

# Finite mixture of extreme value distribution for rainfall frequency analysis

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**सार** — 3-4 वृद्धिर्वासी (आउटलायर्स) से युक्त भुज स्टेशन (गुजरात) में (70 वर्षों) के 24 घंटेवार अधिकतम वर्षा अनुक्रमों के आवृत्ति विश्लेषण के लिए ई वी<sub>1</sub> और ई वी<sub>2</sub> वंशों के मिश्रण का प्रयास किया गया। दूसरे मामले के अध्ययन में ई वी<sub>1</sub> और ई वी<sub>2</sub> के मिश्रण का प्रयोग, दिन हाटे (पश्चिम बंगाल) की (58 वर्षों) चरम वर्षा की व्याख्या करने के लिए किया गया। ई वी वंशों के परिमित मिश्रण उन सब मामलों में सही बैठते हैं जहाँ पर ई वी<sub>1</sub> अकेला प्रत्यागमन काल मानों का अधिक या कम आकलन करता है।

**ABSTRACT.** A mixture of EV<sub>1</sub> and EV<sub>2</sub> distributions has been tried for the frequency analysis of 24-hourly maximum rainfall series (70 years) at Bhuj containing 3-4 outliers. In a second case study, a mixture of EV<sub>1</sub> and EV<sub>2</sub> is used to describe extreme rainfall (58 years) of Dinhat. A finite mixture of EV distributions provides a better fit in all those cases where EV<sub>1</sub> alone over or under estimates return period values.

## 1. Introduction

The main distributions used for analysis of annual maximum precipitations series are log normal, extreme value types 1, 2 & 3 (EV<sub>1</sub>, EV<sub>2</sub> & EV<sub>3</sub>), Pearson and log Pearson type III. Though the observed series may satisfy the criteria for accuracy, consistency, homogeneity, stationarity, randomness and sufficiency of record length, a problem which affects the applicability of these distributions is the presence of outliers. In the present paper, attempt has been made to combine EV<sub>1</sub>, linearly with EV<sub>2</sub> to tackle the outliers and with EV<sub>3</sub> to tackle the cases exhibiting small range of extremes. The main papers consulted are by Boes (1966), Chow (1953), Gumbel (1941) and Jenkinson (1955).

## 2. Mixture of two distributions

### List of symbols used

- $E_1(x)$  — Distribution function of extreme value type 1
- $E_2(x)$  — Distribution function of extreme value type 2
- $E_3(x)$  — Distribution function of extreme value type 3
- $E(x)$  — Distribution function of mixed distribution
- $E_n(x)$  — Sample distribution function
- $\theta$  — Mixing proportion
- $u, a, k$  — Parameters of extreme value distribution
- $y$  — Reduced variate of EV distribution
- $x_1, x_2, x_3, x_T$  —  $T$ -year value in respect of EV<sub>1</sub>, EV<sub>2</sub>, EV<sub>3</sub> and mixed distribution respectively

$v(x_1), v(x_2), v(x_3)$  &  $v(x_T)$  — Variances of statistics  $x_1, x_2, x_3$  &  $x_T$

SE ( $x_T$ ) — Standard error of  $x_T$ .

2.1. If  $E_1(x)$  and  $E_2(x)$  are two distribution functions, a linear combination may be defined as:

$$E_\theta(x) = (\frac{1}{2} + \theta) E_1(x) + (\frac{1}{2} - \theta) E_2(x) \quad (1)$$

where,  $-\frac{1}{2} \leq \theta \leq \frac{1}{2}$

$$\therefore \theta = \frac{E_\theta(x) - \frac{1}{2} [E_1(x) + E_2(x)]}{E_1(x) - E_2(x)} \quad (2)$$

As the nature of prior information about  $E_\theta(x)$  is generally not known, it may be estimated by empirical distribution function  $E_n(x)$  of the observed sample ( $x_1, x_2, \dots, x_n$ ). An estimator for  $\theta$  is:

$$\hat{\theta} = \frac{E_n(x) - \frac{1}{2} [E_1(x) + E_2(x)]}{E_1(x) - E_2(x)} \quad (3)$$

$\hat{\theta}$  provides unbiased and consistent estimate of  $\theta$  with variance of order  $(1/n)$ . However, when  $\hat{\theta}$  lies outside the interval  $(-1/2, 1/2)$ , we may truncate the values going beyond  $-1/2$  or  $1/2$ , although by doing so unbiasedness of the estimator is lost.

