Moment Properties of Generalized Order Statistics from Exponential-Weibull Lifetime Distribution

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Abstract. In this paper, simple explicit expressions and some recurrence relations satisfied by single and product moments of generalized order statistics from exponential-Weibull lifetime distribution have been obtained. These relations are deduced for moments of order statistics and upper record values. Further, conditional moments and a recurrence relation for single moments of the generalized order statistics are used to characterize this distribution and some computational works are also carried out.

Keywords: Exponential-Weibull lifetime distribution, generalized order statistics, record values, order statistics, single moments, product moments, conditional moments, recurrence relations, characterization.

1 Introduction

The concept of generalized order statistics (gos) was introduced by Kamps [1]. A variety of order models of random variables is contained in this concept, such as order statistics, upper record values, progressive Type II censoring order statistics, sequential order statistics and Pfeifer's records.

Let X_1, X_2, \ldots be a sequence of independent and identically distributed *(iid)* random variables *(rv)* with distribution function *(df)* F(x) and probability density function *(pdf)* f(x). Let k > 0, $n \in N$, $m \in \Re$ and $\gamma_r = k + (n - r)(m + 1) > 0$. If the random variables X(r, n, m, k), $r = 1, 2, \ldots, n$ possess a joint *pdf* of the form

$$k\left(\prod_{j=1}^{n-1}\gamma_{j}\right)\left(\prod_{i=1}^{n-1}\left[1-F(x_{i})\right]^{m}f(x_{i})\right)\left[1-F(x_{n})\right]^{k-1}f(x_{n})$$
(1)

on the cone $F^{-1}(0) < x_1 \leq ... \leq x_n < F^{-1}(1)$, then they are called *gos* of a sample from a distribution with df = F(x). Note that in the case m = 0, k = 1, this model reduces to the joint pdf of the ordinary order statistics and when m = -1, we get the joint pdf of the k – th upper record values. In view of (1), the marginal pdf of the r – th gos X(r, n, m, k) is given by

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

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

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

where

$$\begin{split} \overline{F}(x) &= 1 - F(x) , \qquad C_{r-1} = \prod_{i=1}^{r} \gamma_i , \qquad r = 1, 2, \dots, n \\ h_m(x) &= \begin{cases} -\frac{1}{m+1} (1-x)^{m+1} , & m \neq -1 \\ -\ln(1-x) , & m = -1 \end{cases} \end{split}$$

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

Several authors utilized the concept of *gos* in their studies. References may be made to Kamps and Gather [2], Keseling [3], Cramer and Kamps [4], Ahsanullah [5], Habibullah and Ahsanullah [6], Raqab [7], Kamps and Cramer [8], Ahmad and Fawzy [9], Beiniek and Syznal [10], Al-Hussaini and Ahmad [11], Cramer *et al.* [12], Khan and Alzaid [13], Jaheen [14], Khan *et al.* ([15],[16]), Khan and Zia [17] among others.

Kamps [18] investigated the importance of recurrence relations of order statistics in characterization. In this paper, we study the generalized order statistics from exponential-Weibull lifetime distribution and derive explicit expressions for single moments. We also establish some simple recurrence relationships for the single and the product moments. Further, various deductions and particular cases are discussed. At the end, the characterization results based on conditional expectation and recurrence relations are presented and some computational works are also carried out.

A random variable X is said to have exponential-Weibull lifetime distribution (Cordeiro et al. [19]) if its pdf is of the form

$$f(x) = (\alpha + \beta \theta x^{\theta-1})e^{-(\alpha x + \beta x^{\theta})}, x > 0, \alpha > 0, \beta > 0, \theta > 0$$

$$\tag{4}$$

and the corresponding df is

$$F(x) = 1 - e^{-(\alpha x + \beta x^{\theta})}, x > 0, \alpha > 0, \beta > 0, \theta > 0$$
(5)

It can be seen that

$$f(x) = (\alpha + \beta \theta x^{\theta - 1})\overline{F}(x)$$
(6)

We can obtain several special models from relation (4). The exponential and Weibull distributions are the special cases for $\theta = 1$ or $\theta = 1$, $\alpha = 0$ or $\beta = 0$ and $\alpha = 0$, respectively. The Rayleigh distribution arises when $\alpha = 0$ and $\theta = 2$. The two-parameter linear failure rate distribution is obtained when $\theta = 2$.

The relation (6) will be used to derive explicit expressions and some recurrence relations for the moments of gos from exponential-Weibull lifetime distribution.

