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## On Greenwood goodness-of-fit test

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#### 1. Introduction

# Let $X_{1,n} \le X_{2,n} \le ... \le X_{n-1,n}$ be the order statistics of a sample of size n-1 drawn from a continuous distribution concentrated on the interval [0,1]. The disjoint *s*-spacings are defined by $W_{k,n}^{(s)} = X_{ks,n} - X_{(k-1)s,n}$ , k=1,2,...,N', $W_{N'+1,n}^{(s)} = 1 - X_{N's,n}$ , with notation $X_{0,n} = 0$ and $X_{n,n} = 1$ and $N' = \lfloor n/s \rfloor$ is the largest integer that does not exceeds n/s. We assume that *s* may increase together with *n* and s = o(n).

The goodness-of-fit problem is to test if the sample drawn from Uniform [0,1] distribution. A large class of the tests of uniformity based on spacings has the general form

$$S_{N}^{f} = \sum_{k=1}^{N} f(nW_{k,n}^{(s)}),$$
(1.1)

where  $N = \lceil n/s \rceil$  is the smallest integer greater than or equal to n/s and f(u) is a real-valued function. A test based on the statistic  $S_N^f$  is called, for brevity, *f*-test. Three most popular cases of  $S_N^f$  are

 $G_N^2 = \sum_{k=1}^N (nW_{k,n}^{(s)})^2,$ 

#### ABSTRACTS

The Greenwood test is based on the sum of squares of disjoint spacings. For the goodness-of-fit problem within the class of symmetric tests, the Greenwood test is known to be optimal in terms of the Pitman asymptotic efficiency (AE), whereas it is much inferior to the log-spacings test in terms of the Bahadur AE. In this paper we extend these properties of Greenwood test showing that it remains optimal in terms of the weak intermediate and intermediate AE, but it is much inferior to those tests satisfying the Cramer condition in terms of the strong intermediate AE.

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called Greenwood's statistic

$$L_N = \sum_{k=1}^N \log(nW_{k,n}^{(s)}),$$

called log-spacings statistic, and

$$R_N = \sum_{k=1}^N \left| n W_{k,n}^{(s)} - s \right|,$$

called Rao's statistic.

Considerable attention has been devoted in literature to statistics of type (1.1). For detailed information, applications, and references, consult, the papers by Greenwood (1946), Darling (1953), Le Cam (1958), Pyke (1965, 1972), Rao (1972), Rao and Sethuraman (1975), Deheuvels (1985), Jammalamadaka et al. (1989), Zhou and Jammalamadaka (1989), Ghosh and Jammalamadaka (2000), Jammalamadaka and Goria (2004), Mirakhmedov (2005, 2006), Gatto and Jammalamadaka (2006), Mirakhmedov and Naeem (2008a, b) and Mirakhmedov et al. (2009).

We are interested here in asymptotic efficiencies (AE) of *f*-tests, giving most attention to the Greenwood test. There are several concepts of AE, which differ by conditions imposed on the asymptotic behavior of the size  $\alpha_n$ , the power  $\beta_n$  and the alternative  $H_1$ . The most celebrated is Pitman concept, in which we deal with a sequence of alternatives  $H_{1,n}$  converging to  $H_0$  at a rate necessary to keep  $\alpha_n \rightarrow \alpha$ ,  $\beta_n \rightarrow \beta$ ,  $0 < \alpha < \beta < 1$ , as  $n \rightarrow \infty$ . This is a family of Pitman alternatives (we denote it as  $P_{alt}$ ). Under the Bahadur concept, a family of fixed alternatives (more precisely, the alternatives that do not approach the hypothesis)  $B_{alt}$  is considered; it is supposed that the asymptotic power  $\beta < 1$  and, as consequence, AE of a test is characterized by the *exponential* rate of decrease of the size  $\alpha_n$ . Thus, the Pitman and Bahadur concepts apply one of the following principles: move the alternatives steadily closer to the hypothesis (the Pitman concept) or decrease the size of a test to zero (the Bahadur concept). The intermediate AE concept, as defined by Kallenberg (1983), (see also lvchenko and Mirakhmedov, 1995; Inglot, 1999), applies both principles simultaneously: it is assumed that  $\beta_n \rightarrow \beta$ ,  $0 < \beta < 1$ , while simultaneously  $H_{1,n} \rightarrow H_0$  (more slowly than in the Pitman case) and  $\alpha_n \rightarrow 0$  (more slowly than in the Bahadur approach). This intermediate approach is somewhere between of Pitman and Bahadur approaches of efficiency.

