

# Moments of Power Function Distribution Based on Ordered Random Variables and Characterization

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## ABSTRACT

*In this paper simple expressions for single and product moments of generalized order statistics from the power function distribution have been obtained. The results for order statistics and records are deduced from the relations derived. Further, a characterizing result of this distribution on using the conditional moments of the generalized order statistics is discussed.*

**Keywords:** Generalized order statistics, order statistics, records, power function distribution, moments and characterization.

## 1. Introduction

The concept of generalized order statistics (*gos*) was introduced by Kamps (1995). Several models of ordered random variables such as order statistics, record values, sequential order statistics, progressive type II censored order statistics and Pfeifer's record values can be discussed as special cases of the *gos*. Suppose  $X(1, n, m, k), \dots, X(n, n, m, k)$ , ( $k \geq 1$ ,  $m$  is a real number), are  $n$  *gos* from an absolutely continuous distribution function (*df*)  $F(x)$  with probability density function (*pdf*)  $f(x)$ , if their joint *pdf* is of the form

$$k \left( \prod_{j=1}^{n-1} \gamma_j \right) \left( \prod_{i=1}^{n-1} [1 - F(x_i)]^m f(x_i) \right) (1 - F(x_n))^{k-1} f(x_n) \quad (1)$$

on the cone  $F^{-1}(0) < x_1 \leq \dots \leq x_n < F^{-1}(1)$ ,

where  $\gamma_j = k + (n - j)(m + 1) > 0$  for all  $j$ ,  $1 \leq j \leq n$ ,  $k$  is a positive integer and  $m \geq -1$ .

If  $m = 0$  and  $k = 1$ , then this model reduces to the ordinary  $r$ -th order statistic and (1) will be the joint *pdf* of  $n$  order statistics  $X_{1:n} \leq X_{2:n} \leq \dots \leq X_{n:n}$  from *df*  $F(x)$ . If  $k = 1$  and  $m = -1$ , then (1) will be the joint *pdf* of the first  $n$  record values of the identically and independently distributed (*iid*) random variables with *df*  $F(x)$  and corresponding *pdf*  $f(x)$ .

In view of (1), the marginal *pdf* of the  $r$ -th *gos*,  $X(r, n, m, k)$ ,  $1 \leq r \leq n$ , is

$$f_{X(r, n, m, k)}(x) = \frac{C_{r-1}}{(r-1)!} [\bar{F}(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)), \quad (2)$$

and the joint *pdf* of  $X(r, n, m, k)$  and  $X(s, n, m, k)$ ,  $1 \leq r < s \leq n$ , is

$$f_{X(r, n, m, k) X(s, n, m, k)}(x, y) = \frac{C_{s-1}}{(r-1)!(s-r-1)!} [\bar{F}(x)]^m f(x) g_m^{r-1}(F(x)) \\ \times [h_m(F(y)) - h_m(F(x))]^{s-r-1} [\bar{F}(y)]^{\gamma_s-1} f(y), \quad x < y, \quad (3)$$

where

$$\bar{F}(x) = 1 - F(x), \quad C_{r-1} = \prod_{i=1}^r \gamma_i$$

$$h_m(x) = \begin{cases} -\frac{1}{m+1} (1-x)^{m+1}, & m \neq -1 \\ -\ln(1-x), & m = -1 \end{cases}$$

and

$$g_m(x) = h_m(x) - h_m(0), \quad x \in [0, 1).$$

Several authors utilized the concept of *gos* in their work. References may be made to Kamps and Gather (1997), Keseling (1999), Cramer and Kamps (2000), Ahsanullah (2000, 2004), Habibullah and Ahsanullah (2000), Pawlas and Szynal (2001), Raqab (2001), Kamps and Cramer (2001), Ahmad and Fawzy (2003), Bieniek and Szynal (2003), Al-Hussaini and Ahmad (2003), Saran and Pandey (2003), Cramer *et al.* (2004), Jaheen (2005), Al-Hussaini, *et al.* (2005) Khan *et al.* (2006), Khan *et al.* (2007), Ahmad (2007, 2008), Khan *et al.* (2010) among others.

In this paper we have obtained simple expressions for the exact moments of generalized order statistics from the power function distribution. Results for order statistics and record values are deduced as special cases and a characterization of this distribution is obtained by using conditional moments of generalized order statistics.