2 Relations for Single Moments

We shall first establish the existence of $E[X^{j}(r, n, m, k)]$. Using (2), we have when $m \neq -1$

$$E[X^{j}(r,n,m,k)] = \frac{C_{r-1}}{(r-1)!} \int_{0}^{\infty} x^{j} [\overline{F}(x)]^{\gamma_{r}-1} f(x) g_{m}^{r-1}(F(x)) \,\mathrm{d}\,x$$
(7)

By using binomial expansion, (7) can be rewritten as

$$E[X^{j}(r,n,m,k)] = \frac{C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{u=0}^{r-1} (-1)^{u} {\binom{r-1}{u}} \int_{0}^{\infty} x^{j} [\overline{F}(x)]^{\gamma_{r-u}-1} f(x) \, \mathrm{d} x \tag{8}$$

Further, on using (6) in (8), we obtain

$$E[X^{j}(r,n,m,k)] = A \int_{0}^{\infty} x^{j} [\overline{F}(x)]^{\gamma_{r-u}} (\alpha + \beta \theta x^{\theta-1}) dx$$
$$= -\frac{A}{\gamma_{r-u}} \int_{0}^{\infty} x^{j} \left(\frac{d \left(e^{-\gamma_{r-u}(\alpha x + \beta x^{\theta})} \right)}{dx} \right) dx$$

where $A = \frac{C_{r-1}}{(r-1)!(m+1)^{r-1}} \sum_{u=0}^{r-1} (-1)^u \binom{r-1}{u}.$ Integrating by parts now yields

$$E[X^{j}(r,n,m,k)] = \frac{jA}{\gamma_{r-u}} \int_{0}^{\infty} x^{j-1} e^{-\gamma_{r-u}(\alpha x + \beta x^{\theta})} dx$$

On expanding $e^{-\gamma_{r-u} \alpha x}$ in Taylor series, we get

$$E[X^{j}(r, n, m, k)] = A^{*} \int_{0}^{\infty} x^{j+\nu-1} e^{-\gamma_{r-u}\beta x^{\theta}} dx$$
(9)

Journal of Advanced Statistics, Vol. 1, No. 3, September 2016 https://dx.doi.org/10.22606/jas.2016.13004

where $A^* = \frac{jA}{\gamma_{r-u}} \sum_{v=0}^{\infty} (-1)^v \frac{(\alpha \gamma_{r-u})^v}{v!}$. We have [20]

$$\int_{0}^{\infty} x^{m} e^{-\beta x^{n}} dx = \frac{\Gamma(m+1) / n}{n\beta^{(m+1)/n}}, \beta, m, n > 0$$
(10)

Now on substituting (10) in (9), we have

$$E[X^{j}(r,n,m,k)] = \frac{jC_{r-1}}{(r-1)!\,\theta(m+1)^{r-1}} \sum_{v=0}^{\infty} \sum_{u=0}^{r-1} (-1)^{u+v} \binom{r-1}{u} \times \frac{\alpha^{v}(\gamma_{r-u})^{v-1-(j+v)/\theta}}{v!\,\beta^{(j+v)/\theta}} \,\Gamma\{(j+v)\,/\,\theta\}$$
(11)

When m = -1,

$$E[X^{j}(r,n,m,k)] = \frac{0}{0} \text{ as } \sum_{u=0}^{r-1} (-1)^{u} \binom{r-1}{u} = 0$$

Since (11) is of the form $\frac{0}{0}$ at m = -1, therefore, we have

$$E[X^{j}(r,n,m,k)] = B\sum_{u=0}^{r-1} (-1)^{u} {\binom{r-1}{u}} \frac{[k+(n-r+u)(m+1)]^{v-\{(j+v)/\theta\}-1}}{(m+1)^{r-1}}$$
(12)
$$\sum_{u=0}^{\infty} (-1)^{v} \alpha^{v} \Gamma\{(j+v)/\theta\}$$

where $B = \frac{jC_{r-1}}{(r-1)!\theta} \sum_{v=0}^{\infty} (-1)^v \frac{\alpha^v \Gamma\{(j+v) / \theta\}}{v! \beta^{(j+v)/\theta}}.$

Differentiating numerator and denominator of (12) (r-1) times with respect to m, we get

$$\begin{split} E[X^{j}(r,n,m,k)] &= B \sum_{u=0}^{r-1} (-1)^{u+(r-1)} \binom{r-1}{u} \\ &\times \frac{[\{(j+v) \ / \ \theta\} + 1 - v][\{(j+v) \ / \ \theta\} + 2 - v] \dots [\{(j+v) \ / \ \theta\} + r - 1 - v](n-r+u)^{r-1}}{(r-1)![k+(n-r+u)(m+1)]^{\{(j+v) \ / \ \theta\} + r-v}} \end{split}$$

