

Some Distributional Properties of Record Values From Generalized Extreme Value Distributions

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Abstract

Several distributional properties of the record values from generalized extreme value distribution are discussed. Minimum variance unbiased estimates for location and scale parameters are presented.

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1 Introduction

A random variable X is said to have the generalized extreme value (GEV) distribution if its cumulative distribution function (cdf) $F(x)$ is of the following form

$$F(x) = \exp\{-(1 - \gamma\sigma^{-1}(x - \mu))^{1/\gamma}\} \quad (1)$$

Where $\gamma \neq 0$, and

$$x < \mu + \sigma\gamma^{-1}, \gamma > 0, \sigma > 0$$

$$x > \mu + \sigma\gamma^{-1}, \gamma < 0, \sigma > 0$$

GEV $(0, 1, -1)$ is unimodal. Gumbel (1958) has given various distributional properties of extreme value distributions. Fisher and Tippet (1925) introduced the extreme value distributions as the limiting distributions of largest and smallest observations. Extreme value distributions have been used in the analysis of data concerning floods, extreme sea levels and air pollution problems. For details see Gumbel (1958), Horwitz (1980), Jenkinson (1955) and Roberts (1979).

Suppose that $X_1, X_2, \dots$ is a sequence of independent and identically distributed random variables with cumulative distribution function (cdf) $F(x)$ and the corresponding pdf $f(x)$.

Let $Y_n = \max(\min) \{X_1, X_2, \dots, X_n\}$. We say X_j is an upper (lower) record value of $\{X_n, n \geq 1\}$ if $Y_j > (<) Y_{j-1}, j > 1$. By definition X_1 is an upper as well as a lower record value. In this paper we will consider the lower record values of the generalized extreme value distribution. For theory of record values, see Ahsanullah (1995, 2004), Arnold, Balakrishnan and Nagaraja (1998) and Nevzorov (2001).

2 Distributional properties of mth lower record value

Let $X_{L(m)}$ be the mth lower record value of $\{X_n, n \geq 1\}$.

Lemma 2.1

$$P[X_{L(m)} \leq x] = \int_{-\infty}^x \frac{(h(u))^{m-1}}{\Gamma(m)} f(u) du \quad (3)$$

where $H(u) = -\ln F(u), 0 < F(u) < 1$.

Proof:

$$\begin{aligned} F_1(x) &= P[X_{L(1)} \leq x] = P[X_1 \leq x] = F(x) \\ F_2(x) &= P[X_{L(2)} \leq x] = \int_{-\infty}^x \int_y^{\infty} \sum_{j=1}^{\infty} (1 - F(u))^{j-1} f(u) f(y) du dy \\ &= \int_{-\infty}^x \int_y^{\infty} \frac{f(u)}{F(u)} f(y) du dy \\ &= \int_{-\infty}^x H(y) f(y) dy \\ F_3(x) &= P[X_{L(3)} \leq x] = \int_{-\infty}^x \int_y^{\infty} \sum_{j=1}^{\infty} (1 - F(u))^{j-1} H(u) f(u) f(y) du dy \\ &= \int_{-\infty}^x \int_y^{\infty} H(u) \frac{f(u)}{F(u)} f(y) du dy \\ &= \int_{-\infty}^x \frac{(H(y))^2}{2!} f(y) dy \end{aligned}$$

The result follows by induction.

If $X \in GEV(\mu, \sigma, \gamma)$, then using (1) and (2) we can write the pdf $F_{(m)}(x)$ of the mth lower record value as

$$f_{(m)}(x) = \{1 - \gamma \sigma^{-1}(x - \mu)\}^{(m-1)/\gamma} f_{(m)}^*(x),$$

Where

$$f_{(m)}^*(x) = \frac{\{1 - \gamma\sigma^{-1}(x - \mu)\}^{\frac{1-\gamma}{\gamma}}}{\sigma \Gamma(m)} \exp\left\{-(1 - \gamma\sigma^{-1})(x - \mu)^{1/2}\right\}, \gamma \neq 0$$

and

$$f_{(m)}(x) = \frac{e^{-m(x-\mu)/\sigma}}{\sigma \Gamma(m)} \exp\left(-e^{-(x-\mu)/\sigma}\right), \quad m = 1, 2, \dots, \gamma = 0.$$

