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## FITTING STATISTICAL DISTRIBUTIONS FOR MAXIMUM DAILY RAINFALL AT GVKK STATION

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**Abstract:** - Agro-climatic characters play an important role in deciding the cropping pattern of a region. The distribution of rainfall is one such climatic character essential to plan farm activities in a given region. The present study was conducted to know the climatic characterization of GVKK station. The secondary data of weather parameters over a period of 38 years (1976-2013) was collected from AICRP on Agro Meteorology. Among the weather parameters, amount of maximum daily rainfall (mm) was considered to fit appropriate probability distributions. The probability distributions viz., Normal, Log-normal, Gamma (1P, 2P, 3P), Generalized Extreme Value (GEV), Weibull (1P, 2P, 3P), Gumbel and Pareto were used to evaluate the best fit for maximum daily rainfall (mm). Kolmogorov-Smirnov test for the goodness of fit of the probability distributions showed that for majority of the data sets on rainfall at different study periods, Weibull (3P) distribution was found to be the best fit. However, all the data sets were scale dominated which indicated large variation in the distribution of rainfall.

**Keywords:** Maximum daily rainfall, Probability distributions and Goodness-of-fit test.

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## INTRODUCTION

Climate is a measure of average pattern of variation in [temperature](#), [humidity](#), [atmospheric pressure](#), [wind](#) speed, [precipitation](#), atmospheric particle count and other [meteorological](#) variables in a given region over long periods of time. Climate is different from [weather](#), in that weather only describes the short-term conditions of these variables in a given region. Climate is usually defined as the "average weather" or as the statistical description in terms of the mean and variability of relevant quantities over a period ranging from months to thousands or millions of years. The classical period is 30 years, as defined by the [World Meteorological Organization](#) (WMO). These quantities are most often surface variables such as temperature, precipitation, and wind. Climate in a wider sense is the state, including a statistical description, of the climate system. Rainfall is one of the most important natural input resources to crop production. Its occurrence and distribution is erratic. Analysis of rainfall data strongly depends on its distribution pattern. It has long been a topic of interest in the fields of Agricultural Statistics in establishing a probability distribution that provides a good fit to daily rainfall data. Several studies have been conducted in India and abroad on rainfall analysis and best fit probability distribution functions such as normal, lognormal, Weibull and gamma type distributions were identified.

The annual and seasonal analysis of rainfall will give general idea about the rainfall pattern of the region, whereas the monthly as well as weekly analysis of rainfall will be of much use as far as agricultural planning is concerned. Knowledge of the distribution of rainfall is essential for successful farming.

Gregory *et al.* (2007) demonstrated the feasibility of fitting cell-by-cell probability distributions to grids of monthly interpolated, Continent-wise data and showed that the gamma distribution was well suited. Suhaila and Jemain (2007) showed that Mixed Exponential was found to be the most appropriate distribution for describing the daily amount of rainfall in Peninsular Malaysia. Lars and Vogel (2008) studied the distribution of wet-day daily rainfall and identified that the 2-parameter Gamma (2P) distribution as the most likely candidate distribution based on traditional goodness of fit tests. Deka *et al.* (2009) compared the goodness of fit test results and generalized logistic distribution was empirically proved to be the most appropriate distribution for describing the annual maximum rainfall series for the majority of the stations in North East India. Sharma and Singh (2010) analyzed the maximum daily rainfall data of Pantnagar for a period of 37 years for annually, seasonally, monthly and weekly, and the best fitted probability distribution was identified using goodness of fit tests. Manikandan *et al.* (2011) analyzed daily rainfall data of 37 years at Tamil Nadu Agricultural University (TNAU)

Campus. Chi-square values revealed that the log-normal distribution was the best fit probability distribution for annual one day maximum rainfall. Bhim Singh *et al.* (2012) analyzed the daily rainfall data of 39 years (1973-2011) in Jhalrapatan area of Rajasthan and log-Pearson type-III distribution was found to be the best fit probability distribution. Oseni and Femi (2012) have fitted several types of statistical distributions to describe rainfall distribution in Ibadan metropolis over a period of 30 years. Mayooran and Laheetharan (2014) have identified best fit probability distributions among 45 standard probability distributions to model annual maximum rainfall in Colombo district for a period of 110 years. The present study was planned for establishing the methodology for identifying the best fit probability distribution for maximum daily rainfall.