On applying L' Hospital rule, we have

$$\lim_{m \to -1} E[X^{j}(r, n, m, k)] = B \frac{[\{(j+v) / \theta\} + 1 - v] \dots [\{(j+v) / \theta\} + r - 1 - v]}{(r-1)! k^{\{(j+v)/\theta\} + r - v}}$$

$$\times \sum_{u=0}^{r-1} (-1)^{u} \binom{r-1}{u} (r-n-u)^{r-1}$$
(13)

But for all integers $n \ge 0$ and for all real numbers x, we have [21]

$$\sum_{i=0}^{n} (-1)^{i} \binom{n}{i} (x-i)^{n} = n!$$
(14)

Substitute (14) in (13) and after simplification, we find that

$$E[X^{j}(r,n,-1,k)] = E[(Y_{r}^{(k)})^{j}] = \frac{j}{(r-1)!\theta} \sum_{v=0}^{\infty} (-1)^{v} \frac{\alpha^{v} \Gamma\{(j+v) / \theta\} \Gamma\{((j+v) / \theta) - v + r\}}{v! k^{\{(j+v)/\theta\} - v} \beta^{(j+v)/\theta} \Gamma\{((j+v) / \theta) - v + 1\}}$$
(15)

where $Y_r^{(k)}$ denotes the k – th upper record values.

Remark 1. Putting $\alpha = 0$ in (11), we get the explicit expression for single moments of gos from the Weibull distribution as given by Kamps [1] pp-101.

Remark 2. Putting $\theta = 1$ in (11), the results for single moments of gos from exponential distribution with parameter $(\alpha + \beta)$ is deduced in the form

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

Remark 3. Setting $\theta = 1$ and $\alpha = 0$ in (11), we get the explicit expression for single moments of gos from the exponential distribution, established by Kamps [1] pp-101.

Remark 4. Setting $\alpha = 0$ in (15), the result for single moments of k – th upper record values is deduced for the Weibull distribution, which verifies the result obtained by Kamps [1] pp-101.

Remark 5. At $\theta = 1$ in (15), the result for single moments of k – th upper record values is deduced for the exponential distribution with parameter $(\alpha + \beta)$ in the form

$$E[X^{j}(r, n, -1, k)] = E[(Y_{r}^{(k)})^{j}] = \frac{\Gamma(j+r)}{(r-1)!k^{j}(\alpha+\beta)^{j}}$$

Remark 6. Setting $\theta = 1$ and $\alpha = 0$ in (15), the result for single moments of k – th upper record values is deduced for the exponential distribution as given by Kamps [1] pp-101.

Special cases

i) Putting m = 0 and k = 1 in (11), the exact expression for the single moments of order statistics from exponential-Weibull lifetime distribution can be obtained as

$$E(X_{r:n}^{j}) = \frac{jC_{r:n}}{\theta} \sum_{v=0}^{\infty} \sum_{u=0}^{r-1} (-1)^{u+v} {\binom{r-1}{u}} \frac{\alpha^{v} (n-r+u+1)^{v-1-(j+v)/\theta}}{v! \beta^{(j+v)/\theta}} \Gamma\{(j+v)/\theta\}$$
(16)

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

ii) Putting k = 1 in (15), we deduce the explicit formula for the single moments of upper records for exponential-Weibull lifetime distribution in the form

$$E[(Y_r^{(1)})^j] = E[X_{U(r)}^j] = \frac{j}{(r-1)!\theta} \sum_{v=0}^{\infty} (-1)^v \times \frac{\alpha^v \, \Gamma\{(j+v)\,/\,\theta\} \, \Gamma\{((j+v)\,/\,\theta) - v + r\}}{v!\,\beta^{(j+v)/\theta} \, \Gamma\{((j+v)\,/\,\theta) - v + 1\}}$$
(17)

Expressions (16) and (17) can be used to obtain the moments of order statistics and upper record values for arbitrary chosen values of α , β , θ and various sample size n = 1, 2, ..., 5. Some numerical computations for the first four moments of order statistics and upper record values from exponential-Weibull lifetime distribution are given in Table 1, 2, respectively.

