

**WEAK RECORDS OF GEOMETRIC DISTRIBUTION
AND SOME CHARACTERIZATIONS**

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ABSTRACT

Suppose $\{x_n, n \geq 1\}$ is a sequence of independent and identically distributed discrete random variables having the common distribution function $F(x)$. Some distributional properties of the weak records under the assumption that $F(x)$ has the geometric distribution are presented. Various properties of the weak record values and some characterizations of the geometric distribution are presented.

Key Words and Phrases Difference equation, negative binomial distribution, Equality in distribution.

1. INTRODUCTION

Let X_1, X_2, \dots be a sequence of independent and identically distributed random variables taking values on $0, 1, \dots$ with distribution function F such that $F(0) = 0$ and $a = \sup\{x | F(x) < 1\} = \infty, F(n) < 1$ for any n . The weak upper record times $U_w(n)$ are defined as follows:

$$U_w(1) = 1$$

$$U_w(n+1) = \min \{j > U_w(n), X_j \geq \max(X_1, X_2, \dots, X_{j-1})\}$$

and the corresponding weak upper record value is defined as $X_{U_w(n+1)}$.

The joint probability mass function (pmf) of $X_{U_w(1)}, X_{U_w(2)}, \dots, X_{U_w(n)}$ is given by

$$P_{w,1,2,\dots,n}(x_1, x_2, \dots, x_n) = \left(\prod_{i=1}^{n-1} \frac{p(x_i)}{P(x_i)} \right) p(x_n), \text{ for } 0 \leq x_1 \leq x_2 \leq \dots \leq x_n < \infty.$$

Here $P(X=x_i) = p(x_i)$, $\bar{P}(x_i) = P(X \geq x_i) = p(x_i) + p(x_i + 1) + \dots$

The marginal pmf's of the upper weak records are given by

$$P(X_{U_w(1)} = x_1) = P_{w,1}(x_1) = p(x_1), x_1 = 0, 1, 2, \dots, \dots$$

$$P(X_{U_w(2)} = x_2) = P_{w,2}(x_2) = R_{w,1}(x_2) p(x_2), x_2 = 0, 1, 2, \dots$$

where

$$R_{w,1}(x_2) = \sum_{0 \leq x_1 \leq x_2} \frac{p(x_1)}{\bar{P}(x_1)}$$

$$P(X_{U_w(n)} = x_n) = P_{w,n}(x_n) = R_{w,n-1}(x_n) p(x_n), (1.1)$$

where

$$R_{w,n-1}(x_n) = \sum_{0 \leq x_1 \leq x_2 \leq \dots \leq x_{n-1}} \prod_{i=1}^{n-1} \frac{p(x_i)}{\bar{P}(x_i)}$$

The joint pdf of $X_{U_w(n)}$ and $X_{U_w(n-1)}$ is

$$\begin{aligned} P(X_{U_w(n)} = x_{n-1}, X_{U_w(n-1)} = x_n) \\ = R_{w,n-1}(x_{n-1}) A_w(x_{n-1}) p(x_n), 0 \leq x_{n-1} \leq x_n \\ = 0, \text{ otherwise,} \end{aligned} \quad (1.2)$$

where

$$\begin{aligned} R_{w,n-1}(x_{n-1}) = \sum_{0 \leq x_1 \leq x_2 \leq \dots \leq x_{n-1} < x_n} \prod_{i=1}^{n-1} \frac{p(x_i)}{\bar{P}(x_i)}, \quad A_w(x_{n-1}) = \frac{p(x_{n-1})}{\bar{P}(x_{n-1})} \\ P(X_{U_w(n)} = x_n | X_{U_w(n-1)} = x_{n-1}) = \frac{p(x_n)}{\bar{P}(x_{n-1})}, \quad 0 \leq x_{n-1} \leq x_n \end{aligned} \quad (1.3)$$