It may be seen that the sample observations form a discrete series.  $E_n(x)$  represents a step deviation function, whereas  $E_1(x)$  and  $E_2(x)$  form continuous curves. A mean curve for  $E_n(x)$  should be estimated for obtaining  $\theta$ .

### 2.2. Mixing of two $EV_1$ distributions with different parameters

Situations may arise when a sample of observations is considered in two parts one drawn from population  $E_1(x; u_1, \alpha_1)$  and the other from  $E_1(x; u_2, \alpha_2)$ . The entire series may be assumed to have come from the mixed distribution:

$$E_\theta(x) = \left(\frac{1}{2} + \theta\right) E_1(x; u_1, \alpha_1) + \left(\frac{1}{2} - \theta\right) E_1(x; u_2, \alpha_2) \quad (4)$$

For large samples  $E_\theta(x)$  may be estimated by  $E_n(x)$  and  $\theta$  by  $\hat{\theta}$  given as :

$$\hat{\theta} = \frac{E_n(x) - \frac{1}{2} [E_1(x; u_1, \alpha_1) + E_1(x; u_2, \alpha_2)]}{E_1(x; u_1, \alpha_1) - E_1(x; u_2, \alpha_2)} \quad (5)$$

The return period values ( $x_T$ ) may be obtained by:

$$x_T = \left(\frac{1}{2} + \hat{\theta}\right)(u_1 + \alpha_1 y_1) + \left(\frac{1}{2} - \hat{\theta}\right)(u_2 + \alpha_2 y_1) \quad (6)$$

$y_1 = -\ln \ln \left(\frac{T}{T-1}\right)$ , is the reduced variate.

### 2.3. Mixing of $EV_1$ with $EV_2$ or $EV_3$

If parameters of  $EV_1$  are  $(u_1, \alpha_1, y_1)$  and those of  $EV_2$  are  $(u_2, \alpha_2, y_2, k)$ , then the return period values ( $x_T$ ) of the series are given by :

$$x_T = \left(\frac{1}{2} + \theta\right) \left(u_1 + \alpha_1 y_1\right) + \left(\frac{1}{2} - \theta\right) \left(u_2 + \frac{\alpha_2}{k} - \frac{\alpha_2}{k} y_2\right) \quad (7)$$

where,  $y_2 = e^{-ky_1} \approx (1 - ky_1)$  (for small values of  $k$ )

$$\therefore x_T = \left(\frac{1}{2} + \theta\right)(u_1 + \alpha_1 y_1) + \left(\frac{1}{2} - \theta\right)(u_2 + \alpha_2 y_1) = u^* + \alpha^* y_1^* \quad (8)$$

where,  $u^* = \bar{u} + \theta u'$ ,

$$\alpha^* = \bar{\alpha} + \theta \alpha'$$

$$\bar{u} = \frac{u_1 + u_2}{2}, \quad \bar{\alpha} = \frac{\alpha_1 + \alpha_2}{2},$$

$$u' = u_1 - u_2, \quad \alpha' = \alpha_1 - \alpha_2.$$

Thus, the mixture corresponds to an  $EV_1$  distribution with parameters  $u^*$  and  $\alpha^*$ .

