

LOWER GENERALIZED ORDER STATISTICS
FROM EXPONENTIATED GAMMA DISTRIBUTION
AND ITS CHARACTERIZATION

R. U. Khan and Devendra Kumar

Department of Statistics and Operations Research
Aligarh Muslim University, Aligarh-202 002, India
Email: *aruke@rediffmail.com*

Paper received on 01 April 2010; revised, 13 October 2010; accepted, 11 November 2010.

Abstract.

In this paper we establish explicit expressions and some recurrence relations for single and product moments of lower generalized order statistics from exponentiated gamma distribution. The results include as particular cases the above relations for moments of order statistics and lower records. Further, using a recurrence relation for single moments we obtain characterization of exponentiated gamma distribution.

Keywords. Lower generalized order statistics, order statistics, lower records, single moments, product moments, recurrence relations, exponentiated gamma distribution and characterization

AMS Subject Classification. 62G30, 62E10

1 Introduction

The concept of generalized order statistics (*gos*) was introduced by Kamps [6] as below: Let $F()$ be an absolutely continuous distribution function (*df*) with probability density function (*pdf*) $f()$. Further, let $n \in \mathbb{N}$, $n \geq 2$, $k > 0$, $\tilde{m} =$

$(m_1, m_2, \dots, m_{n-1}) \in \mathfrak{R}^{n-1}$, $M_r = \sum_{j=r}^{n-1} m_j$, such that $\gamma_r = k + n - r + M_r > 0$,

for all $r \in \{1, 2, \dots, n-1\}$. Then $X(r, n, \tilde{m}, k)$, $r = 1, 2, \dots, n$ are called *gos* if their joint *pdf* is given by

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

on the cone $F^{-1}(0+) < x_1 \leq x_2 \leq \dots \leq x_n < F^{-1}(1)$ of \mathfrak{R}^{n-1} . The model of *gos* contains as special cases, order statistics, sequential order statistics,

Stigler's order statistics and record values. But when $F()$ is an inverse distribution function, we need a concept of lower generalized order statistics (l gos), which was introduced by Pawlas and Szynal [11] as follows:

Let $n \in \mathbb{N}$, $k \geq 1$, $m \in \mathbb{R}$, be the parameters such that

$$\gamma_r = k + n - r + M_r > 0, \quad M_r = \sum_{j=r}^{n-1} m_j, \quad \forall 1 \leq r \leq n.$$

By the l gos from an absolutely continuous distribution function $F()$ with density function $f()$ we mean random variables $X'(1, n, \tilde{m}, k), \dots, X'(n, n, \tilde{m}, k)$ having joint density function of the form

$$k \left(\prod_{j=1}^{n-1} \gamma_j \right) \left(\prod_{i=1}^{n-1} [F(x_i)]^{m_i} f(x_i) \right) [F(x_n)]^{k-1} f(x_n) \quad (1.1)$$

for $F^{-1}(1) > x_1 \geq x_2 \geq \dots \geq x_n > F^{-1}(0)$.

For simplicity we shall assume $m_1 = m_2 = \dots = m_{n-1} = m$.

The *pdf* of r -th l gos is given by

$$f_{X'(r,n,m,k)}(x) = \frac{C_{r-1}}{(r-1)!} [F(x)]^{\gamma_{r-1}} f(x) g_m^{r-1}(F(x)). \quad (1.2)$$

The joint *pdf* of r -th and s -th l gos is

$$\begin{aligned} f_{X'(r,n,m,k), X'(s,n,m,k)}(x, y) &= \frac{C_{s-1}}{(r-1)!(s-r-1)!} [F(x)]^m f(x) g_m^{r-1}(F(x)) \\ &\times [h_m(F(y)) - h_m(F(x))]^{s-r-1} [F(y)]^{\gamma_s-1} f(y), \quad \alpha \leq y < x \leq \beta, \end{aligned} \quad (1.3)$$

where

$$\begin{aligned} C_{r-1} &= \prod_{i=1}^r \gamma_i \\ h_m(x) &= \begin{cases} -\frac{1}{m+1} x^{m+1}, & m \neq -1 \\ -\ln x, & m = -1 \end{cases} \end{aligned}$$

and

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

We shall also take $X'(0, n, m, k) = 0$. If $m = 0$, $k = 1$, then $X'(r, n, m, k)$ reduces to the $(n-r+1)$ -th order statistics, $X_{n-r+1:n}$ from the sample X_1, X_2, \dots, X_n and when $m = -1$, then $X'(r, n, m, k)$ reduces to the r -th k -lower record value [Pawlas and Szynal [11]]. The work of Burkschat *et al.* [3] may also refer for lower generalized order statistics.

