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**ON THE CHARACTERIZATION OF THE
EXPONENTIAL DISTRIBUTION BY EXPECTED VALUE
OF RECORD VALUE WITH RANDOM INDEX**

ABSTRACT. In this paper we give some characterizations of the exponential distribution based on the expected value of record value. These characterizations are considered for distribution functions belonging to IFRA or DFRA. The index of record value has the geometric distribution.

KEY WORDS: record value, expected value, IFRA, DFRA.

1. INTRODUCTION

Let X be a nonnegative random variable, and let $F(x) = P(X < x)$ be its distribution function. Let $\bar{F}(x) = 1 - F(x)$ be the survival function of X , and let $E(X)$ be the expected value of X .

We say that F has increasing failure rate average ($F \in \text{IFRA}$) if $-\frac{1}{x} \log \bar{F}(x)$ is nondecreasing in $x > 0$. Similarly, F has decreasing failure rate average ($F \in \text{DFRA}$) if $-\frac{1}{x} \log \bar{F}(x)$ is nonincreasing in $x > 0$.

It is known (see [1]) that $F \in \text{IFRA}$ if and only if

$$(1) \quad \bar{F}(\alpha x) \geq [\bar{F}(x)]^\alpha \quad \text{for all} \quad 0 < \alpha < 1 \quad \text{and} \quad x > 0,$$

and $F \in \text{DFRA}$ if and only if

$$(2) \quad \bar{F}(\alpha x) \leq [\bar{F}(x)]^\alpha \quad \text{for all} \quad 0 < \alpha < 1 \quad \text{and} \quad x > 0.$$

We say that X is exponentially distributed if

$$(3) \quad F(x) = 1 - e^{-\lambda x} \quad (x > 0) \quad \text{for some} \quad \lambda > 0.$$

We say that ν is geometrically distributed if

$$(4) \quad P(\nu = k) = p(1-p)^{k-1}, \quad k = 1, 2, \dots, \quad \text{for some} \quad 0 < p < 1.$$

Let $(X_n, n \geq 1)$ be a sequence of independent and identically distributed random variables. Define the sequence of record times $(L(n), n \geq 1)$ in the following way $L(1) = 1$, $L(n) = \min\{j: X_j > X_{L(n-1)}\}$, $n \geq 2$.

Then the sequence $(R_n, n \geq 1)$, where $R_n = X_{L(n)}$, is called the sequence of record values of $(X_n, n \geq 1)$.

The following lemmas are valid:

Lemma 1. ([3]). Let $X \geq 0$ be a random variable with distribution function F . Assume that $E(X)$ is finite. Then

$$E(X) = \int_0^{\infty} [1 - F(x)] dx.$$

Lemma 2. ([2]). Let $(X_n, n \geq 1)$ be a sequence of identically distributed random variables with finite expected value $E(X_1)$. Let N be a random variable with values in $\{1, 2, \dots\}$. Moreover, let N be distributed independently of the sequence $(X_n, n \geq 1)$. Let us suppose that the expected value $E(N)$ is finite. Then the expected value of the random variable

$$Y = \sum_{k=1}^N X_k$$

is finite and

$$E(Y) = E(X_1)E(N).$$

2. RESULTS

The following theorem is given in [5] (Theorem 4.5.2, p.129):

Let $(X_n, n \geq 1)$ be a sequence of independent and identically distributed positive random variables with a continuous distribution function F . Assume that the limit

$$(5) \quad \lim_{x \rightarrow 0^+} \frac{F(x)}{x} = \lambda, \quad 0 < \lambda < \infty.$$

Moreover, assume that ν is a geometric random variable independent of the sequence $(X_n, n \geq 1)$, and condition (4) holds. The random variables X_1 and pR_ν are identically distributed if and only if F is a distribution function of the exponential law.