A random variable  $X$  is said to have the three parameter power function distribution if its probability density function (*pdf*) is of the following form

$$f(x) = \frac{\alpha}{\sigma} \left( \frac{\mu + \sigma - x}{\sigma} \right)^{\alpha-1}, \mu < x < \mu + \sigma, -\infty < \mu < \infty, \sigma > 0, \alpha > 0.$$

This is a Pearson's Type-I distribution. If  $\alpha = 1$  then the power function distribution coincides with the uniform distribution on the interval  $(\mu, \mu + \sigma)$ . We will consider in this paper without any loss of generality  $\mu = 0$  and  $\sigma = 1$ , i.e.

$$f(x) = \alpha(1-x)^{\alpha-1}, \quad 0 < x < 1, \alpha > 0 \tag{4}$$

and the corresponding distribution function (*df*)

$$F(x) = 1 - (1-x)^\alpha, \quad 0 < x < 1, \alpha > 0. \tag{5}$$

For applications of the distribution one may refer to Meniconi and Barry (1996), Zaka and Akhter (2013) and Arslan (2014).

## 2. Single Moments

### Lemma 2.1

For the power function distribution as given in (5) and any non-negative finite integers  $a$  and  $b$ , when  $m \neq -1$

$$J_j(a, b) = \frac{1}{(m+1)^b} \sum_{p=0}^j \sum_{u=0}^b (-1)^{u+p} \binom{j}{p} \binom{b}{u} \frac{1}{[a + u(m+1) + 1 + (p/\alpha)]}, \tag{6}$$

for  $m = -1$

$$J_j(a, b) = b! \sum_{p=0}^j (-1)^p \binom{j}{p} \frac{1}{[a + 1 + (p/\alpha)]^{b+1}}, \tag{7}$$

where

$$J_j(a, b) = \int_0^1 x^j [\bar{F}(x)]^a f(x) g_m^b(F(x)) dx. \tag{8}$$

**Proof.** On expanding binomially the term  $g_m^b(F(x)) = \left[ \frac{1}{m+1} \{1 - (\bar{F}(x))^{m+1}\} \right]^b$  in (8), we get

$$J_j(a, b) = A \int_0^1 x^j [\bar{F}(x)]^{a+u(m+1)} f(x) dx, \quad m \neq -1, \tag{9}$$

where

$$A = \frac{1}{(m+1)^b} \sum_{u=0}^b (-1)^u \binom{b}{u}.$$

Setting  $z = [\bar{F}(x)]^{1/\alpha}$  in (9), we get

$$J_j(a, b) = \alpha A \sum_{p=0}^j (-1)^p \binom{j}{p} \int_0^1 z^{\alpha[a+u(m+1)+1]+p-1} dz$$

and hence the result.

When  $m = -1$  the expression (6) is undefined, therefore we have to consider the limit as  $m$  tends to  $-1$ . Write

$$J_j(a, b) = A_1 \sum_{u=0}^b (-1)^u \binom{b}{u} \frac{[a + u(m+1) + 1 + (p/\alpha)]^{-1}}{(m+1)^b}, \tag{10}$$

where

$$A_1 = \sum_{p=0}^j (-1)^p \binom{j}{p}.$$

Differentiating numerator and denominator of (10)  $b$  times with respect to  $m$ , we get

$$J_j(a, b) = A_1 \sum_{u=0}^b (-1)^{u+b} \binom{b}{u} \frac{u^b}{[a + u(m+1) + 1 + (p/\alpha)]^{b+1}}, \quad b > 0.$$

On applying the L' Hospital rule, we have

$$\lim_{m \rightarrow -1} J_j(a, b) = A_1 \sum_{u=0}^b (-1)^{u+b} \binom{b}{u} \frac{u^b}{[a + 1 + (p/\alpha)]^{b+1}}. \tag{11}$$

But for all integers  $n \geq 0$  and for all real numbers  $x$ , from Ruiz (1996)

$$\sum_{i=0}^n (-1)^i \binom{n}{i} (x-i)^n = n!. \tag{12}$$

Therefore,

$$\sum_{u=0}^b (-1)^{u+b} \binom{b}{u} u^b = b!. \tag{13}$$

Now on substituting (13) in (11), we have the result given in (7).

**Theorem 2.1**

For the power function distribution as given in (5) and  $1 \leq r \leq n$ ,  $k = 1, 2, \dots$ ,  $m \neq -1$ ,

$$E[X^j(r, n, m, k)] = \frac{C_{r-1}}{(r-1)!} J_j(\gamma_r - 1, r-1) \tag{14}$$

$$= \frac{C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{p=0}^j \sum_{u=0}^{r-1} (-1)^{u+p} \binom{j}{p} \binom{r-1}{u} \frac{1}{[\gamma_{r-u} + (p/\alpha)]}. \tag{15}$$

**Proof.** From (2) and (8), we have

$$E[X^j(r, n, m, k)] = \frac{C_{r-1}}{(r-1)!} J_j(\gamma_r - 1, r-1).$$

Making use of Lemma 2.1, we establish the relation given in (15).