## MATERIAL AND METHODS

The present study was conducted to know the pattern of rainfall distribution at Gandhi Krishi Vignana Kendra (GKVK) station. The station, GKVK is located at Bengaluru Urban District of Bengaluru North Taluk. It belongs to the Eastern Dry zone (Zone-V). The geographical coordinates of this station are  $77^{\circ}35'$  longitude,  $12^{\circ}58'$  latitude and 930 amsl altitude. The Eastern dry zone includes Kolar, Tumakuru, Bengaluru (Urban), Bengaluru (Rural), Chikkaballapur and Ramanagara. The zone has an average annual rainfall of 944mm (1976-2013). Major crops grown in this zone are ragi, groundnut, rice and maize.

The present study was based on the secondary data on rainfall over a period of 38 years (1976-2013) which was collected from AICRP on Agro Meteorology, University of Agricultural Sciences, GKVK, Bengaluru.

### Fitting probability distributions

The standard probability distributions viz. normal, log-normal, gamma (1P, 2P and 3P), Weibull (1P, 2P and 3P), Gumbel, Pareto and generalised extreme value (GEV) were identified to evaluate the best fit probability distribution for maximum daily rainfall. The description of various probability density functions, range of the variable and the parameters involved are presented in Table 1.

### Description of parameters:

#### ➤ Shape parameter

Shape parameters allow a distribution to take on a variety of shapes, depending on the value of the shape parameter. Examples of shape parameters are skewness and kurtosis.

**Table 1: Description of various probability density functions with range of variable and parameters.**

Distribution	Probability density function	Range	Parameters
<b>Gamma (1P)</b>	$f(x) = \frac{1}{\Gamma(k)} x^{k-1} \exp(-x)$	$0 \leq x < +\infty$	$k =$ Shape parameter
<b>Gamma (2P)</b>	$f(x) = (x)^{k-1} \exp\left[\frac{-(k-\mu)/\beta}{\beta^k \tau(k)}\right]$	$k > 0$ and $\beta > 0$	$\beta =$ Scale parameter
<b>Gamma (3P)</b>	$f(x) = (x - \mu)^{k-1} \exp\left[\frac{-(k-\mu)/\beta}{\beta^k \Gamma(k)}\right]$		$\mu =$ Location parameter
<b>GEV</b>	$f(x) = \begin{cases} \frac{1}{\beta} \exp\left[-(1+kz)^{-\frac{1}{k}}\right] (1+kz)^{-1-1/k} & k \neq 0 \\ \frac{1}{\beta} \exp[-z - \exp(-z)] & k = 0 \end{cases}$	$1+kz > 0$ for $k \neq 0$ $-\infty < x < +\infty$ for $k=0$ where $z = \frac{(x-\mu)}{\beta}$	$k =$ Shape parameter $\beta =$ Scale parameter $\mu =$ Location parameter
<b>Normal</b>	$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{1}{2} \frac{(x-\mu)^2}{2\sigma^2}\right]$	$-\infty < x < +\infty$ $-\infty < \mu < +\infty$ $\sigma > 0$	$\mu =$ Mean $\sigma =$ Standard deviation
<b>Log- normal</b>	$f(x) = \frac{1}{x\sigma\sqrt{2\pi}} \exp\left[\frac{(\ln(x)-\mu)^2}{2\sigma^2}\right]$	$0 < x < +\infty$	$\sigma =$ Scale parameter $\mu =$ Location parameter
<b>Gumbel</b>	$f(x) = \frac{1}{\beta} \exp-(z + e^{-z})$ Where, $z = \frac{x-\mu}{\beta}$	$\beta > 0$ $-\infty < x < +\infty$	$\beta =$ Scale parameter $\mu =$ Location parameter
<b>Pareto</b>	$f(x) = \frac{k\beta^k}{\beta^{k+1}}$	$1 \leq x \leq +\infty$ $k, \beta > 0$ $x \geq 1$	$k =$ Shape parameter $\beta =$ Scale parameter
<b>Weibull (1P)</b>	$f(x) = \beta x^{\beta-1} \exp(-x^\beta)$	$x > 0$ $\beta > 0$	$\beta =$ Shape parameter

<p><b>Weibull (2P)</b></p>	$f(x) = \frac{k}{\beta} \left(\frac{x}{\beta}\right)^{k-1} \exp\left(-\left(\frac{x}{\beta}\right)^k\right)$	$0 \leq x < +\infty$ $k, \beta, > 0$	<p>k = Shape parameter</p> <p><math>\beta</math> = Scale parameter</p>
<p><b>Weibull (3P)</b></p>	$f(x) = \frac{k}{\beta} \left(\frac{x-\mu}{\beta}\right)^{k-1} \exp\left(-\left(\frac{x-\mu}{\beta}\right)^k\right)$		<p><math>\mu</math> = Location parameter</p> <p>(<math>\mu \equiv 0</math> yields the two parameter Weibull distribution)</p>

➤ **Scale parameter**

The scale parameter of a distribution determines the scale of the distribution function. The larger the scale parameter, the more spread out the distribution. The examples of scale parameters include variance and standard deviation.