It can be shown that

$$X_{L(m)}^d \mu + \sigma \gamma^{-1} \{1 - (W_1 + W_2 + \dots + W_m)^\gamma\} \dots \quad (4)$$

For $\gamma \neq 0$

$$X_{L(m)}^d x - \sigma \left(W_1 + \frac{W_2}{2} + \dots + \frac{W_m}{m}\right) \dots \quad (5)$$

For $\gamma = 0$,

where $W_1, W_2, \dots, W_m$ are independently distributed as exponential with cdf $F(x) = 1 - e^{-x}, x \geq 0$ and $X_{L(m)}^d$ GEV $(\mu, \sigma, 0)$.

Using (4), we obtain for $\gamma \neq 0$,

$$E(X_{L(i)}) = \mu + \sigma \gamma^{-1} [1 - \frac{\Gamma(i + \gamma)}{\Gamma(i)}] = \mu + \sigma \alpha, \quad i > -\gamma, \quad (6)$$

$$Var(X_{L(m)}) = \mu^2 \gamma^{-2} \{E(W_1 + W_2 + \dots + W_i)^{2\gamma} - (E(W_1 + W_2 + \dots + W_i)^\gamma)^2\}$$

$$= \mu^2 \gamma^{-2} [\frac{\Gamma(i + 2\gamma)}{\Gamma(i)} - (\frac{\Gamma(i + \gamma)}{\Gamma(i)})^2], i > -2\gamma \quad (7)$$

For $r < s$

$$\begin{aligned} \gamma^2 \sigma^{-2} \text{cov}(x_{L(r)} x_{L(s)}) &= E(\sum_{j=1}^r w_j^\gamma) \sum_{j=1}^s w_j^\gamma - E(\sum_{j=1}^r w_j^\gamma) (E \sum_{j=1}^s w_j^\gamma) \\ &= \int_0^\infty \int_0^\infty u^\gamma (u+v)^\gamma \frac{u^{\gamma-1} e^{-u}}{\Gamma(r)} \cdot \frac{v^{s-\gamma-1} e^{-v}}{\Gamma(m-\gamma)} du dv - \frac{\Gamma(r+\gamma)}{\Gamma(r)} \frac{\Gamma(s+\gamma)}{\Gamma(s)} \\ &= \frac{\Gamma(r+\gamma)}{\Gamma(r)} \cdot \frac{\Gamma(s+2\gamma)}{\Gamma(s)} - \frac{\Gamma(r+\gamma)}{\Gamma(r)} \cdot \frac{\Gamma(s+\gamma)}{\Gamma(s)}, \end{aligned}$$

Since u and v are independent.

Thus

$$\text{cov}(X_{L(r)} X_{L(s)}) = \sigma_0^2 a_r b_s, r < s,$$

Where

$$\sigma_0^2 = \frac{\sigma^2}{\gamma^2}, a_r = \frac{\Gamma(r+\gamma)}{\Gamma(r)} \text{ and } b_s = \frac{\Gamma(s+2\gamma)}{\Gamma(s/\gamma)} - \frac{\Gamma(s+\gamma)}{\Gamma(s)}$$

For $\gamma = 0$, we obtain from (5)

$$\begin{aligned} E(X_{L(r)}) &= \mu + \alpha_r \sigma, \\ Var(X_{L(r)}) &= \sigma^2 V_{r,r}^*, \\ cov(X_{L(r)} X_{L(m)}) &= Var(X_{L(m)}), r < m \end{aligned}$$

with

$$\begin{aligned} \alpha_1^* &= \nu, \text{ the Euler's constant} \\ \alpha_j^* &= \alpha_{j-1}^* - \frac{1}{j-1}, j \geq 2 \\ V_{11}^* &= \frac{x^2}{6} \\ V_{jj}^* &= V_{j-1,j-1} - \frac{1}{(j-1)^2}, j \geq 2. \end{aligned}$$

3 Estimation of parameters

3.1 Minimum variance linear unbiased estimate (MVLUE) of μ and σ for known γ .

If $\gamma > 0$, then the variances of all the lower records are finite and we will consider the first m lower records. If $\gamma < 0$, then we have to consider record values greater than -2γ . Without any loss of generality, we will take $\gamma > 0$ and consider first m lower record values.