**Identity 2.1**

For  $\gamma_r \geq 1$ ,  $k \geq 1$ ,  $1 \leq r \leq n$  and  $m \neq -1$

$$\sum_{u=0}^{r-1} (-1)^u \binom{r-1}{u} \frac{1}{\gamma_{r-u}} = \frac{(r-1)!(m+1)^{r-1}}{\prod_{t=1}^r \gamma_t}. \tag{16}$$

**Proof.** (16) can be proved by setting  $j = 0$  in (15).

**Special cases**

i) Setting  $m = 0$ ,  $k = 1$  in (15), the explicit formula for the single moments of order statistics of the power function distribution can be obtained as

$$E[X_{r:n}^j] = C_{r:n} \sum_{p=0}^j \sum_{u=0}^{r-1} (-1)^{u+p} \binom{j}{p} \binom{r-1}{u} \frac{1}{[n-r+1+u+(p/\alpha)]}, \tag{17}$$

where

$$C_{r,n} = \frac{n!}{(r-1)!(n-r)!}.$$

ii) Putting  $m = -1$  in (15), we deduce the explicit expression for the single moments of upper  $k$  record values for the power function distribution in view of (7) and (14) in the form

$$E[X^j(r, n, -1, k)] = E[(Y_r^{(k)})^j] = k^r \sum_{p=0}^j (-1)^p \binom{j}{p} \frac{1}{[k + (p/\alpha)]^r}$$

and hence for upper records

$$E[(Y_r^{(1)})^j] = E[X_{U(r)}^j] = \sum_{p=0}^j (-1)^p \binom{j}{p} \frac{1}{[1 + (p/\alpha)]^r}. \tag{18}$$

### 3. Product Moments

#### Lemma 3.1

For the power function distribution as given in (2) and non-negative integers  $a$ ,  $b$  and  $c$

$$J_{i,j}(a, 0, c) = \sum_{p=0}^j \sum_{q=0}^i (-1)^{p+q} \binom{j}{p} \binom{i}{q} \times \frac{1}{[c + 1 + (p/\alpha)][a + c + 2 + (p + q)/\alpha]}, \tag{19}$$

where

$$J_{i,j}(a, b, c) = \int_0^1 \int_x^1 x^i y^j [\bar{F}(x)]^a f(x) [h_m(F(y)) - h_m(F(x))]^b \times [\bar{F}(y)]^c f(y) dy dx. \tag{20}$$

**Proof.** From (20), (5) and (4), we have

$$J_{i,j}(a, 0, c) = \alpha \int_0^1 x^i (1-x)^{\alpha(a+1)-1} G(x) dx, \tag{21}$$

where

$$G(x) = \alpha \int_x^1 y^j (1-y)^{\alpha(c+1)-1} dy. \tag{22}$$

By setting  $1 - y = z$  in (22), we get

$$G(x) = \sum_{p=0}^j (-1)^p \binom{j}{p} \frac{(1-x)^{\alpha[c+1+(p/\alpha)]}}{[c+1+(p/\alpha)]}.$$

On substituting the above expression of  $G(x)$  in (21), we find that

$$J_{i,j}(a, 0, c) = \sum_{p=0}^j (-1)^p \binom{j}{p} \frac{\alpha}{[c+1+(p/\alpha)]} \times \int_0^1 x^i (1-x)^{\alpha[a+c+2+(p/\alpha)]-1} dx. \quad (23)$$

Again by setting  $t = 1 - x$  in (23) and simplifying the resulting expression, we derive the relation given in (19).