AE of *f*-tests was studied by several authors. It has been proved that the Pitman AE of *f*-test depends on an asymptotic correlation  $\mu(s_f)$  (see below, relation (2.6)) of the statistic  $S_N^f$  with Greenwood statistic (see Del Pino, 1979; Rao and Kuo, 1984; Jammalamadaka and Tiwari,1987; Jammalamadaka et al., 1989; Mirakhmedov and Naeem, 2008b) and hence, Greenwood test is optimal in terms of Pitman AE. Zhou and Jammalamadaka (1989) have shown that Greenwood test is much inferior to the log-spacings test in terms of Bahadur AE. Finally, Mirakhmedov (2006) and Mirakhmedov and Naeem (2008b) show that the intermediate AE of *f*-tests satisfying Cramér condition  $E \exp\{H|f(Z)|\} < \infty$ ,  $\exists H > 0$ , where *Z* is a random variable with p.d.f.  $\gamma_s(u)=u^{s-1}e^{-u}/\Gamma(s)$ , also depend on  $\mu(s_f)$ . Note that the log-spacings and Rao's statistics satisfy the Cramér condition, whereas Greenwood statistic does not. Thus, the intermediate AE of Greenwood test has been an open problem. In the present paper we address that problem.

The concept of intermediate AE is based on moderate and large deviation theorems for the test statistics. Such theorems for the statistics  $S_N^f$  are presented in Section 3, whereas the main results on AE are discussed in Section 2. For the reader's convenience, the auxiliary assertions are collected in Appendix.

In what follows,  $C_i$  is a positive universal constant, and all asymptotic relations are considered as  $n \rightarrow \infty$ .

#### 2. Asymptotic efficiency

Consider testing the hypothesis  $H_0$  of uniformity of the n-1 independent observations on [0,1]. Under the alternative hypothesis  $H_1$ , we specify the p.d.f. to be

$$p_n(x) = 1 + dl(x)\delta(n), \quad 0 \le x \le 1,$$
(2.1)

where  $\delta(n) \rightarrow 0$  as  $n \rightarrow \infty$ , d > 0 is a constant and l(x) is a function such that

$$\int_{0}^{1} l(x)dx = 0, \quad \int_{0}^{1} l^{2}(x)dx = 1.$$

Assume that the large values of the statistics of type (1.1) reject  $H_0$ . Let  $P_i$ ,  $E_i$ ,  $Var_i$  and  $corr_i$  be the probability, expectation, variance and correlation coefficient, respectively, under  $H_i$ ;  $A_{i,N}$  stand for the asymptotic value of  $N^{-1}E_iR_N^f$ , i=0,1, and Z stands for a random variable (r.v.) with p.d.f.  $\gamma_s(u)=u^{s-1}e^{-u}/\Gamma(s)$ .

Under the alternative (2.1) (see, for example, Mirakhmedov and Naeem, 2008b)

$$Var_1S_N^f = Var_0S_N^f(1+o(1)) = N\sigma^2(f)(1+o(1)),$$

$$x_N(f) \stackrel{def}{=} \sqrt{N} (A_{1N}(f) - A_{0N}(f)) / \sigma(f) = \sqrt{\frac{n(s+1)}{2}} \delta^2(n) \mu(s, f) d^2(1 + o(1)),$$
(2.2)

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