Table 1. First four moments of order statistics

n	r		$\alpha = 1$,	$\theta = 3$			$\alpha = 1$	$\theta = 3$		
		$\beta = 1$				$\beta = 2$				
		E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	
1	1	0.4847	0.4257	0.4214	0.4564	0.4123	0.2975	0.2401	0.2111	
2	1	0.2524	0.1633	0.1207	0.0985	0.2285	0.1236	0.0756	0.0506	
2	2	0.7169	0.6880	0.7221	0.8143	0.5960	0.4714	0.4047	0.3716	
	1	0.1574	0.0835	0.0512	0.0349	0.1492	0.0672	0.0345	0.0195	
3	2	0.4425	0.3229	0.2598	0.2257	0.3871	0.2365	0.1578	0.1129	
	3	0.8541	0.8706	0.9533	1.1086	0.7005	0.5888	0.5282	0.5009	
	1	0.1074	0.0489	0.0260	0.0154	0.1056	0.0413	0.0186	0.0093	
4	2	0.3074	0.1873	0.1268	0.0932	0.2802	0.1447	0.0822	0.0502	
4	3	0.5775	0.4585	0.3928	0.3582	0.4939	0.3282	0.2334	0.1756	
	4	0.9463	1.0080	1.1401	1.3588	0.7693	0.6757	0.6265	0.6093	
	1	0.0776	0.0311	0.0147	0.0078	0.0785	0.0274	0.0110	0.0049	
	2	0.2266	0.1200	0.0712	0.0460	0.2138	0.0971	0.0487	0.0264	
5	3	0.4286	0.2882	0.2103	0.1640	0.3798	0.2163	0.1323	0.0860	
	4	0.6769	0.5721	0.5144	0.4877	0.5700	0.4028	0.3007	0.2354	
	5	1.0137	1.1169	1.2965	1.5766	0.8192	0.7439	0.7079	0.7028	
	r		$\alpha = 2$,	$\theta = 3$			$\alpha = 2$,	$\theta = 3$		
n		$\beta = 1$				$\beta = 2$				
		E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	
1	1	0.2796	0.2131	0.1882	0.1851	0.2524	0.1633	0.1207	0.0985	
2	1	0.1094	0.0567	0.0350	0.0245	0.1074	0.0489	0.0260	0.0154	
2	2	0.4497	0.3696	0.3415	0.3457	0.3974	0.2777	0.2155	0.1816	

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3	1	0.0568	0.0226	0.0109	0.0060	0.0584	0.0210	0.0089	0.0043
	2	0.2148	0.1250	0.0831	0.0613	0.2054	0.1048	0.0601	0.0377
	3	0.5671	0.4918	0.4706	0.4880	0.4934	0.3642	0.2932	0.2535
	1	0.0341	0.0110	0.0044	0.0020	0.0361	0.0107	0.0038	0.0015
4	2	0.1248	0.0573	0.0305	0.0182	0.1254	0.0517	0.0242	0.0125
4	3	0.3048	0.1927	0.1358	0.1044	0.2853	0.1578	0.0959	0.0630
	4	0.6545	0.5915	0.5823	0.6158	0.5628	0.4330	0.3589	0.3170
	1	0.0224	0.0066	0.0020	0.0008	0.0242	0.0061	0.0019	0.0006
	2	0.0806	0.0307	0.0137	0.0069	0.0837	0.0291	0.011	0.0052
5	3	0.1912	0.0973	0.0558	0.0351	0.1880	0.2163	0.0431	0.0236
	4	0.3806	0.2563	0.1891	0.1506	0.3502	0.2061	0.1312	0.0892
	5	0.7230	0.6753	0.6806	0.7321	0.6160	0.4897	0.4158	0.7028

We can note that the relation $\sum_{i=1}^{n} E(X_{i:n}^{j}) = nE(X^{j})$ (David and Nagaraja [22]) is satisfied here.

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		$\alpha = 1$,	$\theta = 3$			$\alpha = 1$,	$\theta = 3$		
n		β :	= 1		$\beta = 2$				
	E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	
1	0.56889	0.45576	0.43111	0.45468	0.49141	0.32772	0.25429	0.21836	
2	0.91044	0.95121	1.08956	1.3383	0.76444	0.65921	0.61777	0.61749	
3	1.14476	1.41461	1.85524	2.55571	0.94710	0.95972	1.02645	1.14873	
4	1.32344	1.84239	2.67656	4.03511	1.0857	1.23354	1.45714	1.78128	
5	1.46938	2.24072	3.53063	5.72872	1.19903	1.48698	1.90049	2.49615	
		$\alpha = 2$,	$\theta = 3$		$\alpha = 2, \theta = 3$				
n		β :	= 1		$\beta = 2$				
	E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	E(X)	$E(X^2)$	$E(X^3)$	$E(X^4)$	
1	0.39750	0.25454	0.20498	0.19060	0.36227	0.20163	0.13772	0.10710	
2	0.69970	0.60646	0.60059	0.65433	0.61746	0.45846	0.38254	0.34751	
3	0.93362	0.98443	1.13275	1.39520	0.80598	0.72009	0.69402	0.71042	
4	1.12136	1.36087	1.75728	2.38843	0.95367	0.97240	1.04633	1.17799	
5	1.27750	1.72650	2.44491	3.60436	1.07519	1.21276	1.42480	1.73455	

Now, we obtain the recurrence relations for single moments of exponential-Weibull lifetime distribution in the following theorem.