We say that the distribution function $F \in \text{GE}(p)$ (the geometric distribution) if the corresponding probability mass function (pmf) is of the following form

$$\begin{aligned} P(X=k) = pq^k, k \in A_n, 0 < p=1-q < 1, A_0 = 0, 1, 2, \dots \\ = 0, \text{ otherwise.} \end{aligned} \quad (1.3)$$

We choose to distinguish between $\text{GE}(p)$ and the larger class of geometric having geometric tail. We write as $F \in \text{GT}_s(p)$ if

$$\begin{aligned} P(X=k) = cpq^x, x \in A_n \\ = 0, \text{ otherwise.} \end{aligned} \quad (1.4)$$

Here c is such that $\sum_{j=n}^{\infty} cpq^j = 1$, i.e. $c = q^{-n}$.

In this paper we will present several distributional properties of $X_{U_w(n)}$ and some characterizations of $\text{GE}(p)$ and $\text{GET}(m,p)$.

2. DISTRIBUTIONAL PROPERTIES OF $X_{U_w(n)}$

$$P(X_{U_w(1)} = x_1) = P_{w,1}(x_1) = p(x_1) = pq^{x_1}, x_1 = 0, 1, 2, \dots$$

$$P(X_{U_w(2)} = x_2) = P_{w,2}(x_2) = R_{w,1}(x_2) p(x_2), x_2 = 0, 1, 2, \dots$$

where

$$R_{w,1}(x_2) = \sum_{0 \leq x_1 \leq x_2} \frac{p(x_1)}{P(x_1)} = \sum_{0 \leq x_1 \leq x_2} p = (x_2 + 1)p$$

$$\text{Thus } P(X_{U_w(2)} = x_2) = P_{w,2}(x_2) = (x_2 + 1)p^2 q^{x_2}, x_2 = 0, 1, 2, \dots$$

$$P(X_{U_w(n)} = x_n) = P_{w,n}(x_n) = R_{w,n-1}(x_n) p(x_n),$$

where

$$R_{w,n-1}(x_n) = \sum_{0 \leq x_1 \leq x_2 \leq \dots \leq x_n} \prod_{i=1}^{n-1} \frac{p(x_i)}{P(x_i)} = \frac{(x_n + 1)(x_n + 2)(x_n + 3) \dots (x_n + n - 1)}{(n - 1)!} p^{n-1}$$

Now

$$\begin{aligned} P(X_{U_w(n)} = x_n) &= P_{w,n}(x_n) = \frac{(x_n + 1)(x_n + 2)(x_n + 3) \dots (x_n + n - 1)}{(n - 1)!} p^n q^{x_n} \\ &= \binom{x_n + n - 1}{n - 1} p^n q^{x_n}, x_n = 0, 1, 2, \dots \end{aligned} \tag{2.1}$$

It is evident that $X_{U_w(n)}$ has the negative binomial distribution. $X_{U_w(n)} \in \text{NB}(n, p)$. We say $X \in \text{NB}(m, p)$ if the pmf of X is as follows:

$$P(X = k) = \binom{k + n - 1}{n - 1} p^n q^k, k = 0, 1, 2, \dots$$

Using the relation (2.10), we can write

$$X_{U_w(n)} \stackrel{d}{=} V_1 + V_2 + \dots + V_n \tag{2.2}$$

where V_i 's are i.i.d. random variable with $P(V_i = x) = pq^x, x = 0, 1, 2, \dots$

The joint pmf of $X_{U_w(n)}$ and $X_{U_w(m)}, n > m$, is given by

$$\begin{aligned} P(X_{U_w(n)} = x_n, X_{U_w(m)} = x_m) &= \binom{x_m + m - 1}{m - 1} \binom{x_n - x_m + n - m}{n - m - 1} p^n q^{x_n}, 0 \leq x_m \leq x_n < \infty \\ &= 0, \text{ otherwise.} \end{aligned}$$