Recurrence relations for single and product moments of lower generalized order statistics from the inverse Weibull distribution are derived by Pawlas and Szynal [11]. Khan *et al.* [8] have established recurrence relations for moments of lower generalized order statistics from exponentiated Weibull distribution. Ahsanullah [1] and Mbah and Ahsanullah [9] characterized the uniform and power function distributions based on distributional properties of lower generalized order statistics respectively. Kamps [7] investigated the importance of recurrence relations of order statistics in characterization.

In this paper, we have established explicit expressions and some recurrence relations for single and product moments of l gos from exponentiated gamma distribution. Result for order statistics and k -th lower record values are deduced as special cases and a characterization of exponentiated gamma distribution has been obtained on using a recurrence relation for single moments.

A random variable X is said to have exponentiated gamma distribution (Gupta *et al.* [4]) if its *pdf* is of the form

$$f(x) = \theta x e^{-x} [1 - e^{-x}(x+1)]^{\theta-1}, \quad x > 0, \theta > 0 \quad (1.4)$$

and the corresponding *df* is

$$F(x) = [1 - e^{-x}(x+1)]^{\theta}, \quad x > 0, \theta > 0. \quad (1.5)$$

For $\theta = 1$, the above distribution corresponds to the gamma distribution $G(2, 1)$.

For application of the distribution one may refer to Nadarajah [10] and Shawky and Bakoban [12].

2 Relations for single moments

Note that for exponentiated gamma distribution

$$F(x) = \frac{e^x(1 - e^{-x}(x+1))}{\theta x} f(x). \quad (2.1)$$

The *pdf* (1.2) can be written for the exponentiated gamma distribution with *pdf* (1.4) and *df* (1.5) in the following form

$$f_{X^{(r,n,m,k)}}(x) = \begin{cases} \frac{\theta C_{r-1}}{(r-1)!(m+1)^{r-1}} x e^{-x} (1 - e^{-x}(x+1))^{\theta \gamma_{r-1}} \\ \quad [1 - (1 - e^{-x}(x+1))^{\theta(m+1)}]^{r-1}, & m \neq -1 \\ \frac{\theta^r k^r}{(r-1)!} x e^{-x} (1 - e^{-x}(x+1))^{\theta k-1} \\ \quad [-\ln(1 - e^{-x}(x+1))]^{r-1}, & m = -1 \end{cases} \quad (2.2)$$

By using binomial and logarithmic expansions we can rewrite (2.2) as

$$f_{X'(r,n,m,k)}(x) = \begin{cases} \frac{\theta C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{a=0}^{r-1} \sum_{b=0}^{\infty} \sum_{c=0}^b (-1)^{a+b} \binom{r-1}{a} \\ \quad \binom{\theta(\gamma_r+(m+1)a-1)}{b} x^{c+1} e^{-(b+1)x}, & m \neq -1 \\ \frac{\theta^r k^r}{(r-1)!} \sum_{a=0}^{\infty} \sum_{t=0}^{\infty} \sum_{b=0}^{r-1+t+a} (-1)^a \alpha_t(r-1) \binom{\theta k-1}{a} \\ \quad \binom{r-1+t+a}{b} x^{b+1} e^{-(r+t+a)x}, & m = -1 \end{cases} \quad (2.3)$$

where $\alpha_t(r-1)$ is the coefficient of $e^{-(r-1+t)x}(x+1)^{r-1+t}$ in the expansion of $\left(\sum_{u=1}^{\infty} \frac{e^{-ux}(x+1)^u}{u}\right)^{r-1}$ [see Balakrishnan and Cohan [2], Shawky and Bakoban [13]].