Theorem 1. For the distribution as given in (5) and $n \in N$, $m \in \Re$, $1 \le r \le n$, j = 0, 1, 2, ...

$$E(X^{j}(r,n,m,k)) = \frac{\alpha \gamma_{r}}{j+1} \{ E[X^{j+1}(r,n,m,k)] - E[X^{j+1}(r-1,n,m,k)] \} + \frac{\beta \theta \gamma_{r}}{j+\theta}$$

$$\times \{ E[X^{j+\theta}(r,n,m,k)] - E[X^{j+\theta}(r-1,n,m,k)] \}$$
(18)

Proof. From (2) and (6), we have

$$E[X^{j}(r,n,m,k)] = \frac{C_{r-1}}{(r-1)!} \left\{ \alpha \int_{0}^{\infty} x^{j} [\overline{F}(x)]^{\gamma_{r}} g_{m}^{r-1}(F(x)) \,\mathrm{d}\,x + \beta \theta \int_{0}^{\infty} x^{j+\theta-1} [\overline{F}(x)]^{\gamma_{r}} g_{m}^{r-1}(F(x)) \,\mathrm{d}\,x \right\}$$
(19)

Now (18) can be seen by noting that in view of Athar and Islam [23]

$$E[X^{j}(r,n,m,k)] - E[X^{j}(r-1,n,m,k)] = \frac{jC_{r-2}}{(r-1)!} \int_{0}^{\infty} x^{j-1} [\overline{F}(x)]^{\gamma_{r}} g_{m}^{r-1}(F(x)) \,\mathrm{d}\,x$$

Remark 7. Substituting m = 0, k = 1 in (18), we deduce the recurrence relation for single moments of order statistics from exponential-Weibull lifetime distribution in the form

$$E(X_{r:n}^{j}) = \frac{\alpha(n-r+1)}{j+1} \{ E(X_{r:n}^{j+1}) - E(X_{r-1:n}^{j+1}) \} + \frac{\beta\theta(n-r+1)}{j+\theta} \{ E(X_{r:n}^{j+\theta}) - E(X_{r-1:n}^{j+\theta}) \}$$

At $\theta = 1$, $\alpha = 0$, the result for single moments of order statistics is deduced for exponential distribution as given in Kamps [1] pp-122.

Remark 8. Putting m = -1 in (18), the result for single moments obtained by Khan et al. [24] for upper k – th record values from exponential-Weibull lifetime distribution is deduced.

Remark 9. Setting $\alpha = 0$ in (18), we get the recurrence relation for single moments of gos from the Weibull distribution, obtained by Khan et al. [25] for $j = j - \theta$.

Remark 10. Assuming $\beta = 0$ and $\alpha = 1$ in (18), the result for single moments of gos is deduced for exponential distribution, established by Pawlas and Syznal [26].

Remark 11. By putting $\theta = 2$ in (18), the result for single moments of gos obtained by Ahmad [27] with $\beta = v/2$ for linear failure rate distribution is deduced.

Remark 12. By putting $\alpha = 0$ and $\theta = 2$ in (18), the recurrence relation for single moments of gos is deduced for Rayleigh distribution in the form

$$E(X^{j}(r, n, m, k)) = \frac{2\beta\gamma_{r}}{j+2} \{ E[X^{j+2}(r, n, m, k)] - E[X^{j+2}(r-1, n, m, k)] \}$$

3 Relations for Product Moments

Theorem 2. For the given exponential-Weibull distribution in (5) and $n \in N$, $m \in \Re$, $1 \le r < s \le n$ and $i, j \ge 0$,

$$E[X^{i}(r,n,m,k)X^{j}(s,n,m,k)] = \frac{\alpha\gamma_{s}}{(j+1)} \{ E[X^{i}(r,n,m,k)X^{j+1}(s,n,m,k)] - E[X^{i}(r,n,m,k)X^{j+1}(s-1,n,m,k)] \} + \frac{\beta\gamma_{s}}{(j+\theta)} \times \{ E[X^{i}(r,n,m,k)X^{j+\theta}(s,n,m,k)] - E[X^{i}(r,n,m,k)X^{j+\theta}(s-1,n,m,k)] \}$$

