

# **The Asymmetric Distribution of Rainfall Frequency and Amounts in India**

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## ABSTRACT

25 Studies of rainfall are usually based on the total amount precipitating throughout a certain  
26 period. Compared to rain rates associated with extreme events, the rain rates associated with  
27 periods when most of the rainfall occur are not studied extensively. In this study, the  
28 characteristics of daily precipitation in India are explored using two metrics - rain frequency  
29 peak (the most frequent non-zero rain rate) and rain amount peak (the rain rate at which the  
30 most amount of rain falls). These metrics are computed over India using local and global  
31 datasets to investigate the characteristics of typical daily precipitation accumulations. These  
32 values are sensitive to the dataset used for this analysis since it is expected that the temporal  
33 and spatial resolution of the rainfall data will influence the rain frequency peak and rain  
34 amount peak. The maximum rain frequency peak for any region in India reaches up to a value  
35 of 35 mm/day while the maximum rain amount peak reaches up to a value as high as 90  
36 mm/day. Findings of the study indicate that the two metrics defined are sufficient to examine  
37 various measures of precipitation characteristics and unravel regional differences in a simple  
38 manner.

39 Keywords: Precipitation characteristics, rain frequency peak, rain amount peak

## 40 1. Introduction

41 Extreme precipitation is a widely studied topic as they frequently lead to widespread flooding  
42 and damage to life and property, while events that occur most frequently and contribute to the  
43 most amount of precipitation are under studied. The typical precipitation over a particular  
44 area has generally been quantified at a climatological scale, usually averaged over a month,  
45 season, or longer. While this approach is useful to distinguish wet and dry areas, it fails to  
46 inform us of the nature and quantity of typical daily precipitation over a location, which  
47 arguably influences majority of our decisions with regards to rain compared to extreme  
48 events that are infrequent and atypical. While total daily precipitation is a frequently  
49 communicated metric, it is insufficient to define a typical rainy day's precipitation since  
50 rainfall varies with each month, season, and year.

51 The total frequency and total intensity of precipitation are two commonly used metrics that  
52 represent how often it rains and how heavy a rain event is. Englehart and Douglas (1985)  
53 found that precipitation frequency, given by the number of days per month or season  
54 receiving greater than a specified amount of rainfall, is more normally distributed and more

55 spatially coherent than total precipitation. A few other studies have investigated total  
56 frequency and intensity in rainfall observations (Biasutti and Yuter, 2013; Chen et al., 1996;  
57 Dai et al., 2007; Sun et al., 2006). Gehne (2016) and Herold (2016) both revealed  
58 considerable uncertainty in observational products.

59 As extreme rainfall events lead to natural disasters such as floods, much of the literature is  
60 focused on them (e.g., Ricko et al., 2016). But these extreme events are infrequent and  
61 inadequate in determining the typical daily precipitation. Finding appropriate ways to  
62 distinguish light and heavy rain events by looking at their statistical distributions is a topic of  
63 active research. . Some of the earliest work focused on the rain frequency distribution, first  
64 qualitatively (Dai, 2001; Petty, 1995) and then utilizing categorical bins (e.g., Dai, 2006).  
65 Later, Watterson and Dix (2003) computed the amount of rain falling in each categorical bin,  
66 referred to as the rain amount distribution or the rain volume distribution in certain studies  
67 (We use the term rain amount distribution throughout this study). Note that the sum of the  
68 rain amount distribution for all the bins gives us the total precipitation. Sun (2007) utilised  
69 bins that were linearly spaced in rain rate to quantify the rain distribution. Though it provides  
70 a mathematical basis for the analysing the distribution of rain, it has imperfect sampling  
71 properties since the daily accumulation of rain (rain rate) spans orders of magnitude.  
72 Similarly, Pendergrass and Hartmann (2014) used logarithmically spaced rain rate bins to  
73 compute the rain amount and the rain frequency distributions. All rain distributions such as  
74 rain frequency distribution, rain amount distribution, etc. depend significantly on the  
75 selection of bin structure for rain rates.

76 Similar investigations into statistical properties of precipitation have also been done over  
77 India. In one of the earliest studies, Ananthakrishnan (1970) studied the space-time variations  
78 of the Indian rainfall by examining the pentad (5 day totals) normal rainfall curves of the  
79 representative stations of different regions of the country. Further, Ananthakrishnan and  
80 Soman (1989) performed a statistical analysis of the daily rainfall series of 15 Indian stations  
81 to develop normalized rainfall curves (NRC) that associated cumulated percentage rain  
82 amount and the cumulated percentage number of rain days. Goswami et al. (2006)  
83 demonstrated that despite considerable year-to-year variability, there are significant increases  
84 in the frequency and the intensity of extreme monsoon rain events over central India during  
85 the period of 1951 to 2000. While studies focused on extreme rainfall events are plenty in  
86 India, work focused on the typical precipitation events in India is sparse.

87 In order to focus on typical events, we need metrics derived from the distribution that  
88 quantifies the characteristics of typical daily precipitation. In one such study, Pendergrass and  
89 Deser (2017) has proposed a few precipitation metrics to quantify the characteristics of daily  
90 rainfall by analysing the distributions of rain computed using the observational precipitation  
91 datasets. In this study, we adopt metrics proposed in Pendergrass and Deser (2017) which  
92 focused on the global climatological characteristics of typical daily precipitation. Here, we  
93 not only present quantitative values to answer the question asked at the very beginning "How  
94 much rain falls on a typical rainy day?" but also analyse the seasonal and spatial variation of  
95 the precipitation metrics(defined below) for India. Further, this study extends to make a  
96 comparison between these metrics and the rain distributions computed for more than one  
97 dataset.

