

CHARACTERIZATIONS OF THE EXPONENTIAL DISTRIBUTION BY RECORD VALUES

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ABSTRACT. This paper presents characterizations based on the identical distribution and the finite moments of the exponential distribution by record values. We prove that $X \in EXP(\sigma)$, $\sigma > 0$, if and only if $X_{U(n+k)} - X_{U(n)}$ and $X_{U(n)} - X_{U(n-k)}$ for $n > 1$ and $k \geq 1$ are identically distributed. Also, we show that $X \in EXP(\sigma)$, $\sigma > 0$, if and only if $E(X_{U(n+k)} - X_{U(n)}) = E(X_{U(n)} - X_{U(n-k)})$ for $n > 1$ and $k \geq 1$.

1. Introduction

Let $\{X_n, n \geq 1\}$ be a sequence of independent and identically distributed(i.i.d.) random variables with cumulative distribution function(cdf) $F(x)$ and probability density function(pdf) $f(x)$. Suppose

$$Y_n = \max\{X_1, X_2, \dots, X_n\}$$

for $n \geq 1$. We say X_j is an upper record value of this sequence, if $Y_j > Y_{j-1}$ for $j > 1$. By definition, X_1 is an upper as well as a lower record value.

The indices at which the upper record values occur are given by the record times $\{U(n), n \geq 1\}$, where $U(n) = \min\{j | j > U(n-1), X_j > X_{U(n-1)}, n \geq 2\}$ with $U(1) = 1$.

We assume that all upper record values $X_{U(i)}$ for $i \geq 1$ occur at a sequence $\{X_n, n \geq 1\}$ of i.i.d. random variables.

A continuous random variable X has the exponential distribution with parameter $\sigma > 0$ if it has a cdf $F(x)$ of the form

$$(1) \quad F(x) = \begin{cases} 1 - e^{-\frac{x}{\sigma}}, & x > 0, \sigma > 0 \\ 0, & \text{otherwise.} \end{cases}$$

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A notation that designates that X has the cdf (1) is $X \in EXP(\sigma)$.

Some characterizations based on the identical distribution are known. Ahsanullah [1, 2] characterized that $X \in EXP(\sigma)$, $\sigma > 0$ if and only if $X_{U(n)}$ and $Z_1 + Z_2 + \cdots + Z_n$ for $n > 1$ are identically distributed, where Z_1, Z_2, \dots, Z_n are i.i.d. $EXP(\sigma)$. And Ahsanullah [1, 2] proved that $X \in EXP(\sigma)$, $\sigma > 0$, if and only if $X_{U(n)} - X_{U(m)}$ and $X_{U(n-m)}$ for $1 < m < n$ are identically distributed. Moreover Ahsanullah [1, 2] showed that $X \in EXP(\sigma)$, $\sigma > 0$, if and only if $X_{U(n+1)} - X_{U(n)}$ and $X_{U(n)} - X_{U(n-1)}$ for $n > 1$ are identically distributed and $X \in EXP(\sigma)$, $\sigma > 0$, if and only if $E[X_{U(n+1)} - X_{U(n)}] = E[X_{U(n)} - X_{U(n-1)}]$ for $n > 1$. We extend the above results from the record times $(U(n-1), U(n), U(n+1))$ to the record times $(U(n-k), U(n), U(n+k))$.

In this paper, we will give characterizations based on the identical distribution and the expectation of the exponential distribution by record values.

2. Main results

THEOREM 1. *Let $\{X_n, n \geq 1\}$ be a sequence of i.i.d. random variables with cdf $F(x)$ which is an absolutely continuous with pdf $f(x)$ and $F(x) < 1$ for $x > 0$. Then $F(x) = 1 - e^{-\frac{x}{\sigma}}$ for $x > 0$ and $\sigma > 0$, if and only if $X_{U(n+k)} - X_{U(n)}$ and $X_{U(n)} - X_{U(n-k)}$ for $n > 1$ and $k \geq 1$ are identically distributed and $F(x)$ belongs to C_2 .*

Proof. If $F(x) = 1 - e^{-\frac{x}{\sigma}}$ for $x > 0$ and $\sigma > 0$, then it can easily be seen that $X_{U(n+k)} - X_{U(n)}$ and $X_{U(n)} - X_{U(n-k)}$ for $n > 1$ and $k > 1$ are identically distributed.