$$(20)$$

Proof. From (3) and (6), we have

$$E[X^{i}(r,n,m,k)X^{j}(s,n,m,k)] = \frac{\alpha C_{s-1}}{(r-1)!(s-r-1)!} \int_{0}^{\infty} \int_{x}^{\infty} x^{i} y^{j} [\overline{F}(x)]^{m} f(x) g_{m}^{r-1}(F(x)) \times [h_{m}(F(y)) - h_{m}(F(x))]^{s-r-1} [\overline{F}(y)]^{\gamma_{s}} \, \mathrm{d} y \, \mathrm{d} x + \frac{\beta \theta C_{s-1}}{(r-1)!(s-r-1)!} \int_{0}^{\infty} \int_{x}^{\infty} x^{i} y^{j+\theta-1} \times [\overline{F}(x)]^{m} f(x) g_{m}^{r-1}(F(x)) [h_{m}(F(y)) - h_{m}(F(x))]^{s-r-1} [\overline{F}(y)]^{\gamma_{s}} \, \mathrm{d} y \, \mathrm{d} x$$
(21)

In view of Athar and Islam [23], note that

$$E[X^{i}(r,n,m,k)X^{j}(s,n,m,k)] - E[X^{i}(r,n,m,k)X^{j}(s-1,n,m,k)] = \frac{jC_{s-2}}{(r-1)!(s-r-1)!}$$

$$\times \int_{0}^{\infty} \int_{x}^{\infty} x^{i}y^{j-1}[\overline{F}(x)]^{m}f(x)g_{m}^{r-1}(F(x))[h_{m}(F(y)) - h_{m}(F(x))]^{s-r-1}[\overline{F}(y)]^{\gamma_{s}} dydx$$
(22)

Substitute (22) in (21) and after simplification, we get the result given in (20).

Remark 13. At i = 0 in (20), the recurrence relation for product moments reduces to relation for single moments as obtained in (18).

Remark 14. Putting m = 0 and k = 1 in (20), we obtain the recurrence relation for the product moments of order statistics from the exponential-Weibull lifetime distribution as

$$E(X_{r:n}^{i}X_{s:n}^{j}) = \frac{\alpha(n-s+1)}{(j+1)} \{ E(X_{r:n}^{i}X_{s:n}^{j+1}) - E(X_{r:n}^{i}X_{s-1:n}^{j+1}) \} + \frac{\beta(n-s+1)}{(j+\theta)} \times \{ E(X_{r:n}^{i}X_{s:n}^{j+\theta}) - E(X_{r:n}^{i}X_{s-1:n}^{j+\theta}) \}$$

At $\theta = 1$, $\alpha = 0$, the relation for product moments of order statistics is deduced for exponential distribution as

$$E(X_{r:n}^{i}X_{s:n}^{j}) = \frac{\beta(n-s+1)}{(j+1)} \{ E(X_{r:n}^{i}X_{s:n}^{j+1}) - E(X_{s:n}^{i}X_{s-1:n}^{j+1}) \}$$

Remark 15. Setting m = -1 in Theorem 2, the relation for the product moments in Khan et al. [24] for upper k – th record values from the exponential-Weibull lifetime distribution is deduced.

Remark 16. Setting $\alpha = 0$ in (20), we get the recurrence relation for the product moments of gos from the Weibull distribution as obtained by Khan et al. [25] for $j = j - \theta$.

Remark 17. Assuming $\beta = 0$ and $\alpha = 1$ in (20), the result for the product moments of gos is deduced for the exponential distribution, established by Pawlas and Syznal [26].

Remark 18. By putting $\theta = 2$ in (20), the result for product moments of gos obtained by Ahmad [27] with $\beta = v/2$ for linear failure rate distribution is deduced.

Remark 19. By putting $\alpha = 0$ and $\theta = 2$ in (20), the recurrence relation for the product moments of gos is deduced for Rayleigh distribution in the form

$$\begin{split} E[X^{i}(r,n,m,k)X^{j}(s,n,m,k)] &= \frac{2\beta\gamma_{s}}{(j+2)} \{ E[X^{i}(r,n,m,k)X^{j+2}(s,n,m,k)] \\ &- E[X^{i}(r,n,m,k)X^{j+2}(s-1,n,m,k)] \} \end{split}$$

4 Characterizations

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

$$f_{X(s,n,m,k)|X(r,n,m,k)}(y \mid x) = \frac{C_{s-1}}{(s-r-1)!C_{r-1}} [\overline{F}(x)]^{m-\gamma_r+1} \times [h_m(F(y)) - h_m(F(x))]^{s-r-1} [\overline{F}(y)]^{\gamma_s-1} f(y), x < y$$
(23)

Theorem 3. Let X be a non-negative random variable having an absolutely continuous df = F(x) with F(0) = 0 and $0 \le F(x) \le 1$ for all x > 0, then

$$E[\xi\{X(s,n,m,k)\} \mid X(l,n,m,k) = x] = e^{-(\alpha x + \beta x^{\theta})} \prod_{j=1}^{s-l} \left(\frac{\gamma_{l+j}}{\gamma_{l+j} + 1}\right), l = r, r+1$$
(24)

if and only if

$$\overline{F}(x) = e^{-(\alpha x + \beta x^{\theta})}, x > 0, \alpha > 0, \beta > 0, \theta > 0$$
(25)

where $\xi(y) = e^{-(\alpha y + \beta y^{\theta})}$.