Theorem 3.1.

Suppose $\gamma > 0$, then the MVLUE $\hat{\mu}$ and $\hat{\sigma}$ based on the first m lower record values $r_1, r, \dots, r_m$ are given by

$$\hat{\mu} = -\alpha' V^{-1} (1\alpha' - \alpha 1') V^{-1} r / \Delta \quad (8)$$

and

$$\hat{\sigma} = -1 V^{-1} (1\alpha' - \alpha 1' V^{-1} r / \Delta, \quad (9)$$

$$\begin{aligned} \text{where } \Delta &= (1' V^{-1} 1)^{-1} (\alpha' V^{-1} \alpha) - (1' V^{-1} \alpha)^2, \\ V^{-1} &= (V^{ij}), \\ V^{uu} &= \left(\frac{1+\gamma}{\gamma^2} \right)^2 \frac{1}{\Gamma(1+2\gamma)}, V^{mm} = \frac{m-1+\gamma}{\gamma^2} \frac{b_{m-1}}{b_m} \frac{\Gamma(m)}{\Gamma(m-1+\gamma)} \\ V^{ii} &= \frac{\Gamma(i)}{\Gamma(i+2\gamma)} \cdot \frac{(1+\gamma)^2 + (i-1)(i-1+2\gamma)}{\gamma^2}, j = i+1, i = 1, \dots, m-1 \\ V^{ij} &= 0, |i-j| > 1 \end{aligned}$$

$$1' = (1, 1, \dots, 1), \quad r' = (\gamma_1, \gamma_2, \dots, \gamma_m), \quad \alpha^1 = (\alpha_1, \alpha_2, \dots, \alpha_m).$$

and

$$\alpha_i = 1 - \frac{\Gamma(i + \gamma)}{\Gamma(i)}.$$

The variances and covariances are given by

$$\text{Var}(\hat{\mu}) = \frac{\sigma_0^2}{\Delta} (\alpha' V^{-1} \alpha)$$

$$\text{Var}(\hat{\sigma}) = \frac{\sigma_0^2}{\Delta} (1' V^{-1} 1)$$

and

$$\text{Cov}(\hat{\mu}, \hat{\sigma}) = -\frac{\sigma_0^2}{\Delta} (1' V^{-1} \alpha).$$

Proof:

Let $R = (X_{L(1)}, X_{L(2)}, \dots, X_{L(m)})$, then we can write

$$E(R) = \mu 1 + \sigma_o \alpha$$

and

$$V(R) = \sigma_0^2 V$$

where

$$\alpha' = (a_1, \dots, a_m), \quad \alpha_i = 1 - \frac{\Gamma(i + \gamma)}{\Gamma(i)}, \quad 1' = (1, \dots, 1), \quad V = (v_{ij}), \quad V^{-1} = (V^{ij}).$$

