A NOTE ON SOME TEST STATISTICS AGAINST HNBUE

Bo Bergman and Bengt Klefsjö*

(Received Jan. 27, 1986; revised Feb. 27, 1987)

Summary

Some comments and extensions are given to the paper on statistics for testing exponentiality against the HNBUE property \( \int_{0}^{\infty} F(x)dx \leq \mu \exp(-t/\mu), t \geq 0 \) by Basu and Ebrahimi (1985, Ann. Inst. Statist. Math., 37, 347–359).

1. Introduction and comments

A life distribution \( F \) with survival function \( \bar{F} = 1 - F \) and finite mean \( \mu = \int_{0}^{\infty} F(x)dx \) is said to be harmonic new better (worse) than used in expectation, HNBUE (HNWUE) if \( \int_{0}^{\infty} F(x)dx \leq (\geq) \mu \exp(-t/\mu), t \geq 0 \). Basu and Ebrahimi [2] discuss some test statistics based on a complete ordered sample \( 0 = t(0) \leq t(1) \leq t(2) \leq \cdots \leq t(n) \) from a life distribution \( F \) for testing

\[ H_0: \text{F is exponential (i.e. } \bar{F}(t) = \exp(-\lambda t), t \geq 0) \]

against

\[ H_1: \text{F is HNBUE (but not exponential)} \]

or

\[ H_2: \text{F is HNWUE (but not exponential)} \]

For further notations and details see Basu and Ebrahimi [2] and for properties of HNBUE (HNWUE) life distributions see Klefsjö [9].

The authors present two test statistics denoted by \( A_n \) and \( B_n \). The statistic \( A_n \) is scale dependent and rather complicated in different ways.

* This research was partially supported by Swedish Natural Science Research Council Port Doctorial Fellowship F-PD 1564-101.

Key words and phrases: HNBUE, HNWUE Pitman efficacy value, power.
and will not be discussed here. The statistic

\[ B_n = \sum_{j=1}^{n-1} \ln (1 - S_j) + n, \]

where \( S_j \) is the scaled total time on test (TTT-) statistic at time \( t(j) \), is simpler and independent of scale and accordingly more interesting. In fact \( B_n \) is a linear function of the test statistic

\[ A_i = -2 \sum_{j=1}^{n-1} \ln (1 - S_j) \]

which was discussed by Klefsjö [8] in a partly unpublished report. Also Kochar and Deshpande [13] suggested a linear function of \( B_n \), namely \( \left( \sum_{j=1}^{n-1} \ln (1 - S_j) \right)/n \), as a test statistic for testing exponentiality against HNBUE (but they mainly discuss the consistency problem).

These test statistics are closely related to the test statistic \(-2 \sum_{j=1}^{n-1} \ln S_j\) (discussed e.g. by Epstein [6]), and widely used as an omnibus test for exponentiality.

One reason for studying \( B_n \), or equivalently \( A_i \), is that it is rather easy to get the percentiles under \( H_0 \). Since \( (S_1, S_2, \ldots, S_{n-1}) \) has the same distribution as an ordered sample from a uniform distribution on \([0, 1]\) (see e.g., Barlow and Proschan [1]), it follows that \( A_i \) is distributed according to a \( \chi^2(2(n-1)) \)-distribution. Accordingly we can, for larger values of \( n \), use that \( \sqrt{2 \chi^2(k)} - \sqrt{2k-1} \) is asymptotically \( N(0, 1) \)-distributed (see e.g., Johnson and Kotz [7], p. 176), which is a better approximation than the usual approximation that \( \{ \chi^2(k) - k \}/\sqrt{2k} \) is asymptotically \( N(0, 1) \)-distributed.

Let \( \chi^2_i(k) \) be defined by \( P(\chi^2 > \chi^2_i(k)) = \alpha \), where \( \chi^2 \) is \( \chi^2(k) \)-distributed. The critical values \( \alpha \) of \( B_n \), where \( P(B_n > \alpha) = \alpha \), can then be obtained as

\[ \alpha = \frac{1}{2} \left( 2n - \chi^2_{1-\alpha}(2(n-1)) \right). \]

They seem not to be aware of this judging from Table 1 in Basu and Ebrahimi [2] on percentiles of \( B_n \) based on numerical integration.

Basu and Ebrahimi [2] present in Table 2 power estimates for \( n = 20 \) against some different Weibull and gamma alternatives for \( A_n \), \( B_n \) and two test statistics \( A_i \) and \( B \) presented in detail in Klefsjö [11] and intended for testing exponentiality against increasing failure rate (IFR) and increasing failure rate in average (IFRA), respectively. The paper does not refer if they use percentiles from an asymptotic normal distribution or not and the power estimates presented seem to be inappropriate. In Klefsjö [8] the same power values (among others) were
estimated for $A_s$ using $\chi^2$-percentiles with quite different results. Also the case $n=10$ was studied by Klefsjö [8]. To be sure we recently made a new power simulation. All these power estimates can be found in Table 1. In that table also the power estimates for $B$ presented in Basu and Ebrahimi [2] and Klefsjö [11] are included. The simulations in Klefsjö [11] are based on the exact percentiles which differ from the asymptotic percentiles from a normal distribution. The power estimates by Basu and Ebrahimi [2] and Klefsjö [8], [10] all differ more than 0.03 which is alarming since two independent power estimates based on 2000 replications should be within ±0.03 with high probability.