➤ **Location parameter**

The location parameter determines the position of central tendency of the distribution along the x-axis.. The location parameter defines the shift of the data. Examples of location parameters include the [mean](#), [median](#), and the [mode](#).

**Kolmogorov- Smirnov test (K-S test) for testing the goodness of fit**

The goodness of fit test measures the discrepancy between observed values and the expected values. Kolmogorov- Smirnov test was used to test for the goodness of fit.

In the present investigation, the goodness of fit test was conducted at  $\alpha = 0.05$  level of significance. It was applied for testing the following hypothesis:

H<sub>0</sub>: The maximum daily rainfall data follows a specified distribution.

H<sub>1</sub>: The maximum daily rainfall data does not follow a specified distribution.

This test is used to decide whether a sample comes from a hypothesized continuous probability density function. It is based on the empirical distribution function i.e., on the largest vertical difference between the theoretical and empirical cumulative distribution function.

$$D = \max_{1 \leq i \leq n} \left( F(X_i) - \frac{i-1}{n}, \frac{i}{n} - F(X_i) \right)$$

Where,  $X_i$  = Random sample,  $i = 1, 2, \dots, n$ .

$$CDF = F_n(X) = \frac{1}{n} [\text{Number of observations} \leq x]$$

### Identification of best fit probability distribution.

The KS test was used to identify the best fit probability distribution for maximum daily rainfall for different data sets. The test statistic of the test was computed and tested at ( $\alpha = 0.05$ ) level of significance. Among all the fitted probability distributions for maximum daily rainfall, the distribution with lowest test statistic value and highest p-value was considered as the best fit.

## RESULTS AND DISCUSSION

The methodology presented above was applied to the 38 years weather data in which maximum rainfall in millimetres were collected from AICRP on agro meteorology located at UAS, GKVK, Bengaluru. Accordingly, the data was classified into 28 data sets as mentioned in Table-2. These 28 data sets were classified as 1 annual, 1 seasonal, 5 seasonal months and 21 standard meteorological weeks to study the distribution pattern at different levels.

### Descriptive statistics of maximum daily rainfall

The summary of statistics such as highest, lowest, mean, standard deviation, skewness and coefficient of variation values of maximum daily rainfall are presented in Table 2. The results show that both annual and seasonal maximum daily rainfall was observed to be 200 mm. Monthly maximum daily rainfall during monsoon season ranged from 98.6 mm to 200 mm while weekly maximum daily rainfall was between 43.2 mm to 200 mm. It was also observed that the minimum daily rainfall was 56.8 mm annually, 43.6 mm seasonally and it ranged from 6.4 mm to 16 mm monthly. For all the weeks, the minimum rainfall was found to be 0.0 mm.

Annual mean of the maximum daily rainfall was found to be 94.6 mm, whereas for overall south-west monsoon seasonal months, it was 90.6 mm. During the seasonal months, the mean of the maximum daily rainfall ranged from 32.1mm to 67.6 mm while for weekly, it varied from 5.2 mm to 33.5 mm.

The value of coefficient of variation annually was observed to be 39.2 per cent which was the lowest among all the data sets. Seasonally, the rainfall varied by about 43.6 per cent whereas for seasonal months, the variation ranged from 58.6 to 66.8 percent. For Standard