The conditional pmf of $X_{U_w(n)}$ given $X_{U_w(m)} = x_m$ is

$$\begin{aligned} P(X_{U_w(n)} = x_n | X_{U_w(m)} = x_m) &= \binom{x_n - x_m + n - m - 1}{n - m - 1} p^{n-m} q^{x_n - x_m}, 0 \leq x_m \leq x_n < \infty \\ &= 0, \text{ otherwise.} \end{aligned} \quad (2.3)$$

Let $Z_{n,m} = X_{U_w(n)} - X_{U_w(m)}$, then

$$\begin{aligned} P(Z_{n,m} = z | X_{U_w(m)} = x_m) &= \binom{z + m - n - 1}{n - m - 1} p^{n-m} q^z, 0 \leq z < \infty \\ &= 0, \text{ otherwise.} \end{aligned} \quad (2.4)$$

Thus $Z_{n,m}$ is independent of $X_{U_w(m)}$ and $P(Z_{n,m} = z)$ is given by (2.4). Further $Z_{n,m}$ and $X_{U_w(n-m)}$ are identically distributed. Using (2.3) we can write

$$X_{U_w(n)} \stackrel{d}{=} X_{U_w(n-m)} + Z_{n,m}, \quad n > m \quad (2.5)$$

Here $Z_{n,m}$ is distributed as $NB(n-m, p)$. The following moment relations follow immediately

$$\begin{aligned} E(X_{U_w(n)}) &= np^{-1}q \\ \text{Var}(X_{U_w(n)}) &= np^{-2}q \\ \text{Cov}(X_{U_w(n)}, X_{U_w(m)}) &= \text{Var}(X_{U_w(n-m)}) = (n-m)p^{-2}q \end{aligned}$$

Let $S_{(n)} = X_{U_w(n)} - \frac{np}{q}$, then the sequence $\{S_{(n)}, n=1, 2, \dots\}$ forms a martingale.

Hence

$$E(S_{(n)} | S_{(1)}, \dots, S_{(n-1)}) = S_{(n-1)}. \quad (2.6)$$

3. CHARACTERIZATIONS

Using (2.6), we obtain

$$E(X_{U_w(n+1)} | X_{U_w(n)} = x) = \sum_{x_n = x}^{\infty} x_n p q^{x_n - x} = x + \frac{q}{p} \quad (3.1)$$

and

$$E(X_{U_w(n+2)} | X_{U_w(n)} = x) = \sum_{x_n = x}^{\infty} x_n p q^{x_n - x} = x + \frac{2q}{p} \quad (3.2)$$

There are several characterizations of the $GE(p)$ distribution using the regression relations given by (2.6) and (2.7). See for details see Stepanov(1994), Aliev (1998) and Wesolowski and Ahsanullah (2001). Recently Gupta and Ahsanullah (2004) has proved the following related characterization result.

Result 1. If $\varphi(x)$ is a continuous function, then $E(\varphi(X_{U_w(n+2)} | X_{U_w(n)} = x) = s(x)$ uniquely determines the distribution

For example if $s(j) = j+b$ and $\varphi(j)=j$, then for $k= 0,1,2,\dots,P(x=k) = \frac{2}{2+b} (\frac{2}{2+b})^k$, $b>0$.

It is an open problem whether the regression property

$$E(X_{U_w(n+k)} | X_{U_w(n)} = x) = x + \frac{kq}{p}, k>2,$$

is a characterization property of the GE(p).

We have seen that $Z_{n,m}$ and $X_{U_w(n)}$ are independent. Is it a characteristic property of the geometric distribution? As a partial answer to the question, we have the following theorem for $m = n-1$.

Theorem 2.1. Let $\{X_n, n \geq 1\}$ be a sequence of independent and identically distributed random variables with common distribution function F. Suppose X is concentrated on $0,1,2,\dots$ and $a = \{x|F(x)<1\}=\infty$. Then $X_j \in \text{GET}(n,p)$ iff $Z_{n,n-1}$ and $X_{U_w(n-1)}$ are independent.