Proof. We have from (23) for s > r + 1,

$$E[\xi\{X(s,n,m,k)\} \mid X(r,n,m,k) = x] = \frac{C_{s-1}}{(s-r-1)!C_{r-1}(m+1)^{s-r-1}} \times \int_{x}^{\infty} e^{-(\alpha y + \beta y^{\theta})} \left(\frac{\overline{F}(y)}{\overline{F}(x)}\right)^{r_{s}-1} \left[1 - \left(\frac{\overline{F}(y)}{\overline{F}(x)}\right)^{m+1}\right]^{s-r-1} \frac{f(y)}{\overline{F}(x)} dy$$
(26)

By setting $u = \frac{\overline{F}(y)}{\overline{F}(x)} = \frac{e^{-(\alpha y + \beta y^{\theta})}}{e^{-(\alpha x + \beta x^{\theta})}}$ from (5) in (26), we obtain

$$E[\xi\{X(s,n,m,k)\} \mid X(r,n,m,k) = x] = \frac{C_{s-1} e^{-(\alpha x + \beta x^{\nu})}}{(s-r-1)!C_{r-1}(m+1)^{s-r-1}} \times \int_{0}^{1} u^{\gamma_{s}} (1-u^{m+1})^{s-r-1} \, \mathrm{d} \, u \tag{27}$$

Again by setting $t = u^{m+1}$ in (27), we get

$$E[\xi\{X(s,n,m,k)\} \mid X(r,n,m,k) = x] = \frac{C_{s-1} e^{-(\alpha x + \beta x^{\theta})}}{(s-r-1)! C_{r-1}(m+1)^{s-r}} \int_0^1 t^{\frac{\gamma_s + 1}{m+1} - 1} (1-t)^{s-r-1} dt$$

and hence the necessary part given in (24).

To prove the sufficient part, we have from (23) and (24)

$$\frac{C_{s-1}}{(s-r-1)!C_{r-1}(m+1)^{s-r-1}} \int_{x}^{\infty} e^{-(\alpha y+\beta y^{\theta})} [(\overline{F}(x))^{m+1} - (\overline{F}(y))^{m+1}]^{s-r-1} \times [\overline{F}(y)]^{\gamma_{s}-1} f(y) \,\mathrm{d}\, y$$

$$= H_{s|r}(x) [\overline{F}(x)]^{\gamma_{r+1}}$$
(28)

where

$$H_{s|r}(x) = e^{-(\alpha x + \beta x^{\theta})} \prod_{j=1}^{s-r} \left(\frac{\gamma_{r+j}}{\gamma_{r+j} + 1} \right)$$

Differentiating (28) both sides with respect to x, we get

$$\begin{aligned} & \frac{C_{s-1}[F(x)]^m f(x)}{(s-r-2)! C_{r-1}(m+1)^{s-r-2}} \int_x^\infty e^{-(\alpha y+\beta y^{\theta})} [(\overline{F}(x))^{m+1} - (\overline{F}(y))^{m+1}]^{s-r-2} \\ & \times [\overline{F}(y)]^{\gamma_s-1} f(y) \,\mathrm{d}\, y = H_{s|r}'(x) [\overline{F}(x)]^{\gamma_{r+1}} - \gamma_{r+1} H_{s|r}(x) [\overline{F}(x)]^{\gamma_{r+1}-1} f(x) \end{aligned}$$

or

$$-\gamma_{r+1}H_{s|r+1}(x)[\overline{F}(x)]^{\gamma_{r+2}+m}f(x) = H_{s|r}'(x)[\overline{F}(x)]^{\gamma_{r+1}} - \gamma_{r+1}H_{s|r}(x)[\overline{F}(x)]^{\gamma_{r+1}-1}f(x)$$

Therefore,

$$\frac{f(x)}{\overline{F}(x)} = -\frac{H'_{s|r}(x)}{\gamma_{r+1}[H_{s|r+1}(x) - H_{s|r}(x)]} = \alpha + \beta \theta \, x^{\theta - 1}$$

which proves that

$$F(x) = 1 - e^{-(\alpha x + \beta x'')}, x > 0, \alpha > 0, \beta > 0, \theta > 0$$

Remark 20. At $m \rightarrow -1$ in (24), we get the characterization results from the exponential-Weibull distribution based on k – th upper record values.