**Table 2: Summary statistics for maximum daily rainfall (mm)**

Study Period	Range	Parameters					
		Mean	SD	CV (%)	Skewness	Maximum	Minimum
Annual	1 <sup>st</sup> Jan–31 <sup>st</sup> Dec	94.6	37.1	39.2	1.0	200.0	56.8
Seasonal	1 <sup>st</sup> June- 28 <sup>th</sup> Oct	90.6	39.4	43.6	1.0	200.0	43.6
June	1 <sup>st</sup> June-30 <sup>th</sup> June	32.1	20.9	65.0	1.1	98.6	6.4
July	1 <sup>st</sup> July-31 <sup>st</sup> July	40.5	27.0	66.8	1.8	139.5	10.2
August	1 <sup>st</sup> Aug-31 <sup>st</sup> Aug	43.1	25.9	60.1	0.6	98.8	10.2
September	1 <sup>st</sup> Sept-30 <sup>th</sup> Sept	67.6	39.6	58.6	0.8	158.4	15.4
October	1 <sup>st</sup> Oct- 28 <sup>th</sup> Oct	66.8	43.3	64.8	1.4	200.0	16.0
23 <sup>rd</sup> SMW	4 <sup>th</sup> June-10 <sup>th</sup> June	18.0	18.5	102.8	0.7	58.8	0.0
24 <sup>th</sup> SMW	11 <sup>th</sup> June- 17 <sup>th</sup> June	9.3	11.6	124.6	1.7	49.4	0.0
25 <sup>th</sup> SMW	18 <sup>th</sup> June- 24 <sup>th</sup> June	9.3	14.3	153.3	2.3	56.2	0.0
26 <sup>th</sup> SMW	25 <sup>th</sup> June-1 <sup>st</sup> July	5.2	7.7	148.8	3.6	43.2	0.0
27 <sup>th</sup> SMW	2 <sup>nd</sup> July to 8 <sup>th</sup> July	16.3	21.7	133.7	2.4	98.6	0.0
28 <sup>th</sup> SMW	9 <sup>th</sup> July-15 <sup>th</sup> July	14.5	23.1	159.3	4.5	139.5	0.0
29 <sup>th</sup> SMW	16 <sup>th</sup> July-22 <sup>nd</sup> July	16.7	14.5	86.9	0.7	47.2	0.0
30 <sup>th</sup> SMW	23 <sup>rd</sup> July-29 <sup>th</sup> July	11.5	11.6	101.0	1.5	47.2	0.0
31 <sup>st</sup> SMW	30 <sup>th</sup> July-5 <sup>th</sup> Aug	18.5	20.8	112.4	2.3	97.8	0.0
32 <sup>nd</sup> SMW	6 <sup>th</sup> Aug-12 <sup>th</sup> Aug	14.2	14.6	103.0	1.5	62.2	0.0
33 <sup>rd</sup> SMW	13 <sup>th</sup> Aug-19 <sup>th</sup> Aug	17.2	18.6	108.5	1.9	79.6	0.6
34 <sup>th</sup> SMW	20 <sup>th</sup> Aug-26 <sup>th</sup> Aug	21.6	26.3	121.5	1.8	98.8	0.0
35 <sup>th</sup> SMW	27 <sup>th</sup> Aug-2 <sup>nd</sup> Sept	22.0	22.6	102.6	1.2	73.6	0.0
36 <sup>th</sup> SMW	3 <sup>rd</sup> Sept-9 <sup>th</sup> Sept	29.0	36.4	125.7	1.9	151.8	0.0
37 <sup>th</sup> SMW	10 <sup>th</sup> Sept- 16 <sup>th</sup> Sept	32.9	36.7	111.6	1.5	136.0	0.0
38 <sup>th</sup> SMW	17 <sup>th</sup> Sept- 23 <sup>rd</sup> Sept	23.4	23.6	100.8	1.1	89.5	0.0
39 <sup>th</sup> SMW	24 <sup>th</sup> Sept- 30 <sup>th</sup> Sept	33.5	39.2	116.8	1.6	158.4	0.0
40 <sup>th</sup> SMW	1 <sup>st</sup> Oct-7 <sup>th</sup> Oct	26.4	22.7	85.8	1.0	84.0	0.0
41 <sup>st</sup> SMW	8 <sup>th</sup> Oct-14 <sup>th</sup> Oct	24.8	34.8	140.3	3.6	200.0	0.0
42 <sup>nd</sup> SMW	15 <sup>th</sup> Oct-21 <sup>st</sup> Oct	18.8	19.4	103.1	1.0	64.4	0.0
43 <sup>rd</sup> SMW	22 <sup>th</sup> Oct-28 <sup>th</sup> Oct	18.1	36.6	201.9	3.1	157.6	0.0

Meteorological Weeks (SMW's), the rainfall variation was found to be very high ranging from 86.9 to 201.9 per cent.

The coefficient of skewness for all the data sets ranged from 0.6 to 4.5 indicating positive skewness. From this result, we can conclude that in all the data sets the average rainfall for the period exceeded the modal value.