**Lemma 3.2**

For the condition as stated in Lemma 3.1, and  $J_{i,j}(a, b, c)$  is as given in (20).

$$J_{i,j}(a, b, c) = \frac{1}{(m+1)^b} \sum_{v=0}^b (-1)^v \binom{b}{v} \times J_{i,j}(a + (b-v)(m+1), 0, c + v(m+1)). \quad (24)$$

For  $m \neq -1$

$$J_{i,j}(a, b, c) = \frac{1}{(m+1)^b} \sum_{p=0}^j \sum_{q=0}^i \sum_{v=0}^b (-1)^{p+q+v} \binom{j}{p} \binom{i}{q} \binom{b}{v} \times \frac{1}{[c+1+v(m+1)+(p/\alpha)][a+c+2+b(m+1)+(p+q)/\alpha]}, \quad (25)$$

and for  $m = -1$

$$J_{i,j}(a, b, c) = \sum_{p=0}^j \sum_{q=0}^i (-1)^{p+q} \binom{j}{p} \binom{i}{q} \times \frac{b! \alpha^{b+2}}{[c+1+(p/\alpha)]^{b+1} [a+c+2+(p+q)/\alpha]}. \quad (26)$$

**Proof.** When  $m \neq -1$ , we have

$$\begin{aligned} [h_m(F(y)) - h_m(F(x))]^b &= \frac{1}{(m+1)^b} [(\bar{F}(x))^{m+1} - (\bar{F}(y))^{m+1}]^b \\ &= \frac{1}{(m+1)^b} \sum_{v=0}^b (-1)^v \binom{b}{v} [\bar{F}(y)]^{v(m+1)} [\bar{F}(x)]^{(b-v)(m+1)}. \end{aligned}$$

Now substituting for  $[h_m(F(y)) - h_m(F(x))]^b$  in equation (20), we get

$$J_{i,j}(a,b,c) = \frac{1}{(m+1)^b} \sum_{v=0}^b (-1)^v \binom{b}{v} J_{i,j}(a + (b-v)(m+1), 0, c + v(m+1)).$$

Making use of Lemma 3.1, we derive the relation given in (25).

When  $m \neq -1$ ,  $J_{i,j}(a,b,c)$  is undefined, as  $\sum_{v=0}^b (-1)^v \binom{b}{v} = 0$ , so after applying

L'Hospital rule and (13), (25) can be proved on the lines of (7).

**Theorem 3.1**

For the power function distribution as given in (5) and  $1 \leq r < s \leq n$ ,  $k = 1, 2, \dots$  with  $m \neq -1$ ,

$$\begin{aligned} E[X^i(r,n,m,k) X^j(s,n,m,k)] &= \frac{C_{s-1}}{(r-1)!(s-r-1)!(m+1)^{r-1}} \\ &\quad \times \sum_{u=0}^{r-1} (-1)^u \binom{r-1}{u} J_{i,j}(m+u(m+1), s-r-1, \gamma_s-1), \end{aligned} \tag{27}$$

$$\begin{aligned} &= \frac{C_{s-1}}{(r-1)!(s-r-1)!(m+1)^{s-2}} \sum_{p=0}^j \sum_{q=0}^i \sum_{u=0}^{r-1} \sum_{v=0}^{s-r-1} (-1)^{p+q+u+v} \\ &\quad \times \binom{j}{p} \binom{i}{q} \binom{r-1}{u} \binom{s-r-1}{v} \frac{1}{[\gamma_{s-v} + (p/\alpha)][\gamma_{r-u} + (p+q)/\alpha]}. \end{aligned} \tag{28}$$

**Proof.** From (3), we have

$$\begin{aligned} E[X^i(r,n,m,k) X^j(s,n,m,k)] &= \frac{C_{s-1}}{(r-1)!(s-r-1)!} \int_0^1 \int_x^1 x^i y^j [\bar{F}(x)]^m f(x) \\ &\quad \times g_m^{r-1}(F(x)) [h_m(F(y)) - h_m(F(x))]^{s-r-1} [\bar{F}(y)]^{\gamma_s-1} f(y) dy dx. \end{aligned} \tag{29}$$



On expanding  $g_m^{r-1}(F(x))$  binomially in (29), we get the relation given in (27). Making use of Lemma 3.2, we derive the relation given in (28).

**Identity 3.1**

For  $\gamma_r, \gamma_s \geq 1, k \geq 1, 1 \leq r < s \leq n$  and  $m \neq -1$

$$\sum_{v=0}^{s-r-1} (-1)^v \binom{s-r-1}{v} \frac{1}{\gamma_{s-v}} = \frac{(s-r-1)!(m+1)^{s-r-1}}{\prod_{t=r+1}^s \gamma_t}. \tag{30}$$

**Proof.** At  $i = j = 0$  in (28), we have

$$1 = \frac{C_{s-1}}{(r-1)!(s-r-1)!(m+1)^{s-2}} \sum_{u=0}^{r-1} \sum_{v=0}^{s-r-1} (-1)^{u+v} \binom{r-1}{u} \binom{s-r-1}{v} \frac{1}{\gamma_{s-v} \gamma_{r-u}}.$$

Now on using (16), we get the result given in (30).