We shall first establish the explicit formula for $E[X'^j(r, n, m, k)]$. Using (2.3), we obtain when $m \neq -1$

$$\begin{aligned} E[X'^j(r, n, m, k)] &= \frac{\theta C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{a=0}^{r-1} \sum_{b=0}^{\infty} \sum_{c=0}^b (-1)^{a+b} \binom{r-1}{a} \\ &\quad \times \binom{\theta(\gamma_r+(m+1)a-1)}{b} \binom{b}{c} \int_0^{\infty} x^{j+c+1} e^{-(b+1)x} dx \\ &= \frac{\theta C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{a=0}^{r-1} \sum_{b=0}^{\infty} \sum_{c=0}^b (-1)^{a+b} \binom{r-1}{a} \\ &\quad \times \binom{\theta(\gamma_r+(m+1)a-1)}{b} \binom{b}{c} \frac{\Gamma(j+c+2)}{(b+1)^{j+c+2}} \end{aligned} \quad (2.4)$$

and when $m = -1$ that

$$\begin{aligned} E[X'^j(r, n, -1, k)] &= \frac{\theta^r k^r}{(r-1)!} \sum_{a=0}^{\infty} \sum_{t=0}^{\infty} \sum_{b=0}^{r-1+t+a} (-1)^a \alpha_t(r-1) \binom{\theta k-1}{a} \\ &\quad \times \binom{r-1+t+a}{b} \int_0^{\infty} x^{j+b+1} e^{-(r+t+a)x} dx \\ &= \frac{\theta^r k^r}{(r-1)!} \sum_{a=0}^{\infty} \sum_{t=0}^{\infty} \sum_{b=0}^{r-1+t+a} (-1)^a \alpha_t(r-1) \binom{\theta k-1}{a} \\ &\quad \times \binom{r-1+t+a}{b} \frac{\Gamma(j+b+2)}{(r+t+a)^{j+b+2}}. \end{aligned} \quad (2.5)$$

If θ is a positive integer, the relations (2.4) and (2.5) then give

$$E[X'^j(r, n, m, k)] = \frac{\theta C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{a=0}^{r-1} \sum_{b=0}^{\theta(\gamma_r+(m+1)a)-1} \sum_{c=0}^b (-1)^{a+b} \binom{r-1}{a} \\ \times \binom{\theta(\gamma_r+(m+1)a)-1}{b} \binom{b}{c} \frac{\Gamma(j+c+2)}{(b+1)^{j+c+2}} \quad (2.6)$$

and

$$E[X'^j(r, n, -1, k)] = \frac{\theta^r k^r}{(r-1)!} \sum_{a=0}^{\theta k-1} \sum_{t=0}^{\infty} \sum_{b=0}^{r-1+t+a} (-1)^a \alpha_t (r-1) \binom{\theta k-1}{a} \\ \times \binom{r-1+t+a}{b} \frac{\Gamma(j+b+2)}{(r+t+a)^{j+b+2}}. \quad (2.7)$$

Special cases

1. Putting $m = 0, k = 1$ in (2.6), the explicit formula for single moments of order statistics of the exponentiated gamma distribution can be obtained as

$$E(X_{n-r+1:n}^j) = \theta C_{r:n} \sum_{a=0}^{r-1} \sum_{b=0}^{\theta(n-r+1+a)-1} \sum_{c=0}^b (-1)^{a+b} \binom{r-1}{a} \\ \times \binom{\theta(n-r+1+a)-1}{b} \binom{b}{c} \frac{\Gamma(j+c+2)}{(b+1)^{j+c+2}},$$

where

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

For $n - r + 1 = r$, the result was obtained by Shawky and Bakoban [14].

2. Putting $k = 1$ in (2.5) and (2.7), we deduce the explicit expressions for the moments of lower record values for the exponentiated gamma distribution, established by Shawky and Bakoban [13].

A recurrence relation for moments of l gos from df (1.5) can be obtained in the following theorem.

Theorem 2.1 For the distribution given in (1.5) and for $2 \leq r \leq n, n \geq 2$ and $k = 1, 2, \dots$

$$E[X'^j(r, n, m, k)] - E[X'^j(r-1, n, m, k)] \\ = \frac{j}{\theta \gamma_r} \{E[X'^{j-1}(r, n, m, k)] + E[X'^{j-2}(r, n, m, k)] - E[\phi(X'(r, n, m, k))]\}, \quad (2.8)$$

where $\phi(x) = x^{j-2}e^x$.