$$v_{ij} = a_i b_j, \quad 1 \leq i, j \leq m.$$

$$\begin{aligned} V^{11} &= \frac{a_2}{a_1(a_2 b_1 - a_1 b_2)} = \frac{(1+\gamma)^2}{\gamma^2} \frac{1}{\Gamma(1+2\gamma)}, \\ V^{mm} &= \frac{b_{m-1}}{b_m} \frac{1}{a_m b_{m-1} - a_{m-1} b_m} = \frac{b_{m-1}(m-1+\gamma)}{b_m \gamma^2} \cdot \frac{\Gamma(m)}{\Gamma(m-1+\gamma)}, \\ V^{ii} &= \frac{a_{i+1} b_{i-1} - a_{i-1} b_{i+1}}{(a_i b_{i-1} - a_{i-1} b_i)(a_{i+1} b_i - a_i b_{i+1})} \\ &= \frac{\Gamma(i)}{\Gamma(i+2\gamma)} \cdot \frac{(i+\gamma)^2 + (i-1)(i-1+2\gamma)}{\gamma^2}, \quad i = 2 \dots m-1, \\ V^{ij} &= V^{ji} = -\frac{1}{a_{i+1} b_i - a_i b_{i+1}} = -\frac{\Gamma(i)}{\Gamma(i+2\gamma)} \cdot \frac{i+\gamma}{\gamma^2}, \quad j = i+1, \quad i = 1, \dots, m-1 \end{aligned}$$

and $V^{ij} = 0, |i - j| > 1$.

Using the method of Lloyd (1952) the MVLUEs $\hat{\mu}$ and $\hat{\sigma}$ of μ and σ based on m lower record values $r_1, r_2, \dots, r_m$ are given by

$$\hat{\mu} = -a'V^{-1} (1a' - a1') V^{-1}r/\Delta$$

and

$$\hat{\sigma} = -1'V^{-1} (1a' - a1') V^{-1}r/\Delta,$$

where $\Delta = (1'V^{-1}1) (a'V^{-1}a) - (1'V^{-1}a)^2$.

$$Var(\hat{\mu}) = \frac{\sigma_0^2}{\Delta} [a'V^{-1}a],$$

$$Var(\hat{\sigma}) = \frac{\sigma_0^2}{\Delta} (1'V^{-1}1)$$

$$\text{and } Cov(\hat{\mu}, \hat{\sigma}) = -\frac{\sigma_0^2 1'V^{-1}a}{\Delta}$$

The following table gives the coefficients of the MVLUEs $\hat{\mu}$ and $\hat{\sigma}$ for $2 \leq m \leq 6$ and $\gamma = 1$.

Table 3.1. the coefficients of $\hat{\mu}$, $\hat{\sigma}$ for $\gamma=2$

m		$X_{L(1)}$	$X_{L(2)}$	$X_{L(3)}$	$X_{L(4)}$	$X_{L(5)}$	$X_{L(6)}$
2	$\hat{\mu}$	1.25	-0.25				
2	$\hat{\sigma}$	0.25	-0.25				
3	$\hat{\mu}$	1.1458	-0.0764	-0.0972			
3	$\hat{\sigma}$	0.10410	-0.0069	-0.0972			
4	$\hat{\mu}$	1.125	-0.075	-0.025	-0.025		
4	$\hat{\sigma}$	0.0592	-0.0040	-0.0013	-0.0540		
5	$\hat{\mu}$	1.1195	-0.0746	-0.0249	-0.0107	-0.0093	
5	$\hat{\sigma}$	0.0386	-0.0026	-0.0009	-0.0004	-0.0348	
6	$\hat{\mu}$	1.1182	-0.0746	-0.0249	-0.0107	-0.0053	-0.0028
6	$\hat{\sigma}$	0.0273	-0.0018	-0.0006	-0.0003	-0.0001	-0.0245

3.2 MVLUEs $\hat{\mu}$ and $\hat{\sigma}$ of μ and σ when $\gamma = 0$

The following Theorem gives MVLUEs $\hat{\mu}$ and $\hat{\sigma}$.

Theorem 3.2

Suppose that $\gamma = 0$, then the MVLUEs $\hat{\mu}$ and $\hat{\sigma}$ of μ and σ based on the m observed lower records, $r_1, \dots, r_n$, are respectively

$$\hat{\mu} = r_m - a_m^* \hat{\sigma} \quad (10)$$

and

$$\hat{\sigma} = \frac{1}{m-1} \sum_{i=1}^{m-1} r_i - r_m, \quad (11)$$

where $a_1^* = v$, the Euler constant ($= 0.57722$)

and

$$a_j^* = a_{j-1}^* - \frac{1}{j-1}, \quad j \leq 2.$$

Their corresponding variances and covariance are

$$Var(\hat{\mu}) = \sigma^2 \left\{ V_{mm}^* + \frac{(V_m^*)^2}{m-1} \right\}$$

$$Var(\hat{\sigma}) = \frac{\sigma^2}{m-1}$$

and

$$Cov(\hat{\mu}, \hat{\sigma}) = -\frac{\sigma^2 V_m^*}{m-1}.$$

Proof:

