

Under the conditions (i) and (ii) of this section, J. R. Blum and L. Weiss in [1] prove that

(2.1) 
$$P\left[\lim_{(n_1, n_2; \rho)} \sup_{r \ge 0} |Q_n(r) - Q(r)| = 0\right] = 1$$

where  $\lim_{(n_4,n_2;p)}$  denotes the limit as  $n_4 \to \infty$ ,  $n_2 \to \infty$  in such

a way that  $n_2/n_1 \rightarrow \rho$ ,  $\rho > 0$ , and

(2.2) 
$$Q(r) = \rho^{r} \int_{0}^{1} \frac{g^{2}(u)}{[\rho + g(u)]^{r+1}} du$$

where g(u) is the derivative of  $F_2(F_1^{-1}(u))$ , satisfying conditions (i) and (ii) of this section.

As a special case of (2.1) we have that

(2.3) 
$$P\left[\lim_{(n_{1}, n_{2}; \rho)} \left| Q_{n_{1}}(0) - Q(0) \right| = 0\right] = 1$$

if  $(F_4, F_2) \in G_4$ , where we have now that

(2.4) 
$$Q(0) = \int_{0}^{1} \frac{g^{2}(u)}{[\rho + g(u)]} du.$$

It is also implied by (2.3) that

(2.5) 
$$\lim_{\substack{(n_1, n_2; \rho)}} P(|Q_n(0) - Q(0)| \ge \epsilon) = 0$$

for any  $\epsilon > 0$ , however small, if  $(F_1, F_2) \in G_1$ ; that is

$$Q_{n_1}(0) = \frac{s_0}{n_1+1} \text{ converges in probability to } Q(0) \text{ (expression (2.4))}.$$

Therefore, the test defined by (1.5) and (1.6) is consistent for testing any  $(F_1, F_2) \in G_0$  against any  $(F_1, F_2) \in G_1$  if

(2.6) 
$$\int_{0}^{1} \frac{g^{2}(u)}{[\rho + g(u)]} du > \frac{1}{1+\rho},$$

where we recall from (1.4) that  $1/1+\rho$  is the quantity to which  $S \\ Q_{n_4}(0) = \frac{o}{n_4+1}$  converges in probability if  $(F_1, F_2) \in G_o$ .

The inequality of (2.6) is proved as follows. We have by Schwarz's inequality that

$$(2.7) \int_{0}^{1} \frac{g^{2}(u)d u}{\rho + g(u)} \int_{0}^{1} (\rho + g(u)) d u > \left\{ \int_{0}^{1} \frac{g(u)}{\sqrt{\rho + g(u)}} \sqrt{\rho + g(u)} d u \right\}^{2}$$

that is

$$\left(\int_{0}^{1} \frac{g^{2}(u)d u}{\rho + g(u)}\right) (\rho + 1) > 1$$

which gives

$$\int_{0}^{1} \frac{g^{2}(u)d u}{\rho + g(u)} > \frac{1}{1 + \rho} ,$$

if g(u) differs from unity over a set of positive probability. This condition obtains if  $(F_1, F_2) \in G_1$ , since the derivatives of  $F_2(F_1^{-1}(u)) = g(u)$  and u are assumed to differ over a set of positive probability on (0,1), and under this condition the above strict Schwarz inequality (2.7) holds. This completes the proof of the above theorem.

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McGill University and University of Wisconsin