Figure 1 shows the year wise variation in annual maximum daily rainfall from 1976-2013. It ranged from 56.8 mm (in 2006) to 200 mm (in 1997). The minimum of the maximum daily rainfall, i.e., 56.8 mm occurred during 10<sup>th</sup> SMW (5<sup>th</sup> Mar-11<sup>th</sup> Mar) and the maximum of the maximum daily rainfall of 200 mm occurred during the 41<sup>st</sup> SMW (8<sup>th</sup> Oct-14<sup>th</sup> Oct).

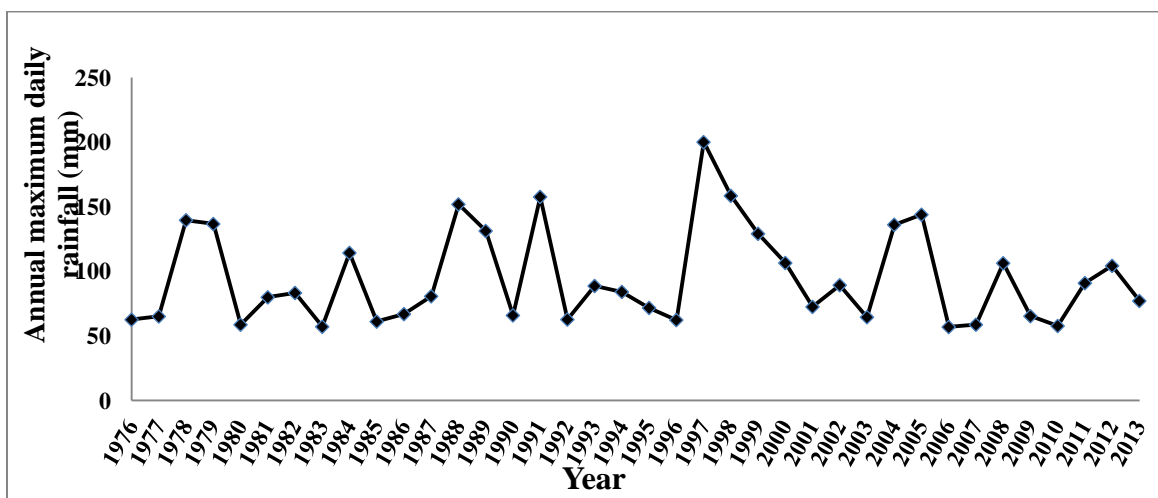


Fig 1: Variation in Maximum daily rainfall (mm) over a study period

### Fitting the probability distributions

Study period probability distribution using goodness of fit tests is given in the Table 3. It was observed from the table that Weibull (3P) distribution was the best fit probability distribution for majority of the Standard Meteorological Weeks, for the annual, seasonal and June month study periods. Log-normal distribution fitted well for July and October months while, Weibull (2P) was the best fit for August month. Gamma (2P) was observed to be the best fit for seasonal month of September along with the standard weeks such as 23<sup>rd</sup>, 32<sup>nd</sup>, 33<sup>rd</sup> and 34<sup>th</sup>. Weibull (3P) distribution fitted well for majority of the data sets since the distribution provides a flexible representation of a variety of shapes. Majority of the study periods were scale-dominated which indicated large variation in the distribution of rainfall.



The distribution parameters helped to determine the pattern of the rainfall distribution in the study region. The joint interpretation of shape and scale parameters conveys the distribution of values in the modeled rainfall data at each location allowing the interpreter a qualitative assessment of the amount and stability of rainfall throughout the season. These parameters reflect the modeled rainfall, and as such also contain errors inherent in the modeled history.

