

THE SAMPLING DISTRIBUTION OF AN ESTIMATOR ARISING IN CONNECTION WITH THE TRUNCATED EXPONENTIAL DISTRIBUTION

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1. Introduction. Let T_1, T_2, \dots, T_N be independent exponentially distributed random variables with $P\{T_j \leq t\} = 1 - e^{-\lambda t}$ for $t \geq 0$, and let $T_{(1)}, \dots, T_{(N)}$ be the corresponding order statistics. For ease of exposition we shall speak of the T_j as failure times of N parallel test items. We consider a situation where observation is truncated at time τ . $D(\tau) = \max\{n: T_{(n)} \leq \tau\}$ is the number of failures observed at or prior to time τ .

$$M(\tau) = \sum_{j=1}^{D(\tau)} T_{(j)}$$

is the total time on test until time τ for items failing at or prior to time τ , and $L(\tau) = M(\tau) + \tau\{N - D(\tau)\}$ is the total time on test until time τ . Then $\lambda^*(\tau) = D(\tau)/L(\tau)$ is a common estimator for λ . The purpose of the present note is to establish the sampling distribution for $\lambda^*(\tau)$.

2. Previous results. Sverdrup (1961) gives large sample properties for $\lambda^*(\tau)$. Bartholomew (1963) studies (small sample) properties of $1/\lambda^*(\tau)$ as an estimator for $\theta = 1/\lambda$, conditional on $D(\tau) > 0$. An interesting application of Bartholomew's result is given by Barlow et al. (1968).

A result closely related to our Theorem 1 is mentioned by Epstein [(1960), pp. 85-86]. Our result (2) was obtained by different methods by Bain and Weeks (1964).

3. Preliminaries. (i) If V_1, \dots, V_m are independent and uniformly distributed over $[0, a]$, and if $Z = \sum_{j=1}^m V_j$, then Z has a probability density

$$\phi_m(z, a) = (a^m \Gamma(m))^{-1} \sum_{\nu=0}^m (-1)^\nu \binom{m}{\nu} (z - \nu a)_+^{m-1}$$

for all z . Here $x_+^0 = 1$ for $x > 0$, $x_+^0 = 0$ for $x \leq 0$, and $x_+^n = \{\max(x, 0)\}^n$ for $n = 1, 2, \dots$.

(ii) Let T be distributed as the T_j and let $U = T$ for $T < \tau$, $U = \tau$ for $T \geq \tau$. Then $P\{U \leq u \mid T \leq t\} = b_t(1 - e^{-\lambda u})$ for $0 \leq u \leq t$, with $b_t = (1 - e^{-\lambda t})^{-1}$, and the corresponding (conditional) density is

$$f_t(u) = b_t \lambda e^{-\lambda u} \quad \text{for } 0 < u < t.$$

(iii) Let f_t^{n*} be the n th convolution of f_t with itself, and let F_t^{n*} be the corresponding distribution function. A simple induction then shows that

$$(1) \quad f_t^{n*}(u) = b_t^n (\lambda t)^n e^{-\lambda u} \phi_n(u, t).$$

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