**Special cases**

i) Putting  $m = 0, k = 1$  in (28), the explicit formula for the product moments of order statistics of the power function distribution can be obtained as

$$E[X_{r:n}^i X_{s:n}^j] = C_{r,s:n} \sum_{p=0}^j \sum_{q=0}^i \sum_{u=0}^{r-1} \sum_{v=0}^{s-r-1} (-1)^{p+q+u+v} \binom{j}{p} \binom{i}{q} \binom{r-1}{u} \binom{s-r-1}{v} \\ \times \frac{1}{[n-s+v+1+(p/\alpha)][n-r+u+1+(p+q)/\alpha]},$$

where

$$C_{r,s:n} = \frac{n!}{(r-1)!(s-r-1)!(n-s)!}.$$

ii) Letting  $m$  tends to  $-1$  in (28), we deduce the explicit expression for the product moments of upper  $k$  record values for the power function distribution in view of (27) and (26) in the form

$$E[(Y_r^{(k)})^i (Y_s^{(k)})^j] = k^s \sum_{p=0}^j \sum_{q=0}^i (-1)^{p+q} \binom{j}{p} \binom{i}{q} \\ \times \frac{1}{[k+(p/\alpha)]^{s-r} [k+(p+q)/\alpha]^r}$$

and hence for upper records

$$E[X_{U(r)}^i X_{U(s)}^j] = \sum_{p=0}^j \sum_{q=0}^i (-1)^{p+q} \binom{j}{p} \binom{i}{q} \frac{1}{[1 + (p/\alpha)]^{s-r} [1 + (p+q)/\alpha]^r}.$$

**Remark 3.1**

At  $j = 0$  in (28), we have

$$E[X^i(r, n, m, k)] = \frac{C_{s-1}}{(r-1)!(s-r-1)!(m+1)^{s-2}} \sum_{q=0}^i \sum_{u=0}^{r-1} \sum_{v=0}^{s-r-1} (-1)^{q+u+v} \times \binom{i}{q} \binom{r-1}{u} \binom{s-r-1}{v} \frac{1}{\gamma_{s-v}[\gamma_{r-u} + (q/\alpha)]}. \tag{31}$$

Making use of (30) in (31) and simplifying the resulting expression, we get

$$E[X^i(r, n, m, k)] = \frac{C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{q=0}^i \sum_{u=0}^{r-1} (-1)^{q+u} \times \binom{i}{q} \binom{r-1}{u} \frac{1}{[\gamma_{r-u} + (q/\alpha)]}$$

as obtained in (15).

**4. Characterization**

Let  $X(r, n, m, k)$ ,  $r = 1, 2, \dots, n$  be gos, then the conditional pdf of  $X(s, n, m, k)$  given  $X(r, n, m, k) = x$ ,  $1 \leq r < s \leq n$ , in view of (2) and (3), is

$$f_{X(s, n, m, k) | X(r, n, m, k)}(y | x) = \frac{C_{s-1}}{(s-r-1)!C_{r-1}} [\bar{F}(x)]^{m-\gamma_r+1} \times [(h_m(F(y)) - h_m(F(x)))]^{s-r-1} [\bar{F}(y)]^{\gamma_s-1} f(y). \tag{32}$$

**Theorem 4.1**

Let  $X(r, n, m, k)$ ,  $r = 1, 2, \dots, n$  be gos based on continuous distribution function  $F(x)$  with  $F(0) = 0$  and  $0 < F(x) < 1$  for all  $x > 0$ , then for two consecutive values  $r$  and  $r + 1$ ,  $2 \leq r + 1 < s \leq n$ , the conditional expectation of gos  $X(s, n, m, k)$  given  $X(l, n, m, k) = x$ , is given as

$$E[X(s, n, m, k) | X(l, n, m, k) = x] = 1 - (1-x) \prod_{j=1}^{s-l} \left( \frac{\gamma_{l+j}}{\gamma_{l+j} + 1/\alpha} \right), \tag{33}$$

$l = r, r + 1$

if and only if  $X$  has the  $df$

$$\bar{F}(x) = (1-x)^\alpha, \quad 0 < x < 1, \quad \alpha > 0.$$

**Proof.** From (32), we have

$$E[X(s, n, m, k) | X(r, n, m, k) = x] = \frac{C_{s-1}}{(s-r-1)!C_{r-1}(m+1)^{s-r-1}} \times \int_x^1 y \left[ 1 - \left( \frac{\bar{F}(y)}{\bar{F}(x)} \right)^{m+1} \right]^{s-r-1} \left( \frac{\bar{F}(y)}{\bar{F}(x)} \right)^{\gamma_s-1} \frac{f(y)}{\bar{F}(x)} dy. \quad (34)$$