98 The rest of this paper is arranged as follows. Section 2 discusses the methodology adopted  
99 and provides information on the observational datasets used for analysis. There is a detailed  
100 case study on the four metropolitan cities of India in Section 3 that illustrates the significance  
101 of the defined precipitation metrics in studying the typical daily precipitation characteristics.  
102 Section 3 presents the quantification of daily precipitation characteristics of India. In Section  
103 4, the rain frequency distribution and the rain amount distribution computed for two different  
104 datasets, i.e. IMD and CHIRPS are compared and subjected to detailed analysis. Section 5  
105 shows the climatological seasonal variation of rain distributions and the precipitation metrics  
106 along latitude and longitude studied in detail. Section 6 contains the precipitation metrics map  
107 for India that provides crucial insights into how the defined precipitation metrics vary in  
108 different regions of India. Section 7 concludes the report by summarising and providing other  
109 necessary information relevant to this study.

## 110 **2. Datasets & Methodology**

### 111 ***2.1 Observational Datasets:***

112 Two gridded precipitation datasets, one national and another global, are utilized in this study.  
113 The first is the high-resolution ( $0.25^\circ \times 0.25^\circ$ ) daily gridded rainfall dataset prepared by the  
114 Indian Meteorological Department (IMD) covering mainland India (Pai et al., 2014). This  
115 long dataset covering the period of 1901-2018 has been generated using rainfall  
116 measurements from more than 7000 gauging stations, and captures the spatial variability of  
117 rainfall across the country better than other gridded datasets due to the rapidly increasing  
118 spatial density of gauging stations. The station data has been converted to gridded data by

119 spatially interpolating using the Inverse Distance Weighted scheme (Shepard, 1968). The  
120 Climate Hazards Group InfraRed Precipitation with Station (CHIRPS) dataset is a relatively-  
121 new quasi-global rainfall product developed by the U.S. Geological Survey (USGS)/Climate  
122 Hazards Group science team. The latest CHIRPS Version 2.0 dataset  
123 (<http://chg.geog.ucsb.edu/data/chirps/>) with a spatial resolution of  $0.05^\circ \times 0.05^\circ$  was used in  
124 our study which incorporates  $0.05^\circ$  resolution satellite imagery with in-situ station data to  
125 create gridded rainfall time series (1981-2020). A common period of 1981-2018 between the  
126 two datasets has been selected to study and compare the typical characteristics of daily  
127 precipitation.

## 128 *2.2 Precipitation Metrics:*

129 In order to relate the characteristics of daily precipitation to their contribution the mean and  
130 variability of rainfall throughout the country, we take recourse to metrics such as rain  
131 frequency peak and rain amount peak.

132 a) **Rain Frequency Peak:** This metric gives us the non-zero rain rate with the maximum  
133 frequency. We have identified the rain rate corresponding to the rain frequency peak  
134 by computing the frequency distribution of daily rainfall wherein frequency is  
135 presented in terms of the percentage of the total number of days. First, we create bins  
136 from 0.1 mm/day to 100 mm/day distributed logarithmically. Note that a day  
137 receiving rain rate less than 0.1 mm/day is counted as a dry day while rain rate greater  
138 than 100 mm/day is counted as an extremely wet day. Though we have calculated the  
139 percentage of dry days in this metric, we haven't calculated the frequency of  
140 extremely wet days regarding them to be extreme events and not crucial for our daily  
141 precipitation analysis. We will, however, present the calculations related to the  
142 extremely wet days in the next metric.

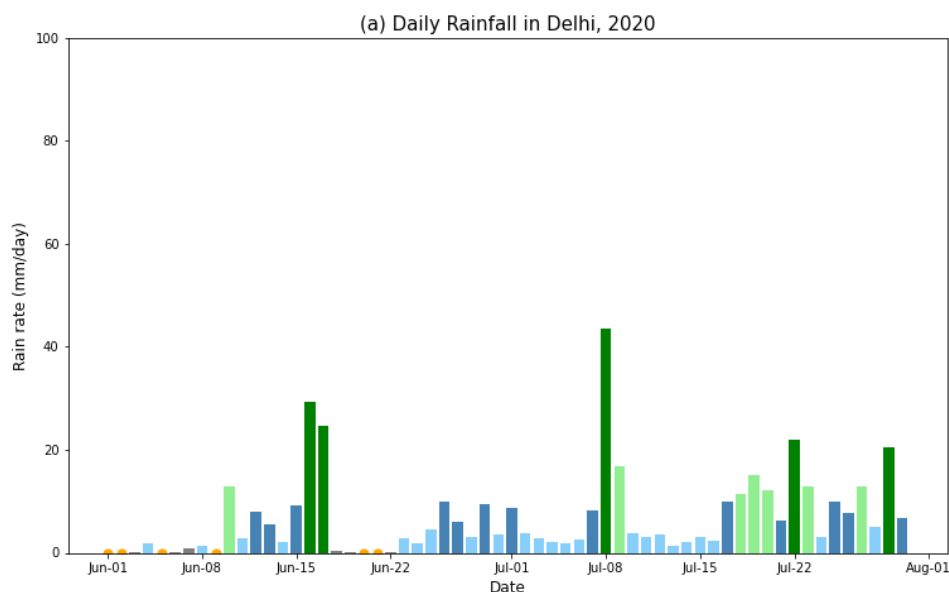
143 b) **Rain Amount Peak:** This metric gives us the rain rate at which the most rain falls.  
144 We have identified the rain rate corresponding to the rain amount peak by computing  
145 the distribution of the amount of precipitation contributed by any individual rain rate  
146 towards total precipitation. The amount is presented in terms of the percentage.  
147 Similar to the rain frequency peak metric, we created bins from 0.1 mm/day to 100  
148 mm/day distributed logarithmically. Note that the rain rate less than 0.1mm/day is  
149 counted as dry days while the rain rate greater than 100mm/day is counted as  
150 extremely wet days. While we have calculated the percentage of precipitation

151 contributed by extremely wet days in this metric, we haven't computed amount of  
 152 precipitation for dry days since their contribution towards total precipitation is  
 153 minimal compared to the contribution of other rain rates.