Remark 21. Setting m = 0, k = 1 in (24), we obtain the characterization results of the exponential-Weibull lifetime distribution based on order statistics.

Following theorem contains characterization of this distribution by a recurrence relation for the single moments of gos.

Theorem 4. Fix a positive integer k and let j be a non-negative integer. A necessary and sufficient condition for a random variable X to be distributed with pdf given by (4) is that

$$E(X^{j}(r,n,m,k)) = \frac{\alpha \gamma_{r}}{j+1} \{ E[X^{j+1}(r,n,m,k)] - E[X^{j+1}(r-1,n,m,k)] \} + \frac{\beta \theta \gamma_{r}}{j+\theta}$$

$$\times \{ E[X^{j+\theta}(r,n,m,k)] - E[X^{j+\theta}(r-1,n,m,k)] \}$$
(29)

Proof. The necessary part follows immediately from equation (18). On the other hand if the relation in (29) is satisfied, then on using (2), we have

$$\begin{split} & \frac{C_{r-1}}{(r-1)!} \int_{0}^{\infty} x^{j} [\overline{F}(x)]^{\gamma_{r}-1} f(x) g_{m}^{r-1}(F(x)) \,\mathrm{d}\, x \\ &= \frac{\alpha \, C_{r-1}}{(r-1)!(j+1)} \int_{0}^{\infty} x^{j+1} [\overline{F}(x)]^{\gamma_{r}} f(x) g_{m}^{r-2}(F(x)) \left\{ \frac{\gamma_{r} g_{m}(F(x))}{\overline{F}(x)} - (r-1) [\overline{F}(x)]^{m} \right\} \mathrm{d}\, x \\ &+ \frac{\beta \theta \, C_{r-1}}{(r-1)!(j+\theta)} \int_{0}^{\infty} x^{j+\theta} [\overline{F}(x)]^{\gamma_{r}} f(x) g_{m}^{r-2}(F(x)) \left\{ \frac{\gamma_{r} g_{m}(F(x))}{\overline{F}(x)} - (r-1) [\overline{F}(x)]^{m} \right\} \mathrm{d}\, x \end{split}$$

Let

$$h(x) = -[(\overline{F}(x)]^{\gamma_r} g_m^{r-1}(F(x))$$
(30)

Differentiating both sides of (30), we get

$$h'(x) = [\overline{F}(x)]^{\gamma_r} f(x) g_m^{r-2}(F(x)) \begin{cases} \frac{\gamma_r g_m(F(x))}{\overline{F}(x)} - (r-1)[\overline{F}(x)]^m \\ \end{cases}$$

Thus,

$$\frac{C_{r-1}}{(r-1)!} \int_{0}^{\infty} x^{j} [\overline{F}(x)]^{\gamma_{r}-1} f(x) g_{m}^{r-1}(F(x)) \,\mathrm{d} x$$

$$= \frac{\alpha C_{r-1}}{(r-1)!(j+1)} \int_{0}^{\infty} x^{j+1} h'(x) \,\mathrm{d} x + \frac{\beta \theta C_{r-1}}{(r-1)!(j+\theta)} \int_{0}^{\infty} x^{j+\theta} h'(x) \,\mathrm{d} x$$
(31)

Integrating right hand side in (31) by parts and using the value of h(x) from (30), we find that

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$$\frac{C_{r-1}}{(r-1)!} \int_0^\infty x^j [\overline{F}(x)]^{\gamma_r - 1} g_m^{r-1}(F(x)) \{ (\alpha + \beta \theta \, x^{\theta - 1}) \overline{F}(x) - f(x) \} \, \mathrm{d}\, x = 0$$
(32)

Applying the extension of Müntz-Szász Theorem, (see for example Hwang and Lin [28]) to (32), we get $f(x) = (\alpha + \beta \theta x^{\theta-1})\overline{F}(x)$

which proves that f(x) has the form as in (6).

Remark 22. Theorem 4 can be used to characterize the exponential, Weibull, linear failure rate and Rayleigh distributions by setting $\beta = 0$, $\alpha = 0$, $\theta = 2$ and $\alpha = 0$, $\theta = 2$, respectively.

Acknowledgments. The authors acknowledge with thanks to the referees and the Editor-in-Chief for their fruitful suggestions and comments which led to the overall improvement in the manuscript.

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