Theorem 1. Let $(X_n, n \geq 1)$ be a sequence of independent and identically distributed positive random variables with a continuous distribution function F . Let condition (5) be satisfied and $F \in \text{IFRA}$. Assume that ν is a geometric random variable independent of the sequence $(X_n, n \geq 1)$, and condition (4)

holds. Let $E(R_\nu) < \infty$. Then X_1 has the distribution function defined in (3) and only if

$$(6) \quad E(X_1) = E(pR_\nu).$$

Proof. Let $F \in \text{IFRA}$. Then $E(X_1) < \infty$. It is known ([5]) that the distribution function of R_ν is of the form

$$(7) \quad F_{R_\nu}(x) = 1 - [\bar{F}(x)]^p, \quad x > 0.$$

Formula (7) gives the equality

$$(8) \quad \bar{F}_{pR_\nu}(z) = [\bar{F}(z/p)]^p \quad \text{for } z > 0.$$

By Lemma 1 we can write condition (6) as

$$(9) \quad \int_0^\infty \{ \bar{F}(z) - [\bar{F}(z/p)]^p \} dz = 0.$$

Substituting $z = py$ in (9) we get

$$(10) \quad \int_0^\infty \{ \bar{F}(py) - [\bar{F}(y)]^p \} p dy = 0.$$

Since $F \in \text{IFRA}$, inequality (1) holds. Hence

$$\bar{F}(py) \geq [\bar{F}(y)]^p \quad \text{for } y > 0.$$

From (10) we obtain

$$(11) \quad \bar{F}(py) = [\bar{F}(y)]^p \quad \text{for almost all } y > 0 \text{ and a fixed } 0 < p < 1.$$

Since $\lim_{x \rightarrow 0^+} \frac{F(x)}{x} = \lambda$, $0 < \lambda < \infty$, it follows from (11) that $\bar{F}(x) = \exp(-\lambda x)$, $x > 0$, $\lambda > 0$, (see [5], p.130).

Now suppose that X_1 has distribution function (3). Then from (7) we obtain that $F_{R_\nu}(y) = 1 - e^{-\lambda py}$ for $y > 0$, $\lambda > 0$, $0 < p < 1$, and $E(pR_\nu) = \frac{1}{\lambda} = E(X_1)$.

Remark 1. Theorem 1 is also true if the condition $F \in \text{IFRA}$ is replaced by the following one: $F \in \text{DFRA}$ and $E X_1 < \infty$. Here in the proof we use inequality (2).

We know that if there are satisfied the assumptions of Theorem 1, then the random variables

$$\sum_{i=1}^{\nu} X_i \quad \text{and} \quad R_\nu$$

are identically distributed if and only if F is a distribution function of the exponential law (see[4]).

Theorem 2. Assume that there are satisfied the assumptions of Theorem 1. Then X_1 has the distribution function defined in (3) if and only if

$$(12) \quad E\left(\sum_{i=1}^{\nu} X_i\right) = E(R_{\nu}).$$

Proof. If X_1 has exponential distribution function (3), then

$$E\left(\sum_{i=1}^{\nu} X_i\right) = E(R_{\nu}) = \frac{1}{p\lambda}.$$

Now let us suppose that condition (12) is satisfied. By Lemma 2 we have

$$E\left(\sum_{i=1}^{\nu} X_i\right) = E(X_1)E(\nu).$$

Since $E(\nu) = \frac{1}{p}$ and Lemma 1 is valid, we conclude that

$$E\left(\sum_{i=1}^{\nu} X_i\right) = \frac{1}{p} \int_0^{\infty} \bar{F}(z) dz.$$

From (8) we obtain

$$E(R_{\nu}) = \frac{1}{p} \int_0^{\infty} [\bar{F}(z/p)]^p dz.$$

Hence formula (12) takes the form

$$\frac{1}{p} \int_0^{\infty} \{\bar{F}(z) - [\bar{F}(z/p)]^p\} dz = 0.$$

Next, analogously as in the proof of Theorem 1, we get that F has form (3).

Remark 2. Theorem 2 is also true if the condition $F \in \text{IFRA}$ is replaced by the following one: $F \in \text{DFRA}$ and $E(X_1) < \infty$.

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