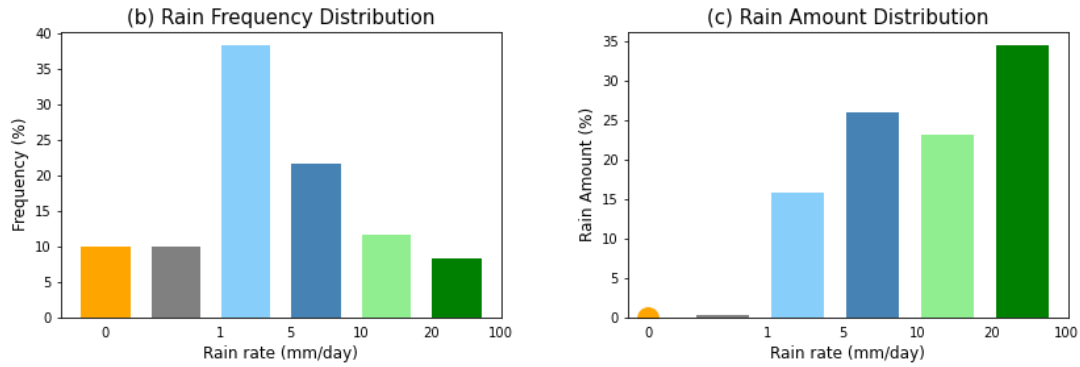
### 154 3. Illustration of Precipitation Characteristics over Metropolitan Cities

155 To illustrate how rainfall frequency and amount distribution describe precipitation  
 156 characteristics, let's examine the rain frequency and rain amount distribution for the four  
 157 metropolitan cities of India, i.e. Delhi, Mumbai, Kolkata, and Chennai (Fig. 1-4).

158 Let's consider the time series of daily precipitation in Delhi for the months of June and July  
 159 of 2020 obtained from the IMD daily rainfall dataset. Note that throughout this report, daily  
 160 precipitation accumulation is referred to by either "rain rate" following (Pendergrass &  
 161 Hartmann, 2014) or "intensity" following Stone et al. (2000). The time series of daily rainfall  
 162 in Delhi is shown in Fig. 1(a). Days with no precipitation are indicated with orange circles on  
 163 the abscissa, and this convention is followed in all other plots to display zero values. Days  
 164 with non-zero precipitation are marked with vertical bars wherein bar length indicates rain  
 165 rate and all the bars are colour coded according to discrete rain-rate bins. This will help to  
 166 visualise the rain frequency distribution (Fig. 1b) and rain amount distribution (Fig. 1c), both  
 167 of which are also colour coded according to the rain rate bins. For Delhi, monsoon starts in  
 168 late June and thus, over these two months, the total precipitation is 405 mm, which is  
 169 equivalent to an average rain rate of 6.75 mm/day.



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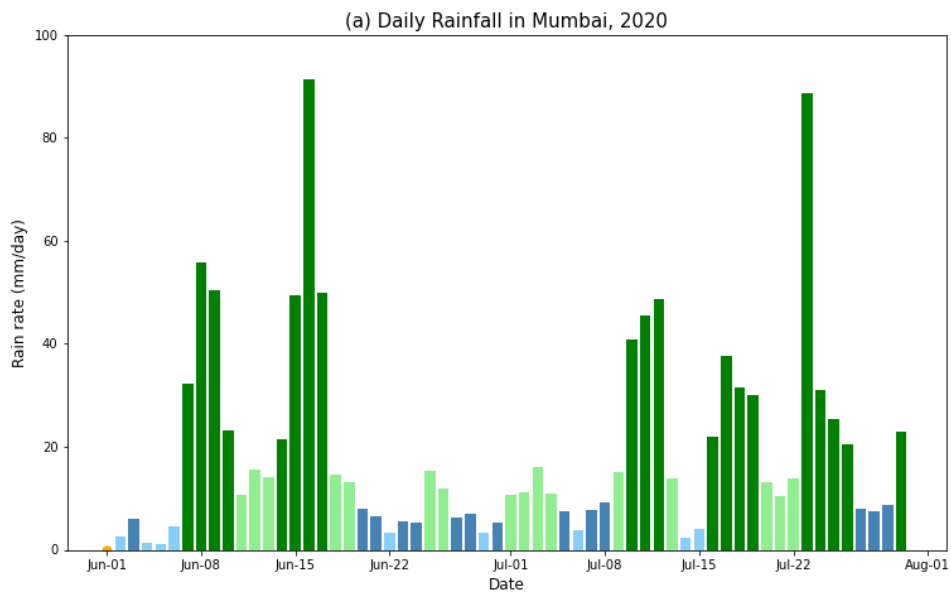