Proof. From (1.2), we have

$$E[X'^j(r, n, m, k)] = \frac{C_{r-1}}{(r-1)!} \int_0^\infty x^j [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx. \quad (2.9)$$

Integrating by parts taking $[F(x)]^{\gamma_r-1} f(x)$ as the part to be integrated, we get

$$\begin{aligned} E[X'^j(r, n, m, k)] &= E[X'^j(r-1, n, m, k)] \\ &\quad - \frac{j C_{r-1}}{\gamma_r (r-1)!} \int_0^\infty x^{j-1} [F(x)]^{\gamma_r} g_m^{r-1}(F(x)) dx \end{aligned}$$

the constant of integration vanishes since the integral considered in (2.9) is a definite integral. On using (2.1), we obtain

$$\begin{aligned} E[X'^j(r, n, m, k)] &= E[X'^j(r-1, n, m, k)] - \frac{j C_{r-1}}{\theta \gamma_r (r-1)!} \left\{ \int_0^\infty x^{j-2} e^x [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \right. \\ &\quad - \int_0^\infty x^{j-1} [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\ &\quad \left. - \int_0^\infty x^{j-2} [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \right\} \end{aligned}$$

and hence the result.

Remark 2.1 Putting $m = 0$, $k = 1$, in (2.8), we obtain a recurrence relation for single moments of order statistics of the exponentiated gamma distribution in the form

$$\begin{aligned} E(X_{n-r+1:n}^j) - E(X_{n-r+2:n}^j) &= \frac{j}{\theta(n-r+1)} \{ E(X_{n-r+1:n}^{j-1}) \\ &\quad + E(X_{n-r+1:n}^{j-2}) - E(\phi(X_{n-r+1:n})) \}. \end{aligned}$$

Remark 2.2 Setting $m = -1$ and $k \geq 1$, in Theorem 2.1, we get a recurrence relation for single moments of lower k -th record values from exponentiated gamma distribution in the form

$$\begin{aligned} E[X'^j(r, n, -1, k)] - E[X'^j(r-1, n, -1, k)] &= \frac{j}{\theta k} \{ E[X'^{j-1}(r, n, -1, k)] + E[X'^{j-2}(r, n, -1, k)] \\ &\quad - E[\phi(X'(r, n, -1, k))] \}. \end{aligned}$$

3 Relations for product moments

On using (1.3), (1.4), (1.5) and binomial, logarithmic expansions the explicit expressions for the product moments of l gos $X'^i(r, n, m, k)$ and $X'^j(s, n, m, k)$,

$1 \leq r < s \leq n$, can be obtained when $m \neq -1$ as

$$\begin{aligned}
& E[X^{r,i}(r, n, m, k) X^{s,j}(s, n, m, k)] \\
&= \frac{C_{s-1}}{(r-1)!(s-r-1)!(m+1)^{s-2}} \sum_{a=0}^{r-1} \sum_{b=0}^{s-r-1} (-1)^{a+b} \binom{r-1}{a} \binom{s-r-1}{b} \\
&\quad \times \int_0^\infty y^j [F(y)]^{\gamma_{s-b}-1} f(y) I(y) dy, \tag{3.1}
\end{aligned}$$

where

$$\begin{aligned}
I(y) &= \int_y^\infty x^i [F(x)]^{(s-r+a-b)(m+1)-1} f(x) dx \\
&= \theta \sum_{c=0}^\infty \sum_{d=0}^c (-1)^c \binom{\theta(s-r+a-b)(m+1)-1}{c} \binom{c}{d} \int_y^\infty x^{i+d+1} e^{-(1+c)x} dx \\
&= \theta \sum_{c=0}^\infty \sum_{d=0}^c (-1)^c \binom{\theta(s-r+a-b)(m+1)-1}{c} \binom{c}{d} \\
&\quad \times \sum_{t=0}^{i+d+1} \frac{e^{-(1+c)y} ((1+c)y)^t \Gamma(i+d+2)}{t! (1+c)^{i+d+2}}.
\end{aligned}$$