Proof. We have seen that $Z_{n,n-1}$ and $X_{U_w(n-1)}$ are independent if $X_j \in \text{GE}(p)$. We will prove here the “if” part. From (1.3), we have

$$P(Z_{n,n-1} = u | X_{U_w(n-1)} = x) = \frac{p(u+x)}{\bar{P}(x)}, \text{ } \cdot, 0 \leq x, u < \infty \tag{3.3}$$

Since $Z_{n,n-1}$ and $X_{U_w(n-1)}$ are independent, we can take

$$\begin{aligned} c(u) &= \frac{p(u+x)}{\bar{P}(x)} \\ &= \frac{\bar{P}(u+x) - \bar{P}(u+x+1)}{\bar{P}(x)} \end{aligned} \tag{3.4}$$

Summing both sides of (3.4) with respect to u from 0 to u_0-1 ($u_0>1$) and writing

$$\begin{aligned} c_1(u_0) &= \sum_{u=0}^{u_0-1} c(u) \text{ and } c_0(u_0) = 1-c_1(u_0), \text{ we get} \\ \bar{P}(x+u) &= c_0(u) \bar{P}(x), x \in A_1 \text{ and } u \in A_0. \end{aligned} \tag{3.5}$$

The general solution of (3.5) is

$$\bar{P}(x) = cp^x, \quad x \in A_1 \quad (3.6)$$

where c is independent of p . Using the boundary condition, $\bar{P}(\infty) = 0$, we obtain

$$\bar{P}(x) = cp^x, \quad x \in A_1, 0 < p < 1. \quad (3.7)$$

The Theorem is proved.

We have seen that

$$X_{U_w(n+1)} \stackrel{d}{=} V_1 + V_2 + \dots + V_{n+1} \quad (3.8)$$

where V_i 's are i.i.d. random variable with $P(V_i = x) = pq^x, x=0,1,2,\dots$

From (3.8) it is evident that

$$X_{U_w(n+1)} \stackrel{d}{=} W_n + W \quad (3.9)$$

where W_n has the same distribution as sum of n i.i.d geometric random variable .and W is independent of W_n and has the same distribution as X_i 's.

The following theorem is a characterization of GET(n,p) based on the property given in (3.9).

Theorem 2.2. Let $\{X_n, n \geq 1\}$ be a sequence of independent and identically distributed random variables with common distribution function F . Suppose X is concentrated on $0,1,2,\dots$ and $a = \{x | F(x) < 1\} = \infty$. Then $X_j \in \text{GET}(n,p)$ iff $X_{U_w(n+1)} \stackrel{d}{=} W_n + W$, where W_n has the same distribution as sum of n i.i.d geometric random variable .and W is independent of W_n and has the same distribution as X_i 's.

Proof. We will give here a proof of the “if” condition.

The condition $X_{U_w(n+1)} \stackrel{d}{=} W_n + X$ will lead to the following functional equation

$$\begin{aligned} p(x) &= \sum_{0 \leq x_0 \leq x_1 \leq \dots \leq x_{n-1} \leq x} \frac{p(x_0)}{\bar{P}(x_0)} \frac{p(x_1)}{\bar{P}(x_1)}, \dots, \frac{p(x_{n-1})}{\bar{P}(x_{n-1})} \\ &= \sum_{u=0}^x \binom{x-u+n-1}{n-1} p^n q^{x-u} p(u). \end{aligned} \quad (3.10)$$

Here $0 < p < 1, q = 1 - p$ and $\bar{P}(x) = 1 - (p(0) + p(1) + \dots + p(x-1))$.

To prove the result we will use the following two lemmas.