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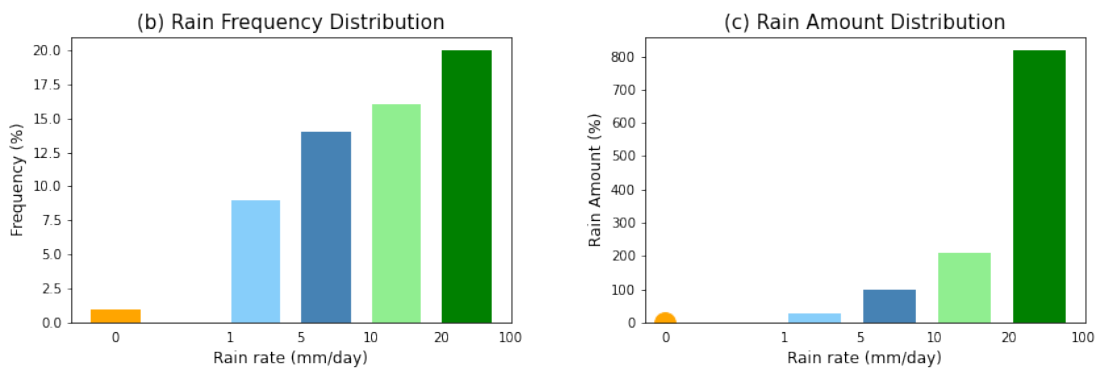
172 **Fig. 1.** Illustrative example of computing the rain frequency distribution and rain amount distribution.  
 173 (a) Time series of daily rainfall during 1<sup>st</sup> June – 30<sup>th</sup> July, 2020 in Delhi, India. Daily precipitation  
 174 accumulation in mm/day indicated by bars which are colour coded by rain rate (*Orange* for zero-  
 175 valued rain rate, grey for non-zero valued rain rates < 1 mm/day, *sky blue* for values between 1 to 5  
 176 mm/day, *dark blue* for values between 5 to 10 mm/day, *light green* for values between 10 to 20  
 177 mm/day, and *dark green* for values > 20 mm/day). Circles on the x axis indicate zero value. (b) Rain  
 178 frequency distribution and (c) Rain amount distribution (histograms are calculated from the example  
 179 time series).

180

181 For quantifying how often rain falls at different rain rates the distribution of rain frequency  
 182 (Fig. 1b) is constructed. The frequency of any logarithmically-spaced rain rate bin is  
 183 calculated as the percentage of the number of days in that bin to the total number of days in  
 184 the time period. The sum of the rain frequency distribution is 100%. We find that the  
 185 percentage of dry days is 13.3%, out of which 10% of days experience zero rain rate (orange)  
 186 and 3.3% of days experience non-zero rain rate below 0.1 mm/day coming from the rain rate  
 187 bin of 0 to 1 mm/day (grey) having a total frequency of 10%. Precipitation between rain rate  
 188 1 to 5 mm/day (sky blue) has the highest frequency of 38.3%, precipitation between rain rate  
 189 5 to 10 mm/day (dark blue) occurs on 21.7% of the total days, 11.7% of the rain happens  
 190 between rain rate 10 to 20 mm/day (light green), and the rain falls between rain rate 20 to 100  
 191 mm/day (dark green) for the remaining 8.3% of the days.



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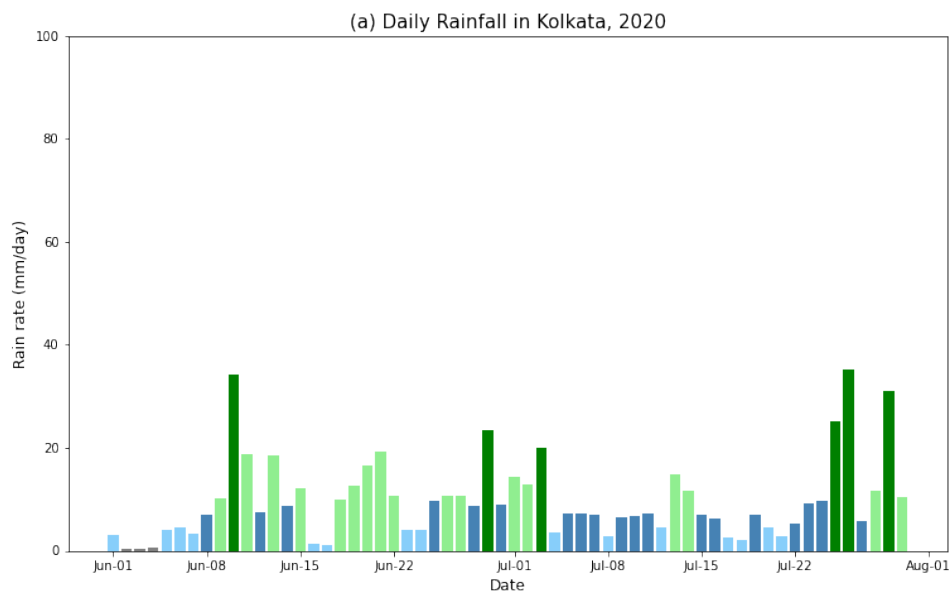
194 **Fig. 2.** (a) Time series of daily rainfall during 1<sup>st</sup> June – 30<sup>th</sup> July, 2020 in Mumbai, India. (b) Rain  
 195 frequency distribution and (c) Rain amount distribution computed for Mumbai’s rainfall time series.