On substituting the above expression of $I(y)$ in (3.1), we find that

$$\begin{aligned}
& E[X^{r,i}(r, n, m, k) X^{s,j}(s, n, m, k)] \\
&= \frac{\theta^2 C_{s-1}}{(r-1)!(s-r-1)!(m+1)^{s-2}} \sum_{a=0}^{r-1} \sum_{b=0}^{s-r-1} \sum_{c=0}^\infty \sum_{d=0}^c \sum_{t=0}^{i+d+1} \sum_{u=0}^\infty \sum_{v=0}^u (-1)^{a+b+c+u} \\
&\quad \times \binom{r-1}{a} \binom{s-r-1}{b} \binom{\theta(s-r+a-b)(m+1)-1}{c} \binom{c}{d} \\
&\quad \times \binom{\theta\gamma_{s-b}-1}{u} \binom{u}{v} \frac{\Gamma(i+d+2)}{t!(1+c)^{i+d+2-t}} \frac{\Gamma(j+t+v+2)}{(c+2+u)^{j+t+v+2}} \tag{3.2}
\end{aligned}$$

and when $m = -1$ that

$$\begin{aligned}
& E[X^{r,i}(r, n-1, k) X^{s,j}(s, n, -1, k)] \\
&= \frac{k^s}{(r-1)!(s-r-1)!} \int_0^\infty y^j [F(y)]^{k-1} f(y) I(y) dy, \tag{3.3}
\end{aligned}$$

where

$$\begin{aligned}
I(y) &= \int_y^\infty x^i [-\ln F(x)]^{r-1} [-\ln F(y) + \ln F(x)]^{s-r-1} \frac{f(x)}{F(x)} dx \\
&= \sum_{a=0}^{s-r-1} \sum_{b=0}^\infty \sum_{t=0}^\infty \sum_{c=0}^{s-2-a+t+b} \sum_{d=0}^{i+c+1} (-1)^{s-r-1} \theta^{s-1-a} \alpha_t(s-2-a) \\
&\quad \times \binom{s-r-1}{a} \binom{s-2-a+t+b}{c} \\
&\quad \times \frac{\Gamma(i+2+c)}{d!(s-1-a+t+b)^{i+2+c-d}} [\ln F(y)]^a e^{-(s-1-a+t+b)y} y^d.
\end{aligned}$$

On substituting the above expression of $I(y)$ in (3.3) and simplifying the resulting equation, we obtain

$$\begin{aligned}
&E[X^{ri}(r, n, -1, k) X^{rj}(s, n, -1, k)] \\
&= \frac{\theta^s k^s}{(r-1)!(s-r-1)!} \sum_{a=0}^{s-r-1} \sum_{b=0}^\infty \sum_{t=0}^\infty \sum_{c=0}^{s-2-a+t+b} \sum_{d=0}^{i+c+1} \sum_{v=0}^\infty \sum_{u=0}^\infty \sum_{w=0}^{a+u+v} (-1)^{s-r-1+a+v} \\
&\quad \times \alpha_t(s-2-a) \alpha_u(a) \binom{s-r-1}{a} \binom{s-2-a+t+b}{c} \binom{\theta k - 1}{v} \binom{a+u+v}{w} \\
&\quad \times \frac{\Gamma(i+2+c) \Gamma(j+d+w+2)}{d!(s-1-a+t+b)^{i+2+c-d} (s+b+t+v+u)^{j+d+w+2}}. \tag{3.4}
\end{aligned}$$

If θ is a positive integer, then the relations (3.2) and (3.4) take the form

$$\begin{aligned}
&E[X^{ri}(r, n, m, k) X^{rj}(s, n, m, k)] \\
&= \frac{\theta^2 C_{s-1}}{(r-1)!(s-r-1)!(m+1)^{s-2}} \sum_{a=0}^{r-1} \sum_{b=0}^{s-r-1} \sum_{c=0}^{\theta(s-r+a-b)(m+1)-1} \sum_{d=0}^c \sum_{t=0}^{i+d+1} \sum_{u=0}^{\theta\gamma_{s-b}-1} \\
&\quad \times \sum_{v=0}^u (-1)^{a+b+c+u} \binom{r-1}{a} \binom{s-r-1}{b} \binom{\theta(s-r+a-b)(m+1)-1}{c} \\
&\quad \times \binom{c}{d} \binom{\theta\gamma_{s-b}-1}{u} \binom{u}{v} \\
&\quad \times \frac{\Gamma(i+d+2) \Gamma(j+t+v+2)}{t!(1+c)^{i+d+2-t} (c+2+u)^{j+t+v+2}}. \tag{3.5}
\end{aligned}$$