For $\gamma = 0$, $E(X_{L(j)}) = \mu + \sigma a_j^*$,
 $a_1^* = v$, the Euler constant ($= 0.57722$)
 and

$$\begin{aligned} a_j^* &= a_{j-1}^* - \frac{1}{j-1}, \quad j \leq 2. \\ Var(X_{L(j)}) &= \sigma^2 V_{jj}^*, \quad j = 1, 2, \dots \\ Cov(X_{L(i)} X_{L(j)}) &= Var(X_{L(j)}), \quad 1 \leq i \leq j, \\ V_{11}^* &= \frac{\pi^2}{6}, V_{jj}^* = V_{j-1, j-1}^* - \frac{1}{(j-1)^2}. \end{aligned}$$

Let $V = (V_{ij}^*)$ and $V^{-1} = (V^{ij})$, then

$$\begin{aligned} 1'V^{-1} &= (0, 0, \dots, 1/V_{mm}^*), \\ a'V^{-1} &= \left(1, 1, \dots, \frac{a_m}{V_{mm}^*} - m - 1\right) \\ a'V^{-1}1 &= a_m/V_{mm}^* \\ a'V^{-1}a &= \frac{a_m^2}{V_{mm}^*} \end{aligned}$$

and

$$\Delta = \frac{m-1}{V_{mm}^*}.$$

Substituting these values in

$$\hat{\mu} = -a'V^{-1} (1a' - a1') V^{-1}r/\Delta$$

and

$$\hat{\sigma} = -1'V^{-1} (1a' - a1') V^{-1}r/\Delta,$$

the result follows.

Table 3.2 gives the coefficients of the MVLUEs $\hat{\mu}$ and $\hat{\sigma}$ for $2 \leq m \leq 6$ and $\gamma = 0$.

Table 3.2. The coefficients of the MVLUEs $\hat{\mu}$ and $\hat{\sigma}$ for $\gamma = 0$.

m		$X_{L(1)}$	$X_{L(2)}$	$X_{L(3)}$	$X_{L(4)}$	$X_{L(5)}$	$X_{L(6)}$
2	$\hat{\mu}$	0.42278	0.57722				
2	$\hat{\sigma}$	1	-1				
3	$\hat{\mu}$	0.46139	0.461239	0.07728			
3	$\hat{\sigma}$	$\frac{1}{2}$	$\frac{1}{2}$	-1			
4	$\hat{\mu}$	0.41403	0.41403	0.41403	-0.25611		
4	$\hat{\sigma}$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	-1		
5	$\hat{\mu}$	0.37653	0.37653	0.37653	0.37653	-0.50611	
5	$\hat{\sigma}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	-1	
6	$\hat{\mu}$	0.32122	0.32122	0.32122	0.32122	0.32122	0.60610
6	$\hat{\sigma}$	0.2	0.2	0.2	0.2	0.2	-1

3.3 Estimate of γ for known values of μ and σ .

We will assume without any loss of generality $\mu = 0$ and $\sigma = 1$.

Let $Y_i = i(X_{L(i)} - X_{L(i+1)})$, then

$E(Y_i) = 1 - \gamma E(X_{L(i)})$ and $E(\sum_{i=1}^{m-1} Y_i) = m - 1 - \gamma E(\sum_{i=1}^{m-1} X_{L(i)})$.

Thus a moment estimator γ^* of γ is

$$\gamma^* = \frac{1 - \bar{y}}{\bar{r}},$$

where

$$\bar{y} = \sum_{i=1}^{m-1} \frac{Y_i}{m-1} \text{ and } \bar{r} = \sum_{i=1}^{m-1} \frac{X_{L(i)}}{m-1}.$$

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