196

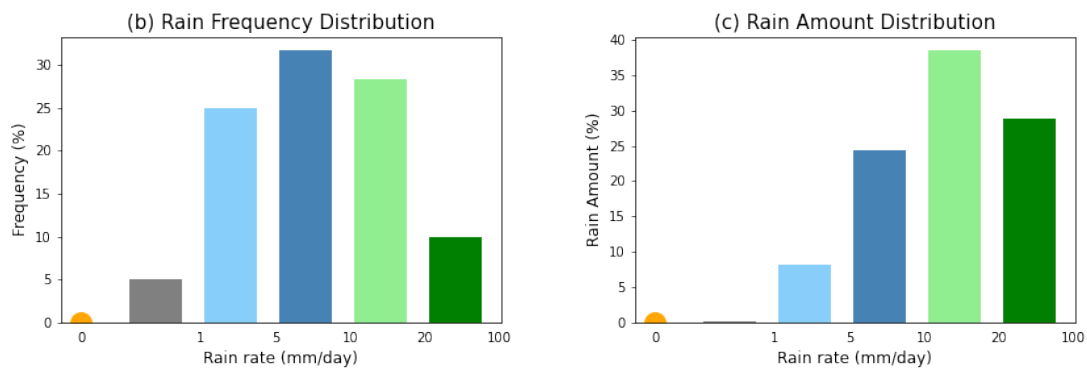
197 For quantifying, how much rain falls at different rain rates the distribution of rain frequency  
 198 as shown in Fig. 1(c). The rain amount for any logarithmically spaced rain rate bin is  
 199 calculated as the sum of all the rain falling within that particular bin, represented by a bar in  
 200 terms of percentage of the total precipitation for the complete time series. The sum of the rain  
 201 amount distribution is also 100%. The bin with low rain rates, i.e. 0 to 1 mm/day (grey  
 202 coloured bar) contributes only 0.4% of the total precipitation while, 1 to 5 mm/day (sky blue  
 203 coloured bar) having the highest frequency, contribute only 15.8% of the total precipitation  
 204 which is minimal. However, the moderate and heavy bins with rain rates between 5 to 10  
 205 mm/day (dark blue), 10 to 20 mm/day (light green) and 20 to 100 mm/day (dark green)



206 containing only 21.7%, 11.7% and 8.3% of the total days respectively contribute around 26%,  
 207 23.2% and 34.5% of the total precipitation respectively. Note that it is the heaviest bin  
 208 corresponding to the rain rates between 20 to 100 mm/day (dark green) with the highest  
 209 contribution, i.e. 34.5% of the total rainfall even though it only has 8.3% of the days. This  
 210 example demonstrates how rain amount distribution highlights the days with heavy rainfall,  
 211 and these heavy precipitation days contribute disproportionately towards the total  
 212 precipitation.

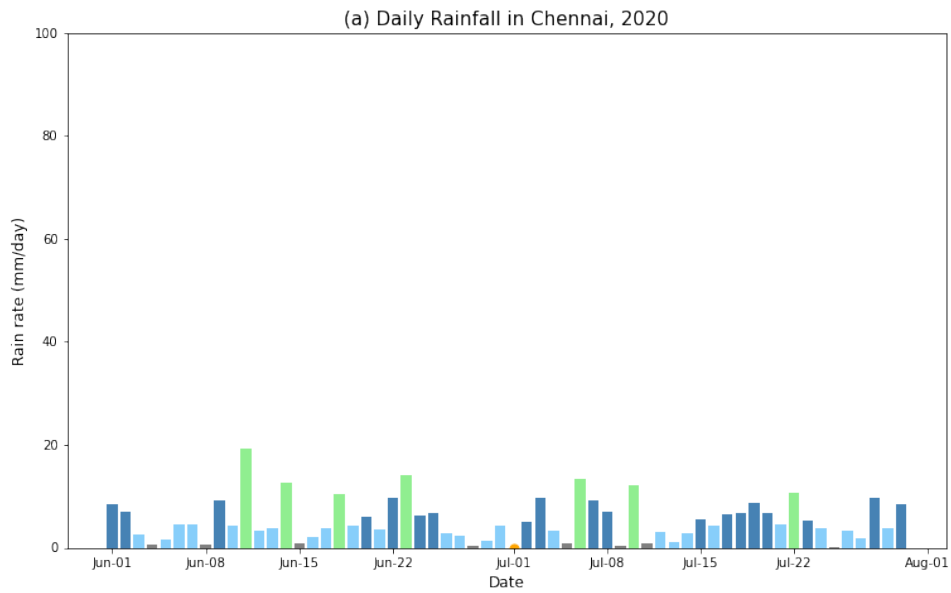


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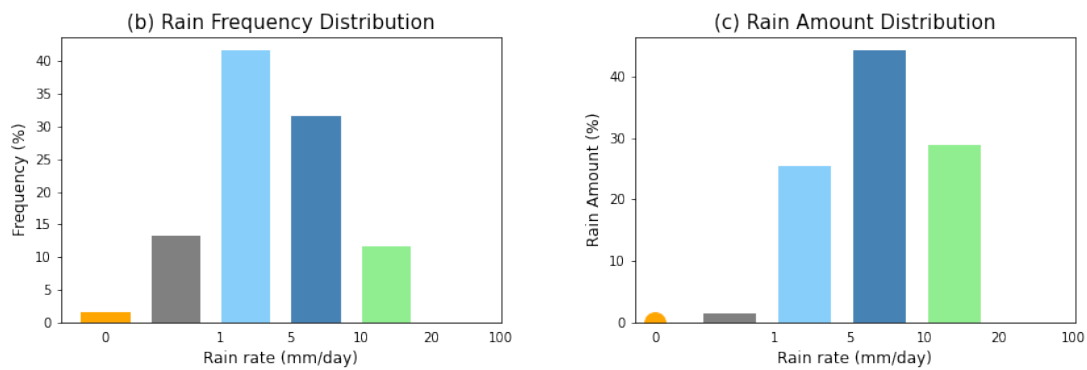


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215 **Fig. 3.** (a) Time series of daily rainfall during 1<sup>st</sup> June – 30<sup>th</sup> July, 2020 in Kolkata, India. (b) Rain  
 216 frequency distribution and (c) Rain amount distribution computed for Kolkata's rainfall time series.



217



218

219 **Fig. 4.** (a) Time series of daily rainfall during 1<sup>st</sup> June – 30<sup>th</sup> July, 2020 in Chennai, India. (b) Rain  
 220 frequency distribution and (c) Rain amount distribution computed for Chennai’s rainfall time series.