$$\begin{aligned}
 & E[X^{ri}(r, n, -1, k)X^{rj}(s, n - 1, k)] \\
 &= \frac{\theta^s k^s}{(r - 1)!(s - r - 1)!} \sum_{a=0}^{s-r-1} \sum_{b=0}^{\infty} \sum_{t=0}^{\infty} \sum_{c=0}^{s-2-a+t+b} \sum_{d=0}^{i+c+1} \sum_{v=0}^{\theta k-1} \sum_{u=0}^{\infty} \sum_{w=0}^{a+u+v} \\
 &\times (-1)^{s-r-1+a+v} \alpha_t(s - 2 - a) \alpha_u(a) \binom{s - r - 1}{a} \binom{s - 2 - a + t + b}{c} \\
 &\times \binom{\theta k - 1}{v} \binom{a + u + v}{w} \\
 &\times \frac{\Gamma(i + 2 + c)\Gamma(j + d + w + 2)}{d!(s - 1 - a + t + b)^{i+2+c-d}(s + b + t + v + u)^{j+d+w+2}}. \tag{3.6}
 \end{aligned}$$

Special cases

1. Putting $m = 0, k = 1$ in (3.5) the explicit formula for the product moments of order statistics of the exponentiated gamma distribution to be obtained as

$$\begin{aligned}
 E(X_{n-r+1:n}^i X_{n-s+1:n}^j) &= \theta^2 C_{r,s;n} \sum_{a=0}^{r-1} \sum_{b=0}^{s-r-1} \sum_{c=0}^{\theta(s-r+a-b)-1} \sum_{d=0}^c \sum_{t=0}^{i+d+1} \\
 &\times \sum_{u=0}^{\theta(n-s+1+b)-1} \sum_{v=0}^u (-1)^{a+b+c+u} \binom{r-1}{a} \\
 &\times \binom{s-r-1}{b} \binom{\theta(s-r+a-b)-1}{c} \binom{c}{d} \binom{\theta(n-s+1+b)-1}{u} \binom{u}{v} \\
 &\times \frac{\Gamma(i+d+2)\Gamma(j+t+v+2)}{t!(1+c)^{i+d+2-t}(c+2+u)^{j+t+v+2}},
 \end{aligned}$$

where

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

For $n - s + 1 = r, n - r + 1 = s$, the result was obtained by Shawky and Bakoban [14].

2. Putting $k = 1$ in (3.4) and (3.6), we deduce the explicit expressions for the product moments of lower record values for the exponentiated gamma distribution, established by Shawky and Bakoban [13].

Making use of (2.1), we can derive recurrence relations for product moments of l gos from (1.5).

Theorem 3.1 For the distribution given in (1.5) and for $1 \leq r < s \leq n$, $n \geq 2$ and $k = 1, 2, \dots$

$$\begin{aligned} & E[X'^i(r, n, m, k)X'^j(s, n, m, k)] - E[X'^i(r, n, m, k)X'^j(s-1, n, m, k)] \\ &= \frac{j}{\theta\gamma_r} \{E[X'^i(r, n, m, k)X'^{j-1}(s, n, m, k)] + E[X'^i(r, n, m, k)X'^{j-2}(s, n, m, k)] \\ & \quad - E[\phi(X'(r, n, m, k)X'(s, n, m, k))]\}, \end{aligned} \quad (3.7)$$

where $\phi(x, y) = x^i y^{j-2} e^y$.

Proof. From (1.3), we have

$$\begin{aligned} & E[X'^i(r, n, m, k)X'^j(s, n, m, k)] \\ &= \frac{C_{s-1}}{(r-1)!(s-r-1)!} \int_0^\infty x^i [F(x)]^m f(x) g_m^{r-1}(F(x)) I(x) dx, \end{aligned} \quad (3.8)$$

where

$$I(x) = \int_0^x y^j [h_m(F(y)) - h_m(F(x))]^{s-r-1} [F(y)]^{\gamma_s-1} f(y) dy.$$

Solving the integral in $I(x)$ by parts and substituting the resulting expression in (3.8), we get