Variance of  $x_T$

$$v(x_T) = \frac{\bar{v}}{2} + 2\theta^2 \bar{v} + \theta v', \quad (9)$$

$$\text{where, } \bar{v} = \frac{1}{2} [v(x_1) + v(x_2)] \text{ \& } v' = v(x_1) - v(x_2)$$

and  $v(x_1)$  &  $v(x_2)$  are the variances when estimated by  $EV_1$  and  $EV_2$  distributions respectively.

### 3. Illustration

The mixture of two distributions is illustrated by two practical case studies in succeeding paras. These cases represent typical 24-hourly annual maximum rainfall series, where a single  $EV_1$  distribution does not provide a reasonable fit to the observed sample.

#### 3.1. Case study I

Daily (24-hr) annual maximum series (mm) for 70 years (1901-1970) was prepared for Bhuj (Gujarat). Maximum likelihood estimates of  $EV_1$  and  $EV_2$  parameters for this data set are :

$$u_1 = 60.4, \alpha_1 = 42.1 ;$$

$$u_2 = 55.3, \alpha_2 = 30.6 \text{ and } k = -0.34.$$

#### 3.2. Fitting of linear mixture of $EV_1$ and $EV_2$ distributions

The distribution function of the observed sample is  $E_n(x) = m/n$ , where  $m$  is number of observations  $\leq x$ .

The distribution functions of  $EV_1$  and  $EV_2$  are :

$$E_1(x) = \exp \left[ \exp \left( \frac{-x - 60.4}{42.1} \right) \right]$$

and

$$E_2(x) = \exp \left[ - \left( 1 + \frac{x - 55.3}{90} \right) \right]^{-2.94}$$

It is possible that the value of  $\theta$  for some observed values of  $x$  may fall outside  $(-0.5, 0.5)$  interval. In such cases we take  $\theta=0.5$  or  $-0.5$  whichever is closer to computed values. Evaluating  $\theta$  from Eqn. (3) for each observed value of  $x$  and taking their arithmetic mean, we get :

$$\theta = -0.1$$

The mixed distribution function,  $E(x)$  is taken as :

$$E(x) = 0.4 E_1(x) + 0.6 E_2(x)$$

$$\text{and } x_T = -8.7 + 16.8 y_1 + 54 e^{-0.31 y_1},$$

$$v(x_T) = 0.16 v(x_1) + 0.36 v(x_2)$$

The return period values ( $x_T$ ) and SE ( $x_T$ ) for the mixed distribution are provided in the Table 1.

All observations fall under 2 SE limits and the mixture distribution  $E(x) = .4 E_1(x) + .6 E_2(x)$  fits better to the observed series than  $EV_1$  alone. A comparison of  $E(x)$  and  $EV_1$  in respect of 3 highest recorded values is given in Table 2.

The above shows marked improvement in the recurrence interval value when we are using mixed distribution though compared to actuals, the values are still very much higher.

TABLE 1

Return period $T$ (yr)	EV <sub>1</sub> $x_1$ (mm)	EV <sub>2</sub> $x_2$ (mm)	Mixed distribution			
			$x_T$ (a) (mm)	SE( $x_T$ ) (b) (mm)	$x_T \pm 2SE(x_T)$	
					(a)-2(b) (mm)	(a)+2(b) (mm)
2	76	67	71	3.9	63	78
5	123	115	118	7.6	104	133
10	155	159	157	18.7	120	194
50	224	304	272	42.9	186	358
100	254	395	339	68.2	202	475
200	283	509	419	104.8	202	628
300	300	590	474	133.8	211	742
500	321	709	554	179.4	195	913
1000	351	908	685	264.3	157	1214
2000	380	1156	847	381.0	85	1509

TABLE 2

$x$ (mm)	Recurrence interval (years)		
	EV <sub>1</sub>	Mixed distribution	Plotting Position
467	exceeds 10,000	256	126
352	714	82	46
308	357	78	28

## 3.3. Case study 2

In this case the highest values exhibit very small increase in the ordered series from 58 years' daily rainfall (mm) recorded at Dinhat; the last few values are 269, 270, 275, 278, 281 and 290 which show hardly 2 cm range.