We will prove the sufficient condition. The joint pdf $f_{n,n+k}(x, y)$ of $X_{U(n)}$ and $X_{U(n+k)}$ is

$$f_{n,n+k}(x, y) = \frac{1}{\Gamma(n)\Gamma(k)} (R(x))^{n-1} r(x) (R(y) - R(x))^{k-1} f(y)$$

for $0 < x < y$, $n > 1$ and $k \geq 1$.

Consider the functions $V = X_{U(n+k)} - X_{U(n)}$ and $W = X_{U(n)}$. It follows that $x_{U(n)} = w$, $x_{U(n+k)} = v + w$ and $|J| = 1$. Thus we can write the joint pdf $f_{V,W}(v, w)$ of V and W as

$$f_{V,W}(v, w) = \frac{1}{\Gamma(n)\Gamma(k)} (R(w))^{n-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w)$$

for $v, w > 0$, $n > 1$ and $k \geq 1$.

The marginal pdf $f_V(v)$ of V is given by

$$\begin{aligned} (2) \quad f_V(v) &= \int_0^\infty f_{V,W}(v, w) dw \\ &= \int_0^\infty \frac{1}{\Gamma(n)\Gamma(k)} (R(w))^{n-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w) dw \end{aligned}$$

for $v > 0$, $n > 1$ and $k \geq 1$.

Also, the joint pdf $f_{n-k,n}(x, y)$ of $X_{U(n-k)}$ and $X_{U(n)}$ is

$$f_{n-k,n}(x, y) = \frac{1}{\Gamma(n-k)\Gamma(k)} (R(x))^{n-k-1} r(x) (R(y) - R(x))^{k-1} f(y)$$

for $0 < x < y$, $n > 1$ and $k \geq 1$.

Let us use the transformation $V = X_{U(n)} - X_{U(n-k)}$ and $W = X_{U(n-k)}$. The Jacobian of the transformation is $|J| = 1$. Thus we can write the joint pdf $f_{V,W}(v, w)$ of V and W as

$$f_{V,W}(v, w) = \frac{1}{\Gamma(n-k)\Gamma(k)} (R(w))^{n-k-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w)$$

for $v, w > 0$, $n > 1$ and $k \geq 1$.

The marginal pdf $f_V(v)$ of V is given by

$$\begin{aligned} (3) \quad f_V(v) &= \int_0^\infty f_{V,W}(v, w) dw \\ &= \int_0^\infty \frac{1}{\Gamma(n-k)\Gamma(k)} (R(w))^{n-k-1} r(w) \\ &\quad \times (R(v+w) - R(w))^{k-1} f(v+w) dw \end{aligned}$$

for $v > 0$, $n > 1$ and $k \geq 1$.

Since $X_{U(n+k)} - X_{U(n)}$ and $X_{U(n)} - X_{U(n-k)}$ for $n > 1$ and $k \geq 1$ are identically distributed, by using (2) and (3), we get

$$\begin{aligned}
 (4) \quad & \int_0^\infty \frac{1}{\Gamma(n)\Gamma(k)} (R(w))^{n-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w) dw \\
 &= \int_0^\infty \frac{1}{\Gamma(n-k)\Gamma(k)} (R(w))^{n-k-1} r(w) \\
 &\quad \times (R(v+w) - R(w))^{k-1} f(v+w) dw
 \end{aligned}$$

for $v > 0$, $n > 1$ and $k \geq 1$.

From (4), we get on simplification

$$\begin{aligned}
 (5) \quad & \int_0^\infty (R(w))^{n-k-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w) \\
 &\quad \times \left(\frac{(R(w))^k}{\Gamma(n)} - \frac{1}{\Gamma(n-k)} \right) dw = 0
 \end{aligned}$$

for $v > 0$, $n > 1$ and $k \geq 1$.

We know that $r(x)$ and $R(x)$ are greater than zero. In addition, we say $F(x)$ belongs to the class C_2 if the hazard rate $r(x)$ is either monotone increasing or decreasing. Consequently $R(v+w) \leq R(w)$ or $R(v+w) \geq R(w)$. Therefore $R(v+w) - R(w) \leq 0$ or $R(v+w) - R(w) \geq 0$.

Thus if $F(x) \in C_2$, then (5) is true if

$$R(v+w) - R(w) = 0$$

for w and any fixed $v > 0$.

Hence

$$r(v+w) = r(w)$$

for w and any fixed $v > 0$.