By setting  $u = \frac{\bar{F}(y)}{\bar{F}(x)} = \left( \frac{1-y}{1-x} \right)^\alpha$  from (5) in (34), we obtain

$$E[X(s, n, m, k) | X(r, n, m, k) = x] = \frac{C_{s-1}}{(s-r-1)!C_{r-1}(m+1)^{s-r-1}} \times \int_0^1 [1 - (1-x)u^{1/\alpha}] u^{\gamma_s-1} (1-u^{m+1})^{s-r-1} du = \frac{C_{s-1}}{(s-r-1)!C_{r-1}(m+1)^{s-r-1}} [I_1 - (1-x)I_2], \quad (35)$$

where

$$I_1 = \int_0^1 u^{\gamma_s-1} (1-u^{m+1})^{s-r-1} du \quad (36)$$

and

$$I_2 = \int_0^1 u^{(1/\alpha)+\gamma_s-1} (1-u^{m+1})^{s-r-1} du. \quad (37)$$

Again by setting  $t = u^{m+1}$  in (36) and (37), we get

$$I_1 = \frac{1}{m+1} \int_0^1 t^{\frac{k}{m+1}+n-s-1} (1-t)^{s-r-1} dt = \frac{(m+1)^{s-r-1} \Gamma(s-r)}{\prod_{j=1}^{s-r} \gamma_{r+j}}.$$

and

$$I_2 = \int_0^1 t^{\frac{k+1/\alpha}{\alpha(m+1)}+n-s-1} (1-t)^{s-r-1} dt$$

$$= \frac{(m+1)^{s-r-1} \Gamma(s-r)}{\prod_{j=1}^{s-r} (\gamma_{r+j} + 1/\theta)}$$

Substituting these expressions for  $I_1$  and  $I_2$  in (35) and simplifying the resulting expression, we derive the relation in (33).

To prove the sufficient part, we have from (32) and (33)

$$\begin{aligned} & \frac{C_{s-1}}{(s-r-1)! C_{r-1} (m+1)^{s-r-1}} \int_x^1 y [(\bar{F}(x))^{m+1} - (\bar{F}(y))^{m+1}]^{s-r-1} \\ & \times [\bar{F}(y)]^{\gamma_s-1} f(y) dy = [\bar{F}(x)]^{\gamma_{r+1}} H_r(x), \end{aligned} \tag{38}$$

where

$$H_r(x) = 1 - (1-x) \prod_{j=1}^{s-r} \left( \frac{\gamma_{r+j}}{\gamma_{r+j} + 1/\alpha} \right).$$

Differentiating (38) both sides with respect to  $x$ , we get

$$\begin{aligned} & - \frac{C_{s-1} [\bar{F}(x)]^m f(x)}{(s-r-2)! C_{r-1} (m+1)^{s-r-2}} \int_0^x y [(\bar{F}(x))^{m+1} - (\bar{F}(y))^{m+1}]^{s-r-2} \\ & \times [\bar{F}(y)]^{\gamma_s-1} f(y) dy = H'_r(x) [\bar{F}(x)]^{\gamma_{r+1}} - \gamma_{r+1} H_r(x) [\bar{F}(x)]^{\gamma_{r+1}-1} f(x) \end{aligned}$$

or

$$\begin{aligned} & - \gamma_{r+1} H_{r+1}(x) [\bar{F}(x)]^{\gamma_{r+2}+m} f(x) \\ & = H'_r(x) [\bar{F}(x)]^{\gamma_{r+1}} - \gamma_{r+1} H_r(x) [\bar{F}(x)]^{\gamma_{r+1}-1} f(x). \end{aligned}$$

Therefore,

$$\frac{f(x)}{\bar{F}(x)} = - \frac{H'_r(x)}{\gamma_{r+1} [H_{r+1}(x) - H_r(x)]} = \frac{\alpha}{(1-x)}$$

which proves that

$$F(x) = 1 - (1-x)^\alpha, \quad 0 < x < 1, \quad \alpha > 0.$$

### Remark 4.1

For  $k = 1$ ,  $m = 0$  and  $k = 1$ ,  $m = -1$ , we obtain the characterization results of the power function distribution based on order statistics and record values, respectively.

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