Lemma 1. If $a \geq 0$ and $b \geq 0$, then

$$\sum_{c=0}^{b+1} \binom{a+c}{a} = \binom{a+b+1}{a+1}$$

The proof is by induction on b . If $b=0$, this is clear, so we assume the result is true for b . Then by the basic property of binomial coefficient

$$\begin{aligned} \sum_{c=0}^{b+1} \binom{a+c}{a} &= \binom{a+b+1}{a} + \sum_{c=0}^b \binom{a+c}{a} \\ &= \binom{a+b+1}{a} + \binom{a+b+1}{a+1} = \binom{a+b+2}{a+1}. \end{aligned}$$

This shows that the result is true for $b+1$ and completes the proof.

Lemma 2. If $x \geq 0$ and $n \geq 1$, then

$$\sum_{0 \leq x_0 \leq x_1 \leq \dots \leq x_{n-1} \leq x} 1 = \binom{x+n}{n} = \sum_{u=0}^x \binom{x-u+n-1}{n-1}.$$

The second equality is an immediate consequence of Lemma 1. To establish the first equality, note that, by decomposing the $s_n(x)$ on the left side according whether $x_{n-1} = 0, 1, 2, \dots, x$, we obtain the recurrence formula

$$s_n(x) = s_{n-1}(0) + s_n(1) + \dots + s_{n-1}(x) \text{ and } s_n(0) = 1.$$

By Lemma 1, $s_n(x) = \binom{x+n}{n}$ satisfies the same recursion and $\binom{n}{n} = 1$. Hence the result.

Now we show that $p(x) = pq^x$ is a solution of (3.10). If $p(x) = pq^x$, then $\bar{P}(x) = 1 - (p(0) + p(1) + \dots + p(x-1)) = 1 - p \frac{1-q^x}{1-q} = q^x$. Inserting the values of p and \bar{P} into (3.10), we obtain the equality by Lemma 2.

Now we show that $p(x) = pq^x$ is the unique solution of (3.10), and we do this by induction over $x \geq 0$. Let $p(x)$ be any solution of (3.10) and set $f(x) = p(x) / pq^x$ for $x \geq 0$. Fix an $x > 0$ and assume $p(k) = pq^k$ for $k=0, 1, 2, \dots, x-1$, then $\bar{P}(x) = q^x$. Inserting this into (3.10) and dividing by $p^{n+1}q^x$, we obtain

$$f(x) \times \sum_{0 \leq x_0 \leq x_1 \leq \dots \leq x_{n-1} \leq x} f(x_0) f(x_1) \dots f(x_{n-1}) = \binom{x+n}{n} + f(x) - 1. \tag{3.11}$$

where we have used Lemma 1 again.

Because $f(0) = f(1) = \dots = f(x-1) = 1$, (3.11) is a polynomial equation in one unknown $f(x)$. Our goal is to show that $f(x) = 1$.

If $f(x) > 1$, then since $f(x_i) \geq 1$ and $f(x) > 1$ in (3.11), we obtain

$$\begin{aligned} f(x) \times \sum_{0 \leq x_0 \leq x_1 \leq \dots \leq x_{n-1} \leq x} 1 &> f(x) \times \sum_{0 \leq x_0 \leq x_1 \leq \dots \leq x_{n-1} \leq x} 1 \\ &= f(x) \binom{x+n}{n} = \binom{x+n}{n} + \binom{x+n}{n} (f(x) - 1) \\ &\geq \binom{x+n}{n} + (f(x) - 1). \end{aligned}$$

This contradicts (3.10). Hence we cannot have $f(x) > 1$.

If $f(x) < 1$, then since $f(x_i) < 1$ and $f(x) < 1$ in (3.11), we obtain

$$\begin{aligned} f(x) \times \sum_{0 \leq x_0 \leq x_1 \leq \dots \leq x_{n-1} \leq x} 1 \\ < f(x) \binom{x+n}{n} = \binom{x+n}{n} + \binom{x+n}{n} (f(x) - 1) \\ \leq \binom{x+n}{n} + (f(x) - 1). \end{aligned}$$

This contradicts (3.10), hence we can't have $f(x) < 1$. This establishes $f(x) = 1$. By induction $f(x) = 1$ for all $x \geq 0$.

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