221

222 For the given two-month period under consideration, the total precipitation of all four cities is  
 223 405 mm for Delhi, 1154 mm (highest amongst all four) for Mumbai, 586 mm for Kolkata,  
 224 and 320 mm for Chennai. While the comparison of rain frequency distribution and rain  
 225 amount distribution of these cities can notably help gain insight into their daily precipitation  
 226 characteristics, let's keep our focus on just the rain frequency peak and rain amount peak (the  
 227 two defined precipitation metrics) for this case study. The rain frequency peak for Delhi and  
 228 Chennai lies in the rain rate bin of 1 to 5 mm/day, for Kolkata lies in the rain rate bin of 5 to  
 229 10 mm/day, while for Mumbai lies in the heaviest bin, i.e. 20 to 100 mm/day. The heavier the  
 230 rain rate bin is, the more amount it contributes towards total precipitation. Since the rain

231 frequency peak of Mumbai lies towards heavy rain rate it has the maximum total rainfall  
232 amount out of all four metro cities.

233 The rain amount peak for Delhi and Mumbai lies in the heaviest rain rate bin of 20 to 100  
234 mm/day which is evident since these heavy precipitation days contribute disproportionately  
235 towards the total precipitation except for Kolkata and Chennai for which it lies in the rain rate  
236 bin of 10 to 20 mm/day and 5 to 10 mm/day respectively. The rainfall in Chennai's case is  
237 highly concentrated in the 5 to 10 mm/day rain rate bin, and the rain frequency of the heavier  
238 bins is also relatively low due to which the rain amount peak for Chennai lies in the moderate  
239 bin. Note that even though the percentage of dry days is just 1.67% for Chennai compared to  
240 13.33% for Delhi's case, Delhi's total precipitation is more than that of Chennai. It is due to  
241 the fact that the frequency of days with heavy rain rate is higher for Delhi which is  
242 contributing much more towards Delhi's total precipitation as also corroborated by the higher  
243 rain amount percentage of the heaviest bin in case of Delhi.

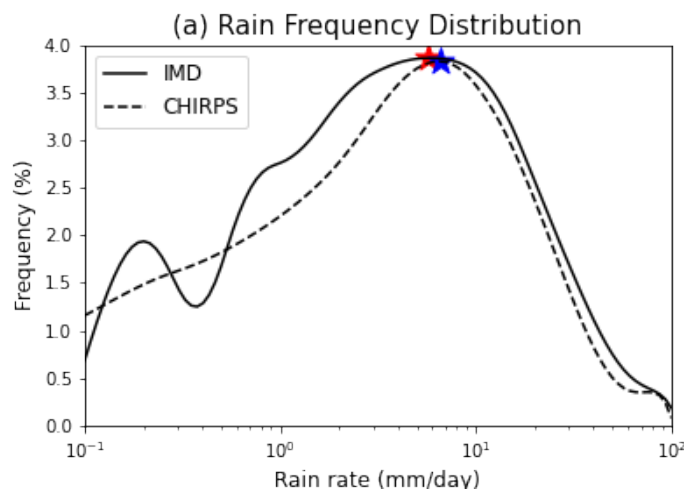
244 While these distributions with the rain rates have significant importance, the rain rate at  
245 which we obtain the peaks for these distributions, i.e. the two defined metrics: rain frequency  
246 peak and rain amount peak are also highly substantial in studying the characteristics of daily  
247 precipitation in India.

#### 248 **4. Quantifying India's typical daily precipitation characteristics**

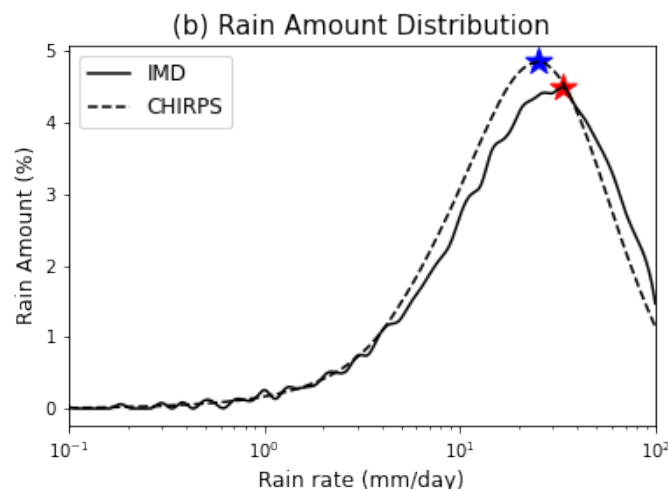
249 In order to develop a more robust understanding of national trends, we present the  
250 distributions of rain frequency and rain amount for India using the IMD daily dataset, as  
251 shown in Fig. 5 illustrate how often rain falls and how heavy it is when it falls at different  
252 rain rates. To analyse the differences that can exist in the defined precipitation metrics when  
253 computed using different datasets, Fig. 5 shows a comparison plot for spatially averaged  
254 annual mean rain frequency distributions (Fig. 5a) and rain amount distributions (Fig. 5b) for  
255 IMD daily dataset and CHIRPS daily global dataset. Note that the CHIRPS dataset is sliced  
256 to only consider the Indian region, i.e. latitude from 6.5°N to 38.5°N & longitude from  
257 66.5°E to 100°E, similar to the range that exists in the IMD dataset. Moreover, the resolution  
258 of CHIRPS dataset (grid size  $0.05^\circ \times 0.05^\circ$ ) was scaled down to match with the resolution of  
259 IMD dataset (grid size  $0.25^\circ \times 0.25^\circ$ ) for accurate comparison of the precipitation metrics  
260 between the two datasets. The two-precipitation metrics, i.e. the rain frequency peak and the  
261 rain amount peak are marked by a red star for IMD dataset and a blue star for CHIRPS  
262 dataset in Fig. 5.