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

the constant of integration vanishes since the integral in $I(x)$ is a definite integral. On using the relation (2.1), we obtain

$$\begin{aligned} & E[X'^i(r, n, m, k)X'^j(s, n, m, k)] \\ &= E[X'^i(r, n, m, k)X'^j(s, n, m, k)] - \frac{jC_{s-1}}{\theta\gamma_s(r-1)!(s-r-1)!} \left\{ \int_0^\infty \int_0^x x^i y^{j-2} e^y \right. \\ & \quad \times [F(x)]^m f(x) g_m^{r-1}(F(x)) [h_m(F(y)) - h_m(F(x))]^{s-r-1} [F(y)]^{\gamma_s-1} f(y) dy dx \\ & \quad - \int_0^\infty \int_0^x x^i y^{j-1} [F(x)]^m f(x) g_m^{r-1}(F(x)) [h_m(F(y)) - h_m(F(x))]^{s-r-1} \\ & \quad \times [F(y)]^{\gamma_s-1} f(y) dy dx - \int_0^\infty \int_0^x x^i y^{j-2} [F(x)]^m f(x) g_m^{r-1}(F(x)) \\ & \quad \left. \times [h_m(F(y)) - h_m(F(x))]^{s-r-1} [F(y)]^{\gamma_s-1} f(y) dy dx \right\} \end{aligned}$$

and hence the result.

Remark 3.1 Putting $m = 0$, $k = 1$, in (3.7), we obtain recurrence relations for product moments of order statistics of the exponentiated gamma distribution in the form

$$\begin{aligned} & E(X_{n-r+1:n}^i X_{n-s+1:n}^j) - E(X_{n-r+1:n}^i X_{n-s+2:n}^j) \\ &= \frac{j}{\theta(n-r+1)} \left\{ E(X_{n-r+1:n}^i X_{n-s+1:n}^{j-1}) + E(X_{n-r+1:n}^i X_{n-s+1:n}^{j-2}) \right. \\ & \quad \left. - E(\phi(X_{n-r+1:n} X_{n-s+1:n})) \right\}. \end{aligned}$$

Remark 3.2 Setting $m = -1$ and $k \geq 1$, in Theorem 3.1, we obtain the recurrence relations for product moments of lower k -th record values from exponentiated gamma distribution in the form

$$\begin{aligned} & E[X^{ri}(r, n, -1, k) X'^j(s, n, -1, k)] - E[X^{ri}(r, n, -1, k) X'^j(s-1, n, -1, k)] \\ &= \frac{j}{\theta k} \{ E[X^{ri}(r, n, -1, k) X'^{j-1}(s, n, -1, k)] \\ & \quad + E[X^{ri}(r, n, -1, k) X'^{j-2}(s, n, -1, k)] \\ & \quad - E[\phi(X'(r, n, -1, k) X'(s, n, -1, k))] \}. \end{aligned}$$

4 Characterization

Theorem 4.1 Let X be a non-negative random variable having an absolutely continuous distribution function $F(x)$ with $F(0) = 0$ and $0 < F(x) < 1$ for all $x > 0$, then

$$\begin{aligned} E[X'^j(r, n, m, k)] &= E[X'^j(r-1, n, m, k)] - \frac{j}{\theta \gamma_r} E[\phi(X'(r, n, m, k))] \\ & \quad + \frac{j}{\theta \gamma_r} E[(X'^{j-1}(r, n, m, k))] + \frac{j}{\theta \gamma_r} E[(X'^{j-2}(r, n, m, k))] \quad (4.1) \end{aligned}$$

if and only if

$$F(x) = [1 - e^{-x}(1+x)]^\theta.$$

Proof. The necessary part follows immediately from equation (2.8). On the other hand if the recurrence relation in equation (4.1) is satisfied, then on using

equation (1.2), we have

$$\begin{aligned}
& \frac{C_{r-1}}{(r-1)!} \int_0^\infty x^j [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\
&= \frac{(r-1)C_{r-1}}{\gamma_r(r-1)!} \int_0^\infty x^j [F(x)]^{\gamma_r+m} f(x) g_m^{r-2}(F(x)) dx \\
&- \frac{jC_{r-1}}{\theta\gamma_r(r-1)!} \int_0^\infty x^{j-2} e^x [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\
&+ \frac{jC_{r-1}}{\theta\gamma_r(r-1)!} \int_0^\infty x^{j-1} [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\
&+ \frac{jC_{r-1}}{\theta\gamma_r(r-1)!} \int_0^\infty x^{j-2} [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx. \quad (4.2)
\end{aligned}$$