By the characterization property of the exponential distribution, we have

$$F(x) = 1 - e^{-\frac{x}{\sigma}}$$

for $x > 0$ and $\sigma > 0$.

This completes the proof. □

THEOREM 2. Let $\{X_n, n \geq 1\}$ be a sequence of i.i.d. random variables with cdf $F(x)$ which is an absolutely continuous with pdf $f(x)$ and $F(x) < 1$ for $x > 0$. Then $F(x) = 1 - e^{-\frac{x}{\sigma}}$ for $x > 0$ and $\sigma > 0$, if and only if $E(X_{U(n+k)} - X_{U(n)}) = E(X_{U(n)} - X_{U(n-k)})$ for $n > 1$ and $k \geq 1$ is finite and $F(x)$ belongs to C_2 .

Proof. If $F(x) = 1 - e^{-\frac{x}{\sigma}}$ for $x > 0$ and $\sigma > 0$, then it can easily be seen that $E(X_{U(n+k)} - X_{U(n)}) = E(X_{U(n)} - X_{U(n-k)})$ for $n > 1$ and $k \geq 1$.

We will prove the sufficient condition. From (2) of Theorem 1, we can write the expectation of $V = X_{U(n+k)} - X_{U(n)}$ as

$$\begin{aligned}
 E(X_{U(n)} - X_{U(n-k)} = V) &= \int_0^\infty v f_V(v) dv \\
 &= \int_0^\infty v \left(\int_0^\infty \frac{1}{\Gamma(n)\Gamma(k)} (R(w))^{n-1} r(w) \right. \\
 (6) \quad &\quad \times (R(v+w) - R(w))^{k-1} f(v+w) dw \Big) dv \\
 &= \int_0^\infty \int_0^\infty \frac{1}{\Gamma(n)\Gamma(k)} (R(w))^{n-1} r(w) \\
 &\quad \times (R(v+w) - R(w))^{k-1} f(v+w) v dw dv
 \end{aligned}$$

for $n > 1$ and $k \geq 1$.

In the same manner as (6), we can write the expectation of $V = X_{U(n)} - X_{U(n-k)}$ as

$$\begin{aligned}
 E(X_{U(n)} - X_{U(n-k)} = V) &= \int_0^\infty v f_V(v) dv \\
 (7) \quad &= \int_0^\infty u \left(\int_0^\infty \frac{(R(w))^{n-k-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w)}{\Gamma(n-k)\Gamma(k)} dw \right) dv \\
 &= \int_0^\infty \int_0^\infty \frac{(R(w))^{n-k-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w) v}{\Gamma(n-k)\Gamma(k)} dw dv
 \end{aligned}$$

for $n > 1$ and $k \geq 1$.

Since $E(X_{U(n+k)} - X_{U(n)}) = E(X_{U(n)} - X_{U(n-k)})$ for $n > 1$ and $k \geq 1$

is finite, by using (6) and (7), we get

$$\begin{aligned}
 & \int_0^\infty \int_0^\infty \frac{1}{\Gamma(n)\Gamma(k)} (R(w))^{n-1} r(w) \\
 (8) \quad & \quad \times (R(v+w) - R(w))^{k-1} f(v+w)v \, dw dv \\
 & = \int_0^\infty \int_0^\infty \frac{(R(w))^{n-k-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w)v}{\Gamma(n-k)\Gamma(k)} \, dw dv
 \end{aligned}$$

for $n > 1$ and $k \geq 1$.

From (8), we get on simplification

$$\begin{aligned}
 (9) \quad & \int_0^\infty \int_0^\infty (R(w))^{n-k-1} r(w) (R(v+w) - R(w))^{k-1} f(v+w)v \\
 & \quad \times \left(\frac{(R(w))^k}{\Gamma(n)} - \frac{1}{\Gamma(n-k)} \right) \, dw dv = 0
 \end{aligned}$$

for $n > 1$ and $k \geq 1$.

In the same manner as Theorem 1, if $F(x) \in C_2$, then (9) is true if

$$R(v+w) - R(w) = 0$$

for w and any fixed $v > 0$.

Hence

$$r(v+w) = r(w)$$

for w and any fixed $v > 0$.

By the characterization property of the exponential distribution, we have

$$F(x) = 1 - e^{-\frac{x}{\sigma}}$$

for $x > 0$ and $\sigma > 0$.

This completes the proof. □

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