263 The first metric shown in Fig. 5a is the rain frequency peak for India. It is defined as the non-  
 264 zero rain rate having the maximum frequency as denoted by the percentage of the total  
 265 number of days. In simple words, it is the most frequent rain rate greater than zero. The rain  
 266 frequency distribution (Fig. 5a) for India attains peak rain rate of 5.7 mm/day in the IMD  
 267 dataset (Fig. 5a) and 6.6 mm/day in the CHIRPS dataset (Fig. 5a), which will be referred to  
 268 as the rain frequency peak. Overall, the rain frequency distribution (with rain rate bins on  
 269 logarithmic scale) is negatively skewed as it falls off rapidly toward heavier rain rates than  
 270 low rain rates. The percentage of dry days, i.e. days with rain rate less than 0.1 mm/day is  
 271 73.9% in case of CHIRPS, slightly greater than 71.5% for IMD dataset. Though the  
 272 frequency for any rain rate for CHIRPS rainfall dataset is mostly lower than the frequency for  
 273 IMD dataset which is also evident by somewhat larger percentage of dry days for CHIRPS,  
 274 the rain frequency peak for CHIRPS occurs at a higher rain rate than IMD.

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278 **Fig. 5.** Comparison between the climatological distribution of annual mean (a) rain frequency and (b)  
279 rain amount for India from IMD daily dataset (solid line) for the years 1979 – 2016 & from CHIRPS  
280 daily global dataset (dashed line) for the years 1981 – 2020. The red star denotes the rain frequency  
281 peak in (a) and the rain amount peak in (b) for IMD data. The blue star denotes the rain frequency  
282 peak in (a) and the rain amount peak in (b) for CHIRPS data.

283

284 The second metric is the rain amount peak which is defined as the rain rate that contributes to  
285 the maximum amount of rain towards total precipitation. The rain amount distribution (Fig.  
286 5b) for India with rain rate bins on logarithmic scale is also negatively skewed with a longer  
287 tail at low rain rates than high rates and has relatively only one peak or maxima. The rain  
288 amount distribution curve for CHIRPS tracks the rain amount distribution curve for IMD for  
289 light rain rates, but then exceeds the IMD curve for moderate rain rates to finally dip below  
290 the IMD curve at heavy rain rates. It is because the total precipitation for the CHIRPS dataset  
291 is lower (CHIRPS also has a higher percentage of dry days than IMD) than the IMD dataset  
292 due to which the rain amount percentage is higher even if the frequency percentage is lower.

293

294 Though the rain amount percentage at moderate rain rates for CHIRPS rainfall dataset is  
295 higher than the percentage for IMD dataset, the rain amount peak for CHIRPS occurs at a  
296 lower rain rate than IMD, displaying the importance of heavy rain rates towards total  
297 precipitation. The rain amount peak for the CHIRPS global dataset is 25.4 mm/day (blue star  
298 in Fig. 5b) compared to the rain amount peak of 33.3 mm/day (red star in Fig. 5b) for IMD. It  
299 is necessary to remind that the values of these precipitation metrics, i.e. the rain frequency  
300 peak and the rain amount peak along with their distributions precipitation dataset used for  
301 analysis are influenced by the inherent uncertainties in these precipitation datasets. These  
302 precipitation metrics can significantly help in defining the typical daily precipitation  
303 characteristics in probably the most straightforward way possible, i.e. just by means of two  
304 numerical values. Although the accuracy of the precipitation metrics is determined by the bin  
305 width used for computations (smaller bins provide higher accuracy but require more sampling  
306 and thus more computational power), they don't depend on the bin width in a systematic  
307 manner.

308

## 309 5. Climatological distribution of rain in India

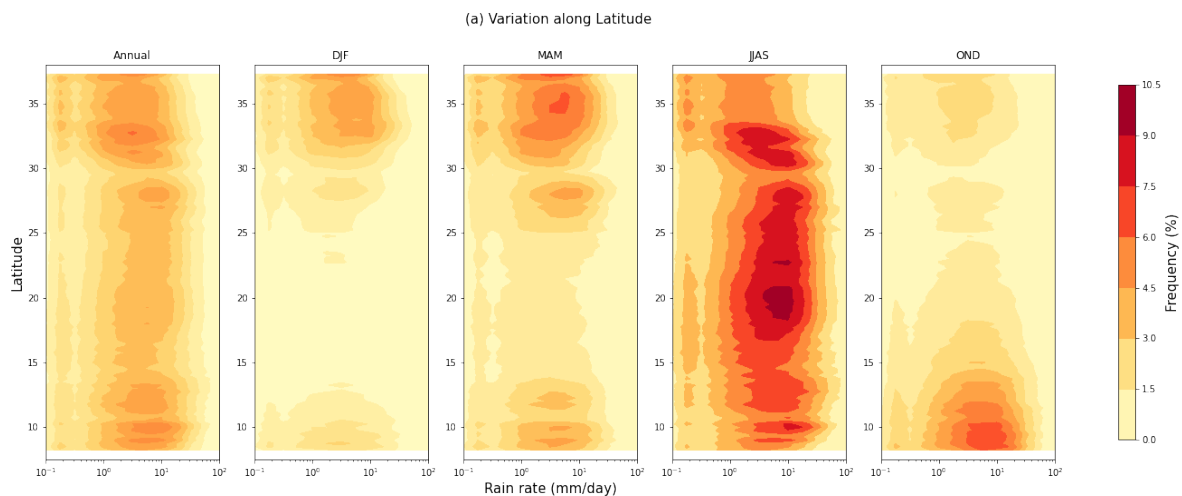
### 310 5.1 Seasonal variation of rain frequency distribution:

311 In order to highlight the spatial and temporal variability of rain frequency and amount, we  
 312 decompose them along the latitudinal and longitudinal transects. To further highlight their  
 313 seasonal variability, we look at the plots of summer monsoon (denoted as JJAS for June, July,  
 314 August and September), pre-monsoon (denoted as MAM for March, April, and May), post-  
 315 monsoon (denoted as OND for October, November and December), and winter (denoted by  
 316 DJF for December, January, and February). Note that the majority of the rainfall in this  
 317 country (80%) come from summer monsoon.