Integrating the first integral on the right hand side of equation (4.2), by parts, we get

$$\begin{aligned}
& \frac{C_{r-1}}{(r-1)!} \int_0^\infty x^j [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\
&= \frac{jC_{r-1}}{\gamma_r(r-1)!} \int_0^\infty x^{j-1} [F(x)]^{\gamma_r} g_m^{r-1}(F(x)) dx \\
&+ \frac{C_{r-1}}{(r-1)!} \int_0^\infty x^j [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\
&- \frac{jC_{r-1}}{\theta\gamma_r(r-1)!} \int_0^\infty x^{j-2} e^x [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\
&+ \frac{jC_{r-1}}{\theta\gamma_r(r-1)!} \int_0^\infty x^{j-1} [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx \\
&+ \frac{jC_{r-1}}{\theta\gamma_r(r-1)!} \int_0^\infty x^{j-2} [F(x)]^{\gamma_r-1} f(x) g_m^{r-1}(F(x)) dx
\end{aligned}$$

which reduces to

$$\begin{aligned}
& \frac{jC_{r-1}}{\gamma_r(r-1)!} \int_0^\infty x^{j-1} [F(x)]^{\gamma_r-1} g_m^{r-1}(F(x)) dx \\
& \left\{ F(x) - \frac{1}{\theta x e^{-x}} f(x) + \frac{1}{\theta} f(x) + \frac{1}{\theta x} f(x) \right\} dx = 0. \quad (4.3)
\end{aligned}$$

Now applying a generalization of the Müntz-Szász Theorem (Hwang and Lin, [5]) to equation (4.3), we get

$$\frac{f(x)}{F(x)} = \frac{\theta x e^{-x}}{[1 - e^{-x}(1+x)]}$$

which prove that

$$F(x) = [1 - e^{-x}(1+x)]^\theta, \quad x \geq 0, \theta > 0.$$

Acknowledgement

The authors would like to thank the referee for carefully reading the paper and for helpful suggestions which greatly improved the paper.

References

- [1] Ahsanullah, M. (2004): A characterization of the uniform distribution by dual generalized order statistics. *Comm. Statist. Theory Methods*, 33, 2921–2928.
- [2] Balakrishnan, N. and Cohan, A. C. (1991): *Order Statistics and Inference: Estimation Methods*. Academic Press, San Diego.
- [3] Burkschat, M., Cramer, E. and Kamps, U. (2003): Dual generalized order statistics. *Metron*, LXI(1), 13–26.
- [4] Gupta, R. C., Gupta, R. D. and Gupta, P. L. (1998): Modelling failure time data by Lehman alternatives. *Comm. Statist. Theory Methods*, 27, 887–904.
- [5] Hwang, J. S. and Lin, G. D. (1984): On a generalized moments problem II. *Proc. Amer. Math. Soc.*, 91, 577–580.
- [6] Kamps, U. (1995): *A Concept of Generalized Order Statistics*. B. G. Teubner Stuttgart.
- [7] Kamps, U. (1998): Characterizations of distributions by recurrence relations and identities for moments of order statistics. In: Balakrishnan, N. and Rao, C. R., *Handbook of Statistics 16, Order Statistics: Theory & Methods*, North-Holland, Amsterdam, 291–311.
- [8] Khan, R. U., Anwar, Z. and Athar, H. (2008): Recurrence relations for single and product moments of dual generalized order statistics from exponentiated Weibull distribution. *Aligarh J. Statist.*, 28, 37–45.
- [9] Mbah, A. K. and Ahsanullah, M. (2007): Some characterization of the power function distribution based on lower generalized order statistics. *Pakistan J. Statist.*, 23, 139–146.
- [10] Nadarajah, S. (2007): The exponentiated gamma distribution with application to drought data. *Calcutta Statist. Assoc. Bull.*, 59 (233–234), 29–54.
- [11] Pawlas, P. and Szynal, D. (2001): Recurrence relations for single and product moments of lower generalized order statistics from the inverse Weibull distribution. *Demonstratio Math.*, XXXIV(2), 353–358.

- [12] Shawky, A. I., Bakoban R. A. (2008): Bayesian and Non-Bayesian estimations on the exponentiated gamma distribution. *Appl. Math. Sci. (Ruse)*, 2(51), 2521–2530.
- [13] Shawky, A. I., Bakoban R. A. (2008): Characterization from exponentiated gamma distribution based on record values. *J. Stat. Theory Appl.*, 7, 263–277.
- [14] Shawky, A. I., Bakoban R. A. (2009): Order statistics from exponentiated gamma distribution and associated inference. *Int. J. Contemp. Math. Sci.*, 4, 71–91.

ProbStat Forum is an e-journal. For details please visit; www.probstat.org.in.