Some deviations can be seen in the results by Basu and Ebrahimi [2] in the light of the asymptotic Pitman efficacy values for $B_n$ (or $A_s$) and $B$; see Table 2.

Table 2. Asymptotic Pitman efficacy values of $B_n$ and $B$.

<table>
<thead>
<tr>
<th></th>
<th>$B_n$ (or $A_s$)</th>
<th>$B$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Weibull</td>
<td>1.00</td>
<td>1.25</td>
</tr>
<tr>
<td>Gamma</td>
<td>0.25</td>
<td>0.50</td>
</tr>
</tbody>
</table>

Basu and Ebrahimi [2] mention the scaled TTT-transform as a basis for their idea to get the test statistics $A_n$ and $B_n$. It is indeed interesting to analyse $B_n$, or equivalently $A_s$, together with scaled TTT-transforms of different life distributions.

It is apparent that $B_n$ and $A_s$ are sensitive to values of $S_j$ which are near to one. Looking at the scaled TTT-transforms in Figure 1 it is rather obvious that $B_n$ and $A_s$ are more sensitive to Weibull distributions with shape parameter $\beta>1$ than to Weibull distributions with $\beta<1$. Also in this light the power estimate 0.59 for $B_n$ when $\beta=0.8$ does not seem to be proper; see Table 1.
2. Some other test statistics

We here also want to mention that the test statistics presented by Basu and Ebrahimi [2] are not the only existing statistics for testing exponentiality against HNBUE. Klefsjö [12] studied two families of test statistics intended for the same problem. In Bergman and Klefsjö [5] these families were extended to the situation when the sample is right censored. Two of the statistics discussed in these papers are

\[ Q_1 = \sum_{j=1}^{n} \left( -\frac{1}{3} + 3 \left( 1 - \frac{j}{n} \right)^3 \right) t(j)/S_n \]

and

\[ Q_2 = \sum_{j=1}^{n} \left( \frac{11}{6} - 8 \left( \frac{j}{n} \right)^3 \right) t(j)/S_n. \]

It is of interest to compare the power estimates in Table 1 and the asymptotic Pitman efficacy values in Table 2 with the corresponding values for \( Q_1 \) and \( Q_2 \). These values can be found in Tables 3 and 4, respectively. The values are reprinted from Klefsjö [12].
Table 3. Power estimates for $Q_1$ and $Q_2$ based on 2000 simulations and exact percentiles when $n=20$.

| $\hat{F}(x)=\exp(-x^{1.4})$ | 0.66 | 0.63 |
| $\hat{F}(x)=\exp(-x^{1.9})$ | 0.98 | 0.97 |
| $\hat{F}(x)=\exp(-x^{2.8})$ | 0.36 | 0.33 |
| $\hat{F}(x)=\int_0^x te^{-t}dt$ | 0.69 | 0.58 |

Table 4. Asymptotic Pitman efficacy values for $Q_1$ and $Q_2$.

<table>
<thead>
<tr>
<th></th>
<th>$Q_1$</th>
<th>$Q_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Weibull</td>
<td>1.51</td>
<td>1.34</td>
</tr>
<tr>
<td>Gamma</td>
<td>0.52</td>
<td>0.39</td>
</tr>
</tbody>
</table>

3. Testing against a distribution which is HNBUE but not NBUE

The Weibull and gamma alternatives mentioned in this paper all have monotone failure rate (IFR or DFR), a much stronger condition than HNBUE or HNWUE (see e.g., Klefsjö [9]). Therefore we have made a power study for some of the mentioned test statistics also against a piecewise exponential distribution which is HNBUE but not NBUE (and accordingly not IFR). The scaled TTT-transform of this distribution can be found in Klefsjö [10]. Here we only mention that $F$ is NBUE exactly when its scaled TTT-transform nowhere is below the diagonal (see e.g., Bergman [3]). A life distribution which is HNBUE but not NBUE accordingly has a scaled TTT-transform which is below the diagonal somewhere in $0 \leq t \leq 1$. However, calculations prove that the scaled TTT-transform is above the diagonal in the beginning and the end of this interval. The power estimates are presented in Table 5. Table 5 indicates that $B_n$ (and $A_n$) is rather effective in detecting life distributions which are HNBUE but not NBUE.

Table 5. Power estimates against a life distribution which is HNBUE but not NBUE based on 2000 simulations and exact percentiles.

<table>
<thead>
<tr>
<th></th>
<th>$B_n$ (or $A_n$)</th>
<th>$A_1$</th>
<th>$B$</th>
<th>$Q_1$</th>
<th>$Q_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$n=20$</td>
<td>0.58</td>
<td>0.08</td>
<td>0.48</td>
<td>0.65</td>
<td>0.38</td>
</tr>
<tr>
<td>$n=10$</td>
<td>0.28</td>
<td>0.07</td>
<td>0.24</td>
<td>0.21</td>
<td>0.14</td>
</tr>
</tbody>
</table>

Linköping Institute of Technology, Sweden
Luleå University, Sweden
REFERENCES