The equation for  $x_T$  is:

$$x_1 = 152.5 + 46.2 y_1$$

and  $SE(x_1) = 36.7(1.11 + 0.52y_1 + 0.61y_1^2)$ .

For EV<sub>3</sub> ( $u_3$ ,  $a_3$ ,  $k$ ) distribution the maximum likelihood estimates for parameters are:

$$u_3 = 157.3, a_3 = 48.3 \text{ \& } k = 0.17$$

$$\therefore x_3 = 448.3 - 2.91 \exp(-0.17 y_1)$$

In fitting a linear mixture of EV<sub>1</sub> and EV<sub>3</sub>,  $\theta$  has been evaluated for each observed value of  $x$  truncated to the nearest limiting value whenever it fell outside the limits  $(-1/2, 1/2)$  and then by simple arithmetic mean:

TABLE 3

$T$	$x_1$ (mm)	$x_3$ (mm)	$x_T$ (a) (mm)	SE( $x_T$ )	$x_T \pm 2SE(x_T)$	
				(b) (mm)	(a)-2(b) (mm)	(a)+2(b) (mm)
2	170	175	173	5.6	162	184
5	222	221	221	7.2	207	236
10	256	248	250	8.7	233	268
50	333	296	307	14.2	278	336
100	365	313	328	17.2	294	363
200	397	327	348	20.5	307	391
300	416	335	359	22.6	316	404
500	439	344	373	25.2	322	424
1000	472	356	391	29.0	333	441
2000	504	366	407	32.6	342	472
5000	546	378	428	37.4	353	503
10000	578	385	443	40.8	361	525

$$\theta = -0.2$$

Mixed distribution function is:

$$E(x) = 0.3 E_1(x) + 0.7 E_3(x)$$

$$x_T = 354.7 + 13.9 y_1 - 198.9 e^{-0.17 y_1}$$

and  $v(x_T) = 0.09 v(x_1) + 0.49 v(x_3)$ .

The results are provided in Table 3.

The mixed distribution  $E(x) = 0.3E_1(x) + 0.7E_3(x)$  provides a better fit to the observed series than  $E_1(x)$  but in the present case,  $E_3(x)$  appears to be equally or more efficient than EV<sub>1</sub>.

## 4. Summary

In a number of cases Gumbels EV<sub>1</sub> fails to describe the observed extreme series with acceptable accuracy. Two of the types are:

- (i) When the series has one or more outliers, whose return period as computed from EV<sub>1</sub> is several times larger than the plotting position ( $T_0$ ) values of the outliers,
- (ii) When the end values (arranged in ascending order) of the series show very slow increase so that their return periods are much less than  $T_0$ .

(2) In case (i) a linear mixture of EV<sub>1</sub> and EV<sub>2</sub>, i.e.,  $E(x) = (\frac{1}{2} + \theta) E_1(x) + (\frac{1}{2} - \theta) E_2(x)$  seems to provide a better fit to the data as illustrated by Bhuj series. This series has 3 outliers but all these have been brought below  $x_T + 2SE(x_T)$  limit using the mixed distribution:

$$E(x) = 0.4 E_1(x) + 0.6 E_2(x)$$

(3) In case (ii) a linear combination of  $EV_1$  and/or  $EV_3$  is suggested. It will enhance the recurrence interval ( $T$ ) for the end values of the series. This is illustrated in sample study 2 using Dinhat data. The highest value of observed series (290 mm) is less  $x_1 - 2SE(x_1)$  when  $EV_1$  alone is used for frequency analysis. But it comes under 2SE to describe the observations.

(4) From the above it may be inferred that the linear combination of  $EV_1$  with  $EV_2$  and/or  $EV_3$  [case (i) and (ii) respectively] would improve generally the results of frequency analysis of extremes as compared those obtained by using  $EV_1$  alone.

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