**Table 3: Study period wise probability distribution using goodness of fit tests.**

Study Period	Range	Best fit	Parameters		
			Shape parameter (k)	Scale parameter ( $\beta$ )	Location parameter ( $\mu$ )
Annual	1 <sup>st</sup> Jan–31 <sup>st</sup> Dec	Weibull (3P)	0.8125	37.1637	56.3503
Seasonal	1 <sup>st</sup> June- 28 <sup>th</sup> Oct	Weibull (3P)	2.1501	74.9242	12.0745
June	1 <sup>st</sup> June-30 <sup>th</sup> June	Weibull (3P)	1.2689	29.7402	4.8815
July	1 <sup>st</sup> July-31 <sup>st</sup> July	Log-normal	3.5139	0.6204	
August	1 <sup>st</sup> Aug-31 <sup>st</sup> Aug	Weibull (2P)	1.7800	48.7270	
September	1 <sup>st</sup> Sept-30 <sup>th</sup> Sept	Gamma (2P)	2.9149	23.1780	
October	1 <sup>st</sup> Oct- 28 <sup>th</sup> Oct	Log-normal	4.0168	0.6175	
23 <sup>rd</sup> SMW	4 <sup>th</sup> June-10 <sup>th</sup> June	Gamma (2P)	0.2105	19.0531	
24 <sup>th</sup> SMW	11 <sup>th</sup> June-17 <sup>th</sup> June	Weibull (3P)	0.7229	8.0669	- 0.1998
25 <sup>th</sup> SMW	18 <sup>th</sup> June-24 <sup>th</sup> June	Weibull (3P)	0.7201	7.3989	- 0.2997
26 <sup>th</sup> SMW	25 <sup>th</sup> June-1 <sup>st</sup> July	Weibull (3P)	0.8702	5.2704	- 0.4514
27 <sup>th</sup> SMW	2 <sup>nd</sup> July to 8 <sup>th</sup> July	Weibull (3P)	0.8197	14.9306	- 0.4496
28 <sup>th</sup> SMW	9 <sup>th</sup> July-15 <sup>th</sup> July	Weibull (3P)	0.8214	13.0598	- 0.4496
29 <sup>th</sup> SMW	16 <sup>th</sup> July-22 <sup>nd</sup> July	Weibull (3P)	1.0279	17.9583	- 0.4506
30 <sup>th</sup> SMW	23 <sup>rd</sup> July-29 <sup>th</sup> July	Weibull (3P)	1.1748	14.1673	- 1.5595
31 <sup>st</sup> SMW	30 <sup>th</sup> July-5 <sup>th</sup> Aug	Weibull (3P)	1.1494	19.7428	- 0.6940
32 <sup>nd</sup> SMW	6 <sup>th</sup> Aug-12 <sup>th</sup> Aug	Gamma (2P)	0.9427	15.0176	
33 <sup>rd</sup> SMW	13 <sup>th</sup> Aug-19 <sup>th</sup> Aug	Gamma (2P)	0.8501	20.1691	
34 <sup>th</sup> SMW	20 <sup>th</sup> Aug-26 <sup>th</sup> Aug	Gamma (2P)	0.6770	31.9156	
35 <sup>th</sup> SMW	27 <sup>th</sup> Aug-2 <sup>nd</sup> Sept	Weibull (3P)	0.9380	22.4655	- 0.6777
36 <sup>th</sup> SMW	3 <sup>rd</sup> Sept-9 <sup>th</sup> Sept	Weibull (3P)	0.6683	24.6470	- 0.4496
37 <sup>th</sup> SMW	10 <sup>th</sup> Sept-16 <sup>th</sup> Sept	Weibull (3P)	0.8899	31.4378	- 0.4496
38 <sup>th</sup> SMW	17 <sup>th</sup> Sept-23 <sup>rd</sup> Sept	Weibull (3P)	0.8559	24.0877	- 1.0237
39 <sup>th</sup> SMW	24 <sup>th</sup> Sept-30 <sup>th</sup> Sept	Weibull (3P)	0.6439	24.7259	- 0.1998
40 <sup>th</sup> SMW	1 <sup>st</sup> Oct-7 <sup>th</sup> Oct	Weibull (3P)	1.4546	34.9827	- 5.1230
41 <sup>st</sup> SMW	8 <sup>th</sup> Oct-14 <sup>th</sup> Oct	Weibull (3P)	0.8737	24.8221	- 1.5266
42 <sup>nd</sup> SMW	15 <sup>th</sup> Oct-21 <sup>st</sup> Oct	Weibull (3P)	0.8635	19.5985	- 1.0179
43 <sup>rd</sup> SMW	22 <sup>th</sup> Oct-28 <sup>th</sup> Oct	Weibull (3P)	0.5184	9.9949	- 0.1998

## CONCLUSION

The present study revealed the distribution pattern of Maximum daily rainfall at GVKK station using appropriate best fit probability distributions. Kolmogorov-Smirnov test statistic and p-values of the probability distributions showed that for majority of the data sets of rainfall at different study periods, Weibull (3P) distribution was found to be the best fit among the 11 fitted distributions. However, majority of the data sets were scale dominated which indicated large variation in the distribution of rainfall.

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