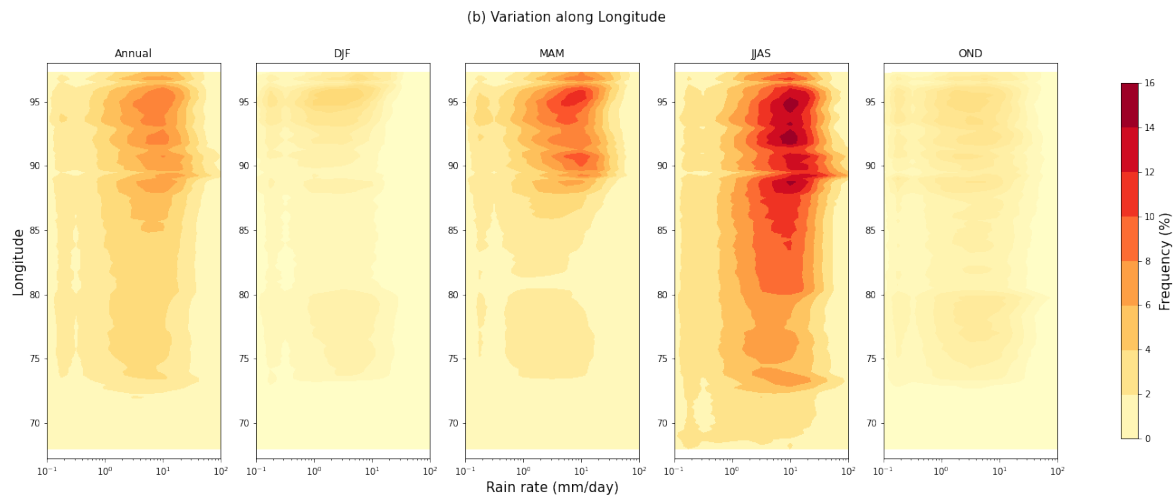
318 The rain frequency distribution of India is decomposed into contributions from different  
 319 latitudes in Fig. 6a and longitudes in Fig. 6b to observe the variation of distribution with  
 320 latitude and longitude. This variation of the rain frequency distribution is plotted for each  
 321 season side by side to study the seasonal variation of typical rainfall characteristics of India  
 322 as well. The rain rate is plotted on the x-axis on the log scale. Latitude or longitude is plotted  
 323 on the y-axis, where latitude varies from 6.5°N to 38.5°N and longitude varies from 66.5°E to  
 324 100.0°E. The rain frequency distribution corresponding to any latitude or longitude is plotted  
 325 using a colour bar (shown with the plot) with the frequency percentage at any rain rate  
 326 defined by the colour's intensity.

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330

331 **Fig. 6.** Climatological zonal, annual-mean rain frequency distribution for India from IMD daily  
 332 dataset for the years 1979 – 2016 and also stratified by seasons. (a) Seasonal variation with latitude  
 333 and (b) Seasonal variation with longitude. (DJF is December, January and February, MAM is March,  
 334 April and May, JJAS is June, July, August and September, and OND is October, November and  
 335 December)

336 The preliminary observation that can be made from the plots in Fig. 6a is that the frequency  
 337 percentage is highest (evident by the high intensity of the colour) in case of JJAS, which is  
 338 quite apparent as these months belong to the rainy season in India. For the DJF plot  
 339 corresponding to the winter season in India, the frequency is relatively low except for higher  
 340 latitudes. This can be accounted to the snow that falls in the northern Himalayan region  
 341 during the winter season, and this snowfall even continues till pre-monsoon (or summer)  
 342 season of MAM. The frequency starts decreasing during the post-monsoon (or autumn)  
 343 season of OND in most parts of India except for lower latitudes. This can be due to the  
 344 retreating monsoon winds that create extreme weather conditions characterised by high  
 345 humidity resulting in rainfall and cyclones in the coastal regions. Since India's maximum part  
 346 at very low latitude (around 10°N) is a coastal area, this leads to high rain frequency  
 347 percentage for these latitudes.

348 There is a break observed between the contours at around 30°N latitude for each rain  
 349 frequency distribution plot in Fig. 6a which can be an indicator of the change of the  
 350 contributing area from plains dominant region to mountains (Himalayas) dominant region. It  
 351 should be noted that while the frequency percentage varies significantly throughout India  
 352 with latitude, the rain frequency peak is relatively similar during the winter and summer  
 353 monsoon seasons. However, there is a sudden increase in the rain frequency peak during the

354 monsoon and autumn seasons for latitudes between 30°N to 40°N that mainly comprises the  
355 Indian plains. It is quite evident that these plain regions receive high rainfall during the  
356 monsoon and post-monsoon season while relatively low rain during the summer season due  
357 to the prevailing high temperatures. Though the plains receive heavy rainfall during JJAS and  
358 OND, the relatively low precipitation during the DJF and MAM counter that high rainfall,  
359 reducing the yearly mean frequency and rain frequency peak. Hence, India's annual mean rain  
360 frequency distribution plot is relatively uniform both in terms of rain frequency peak and the  
361 corresponding frequency percentage.

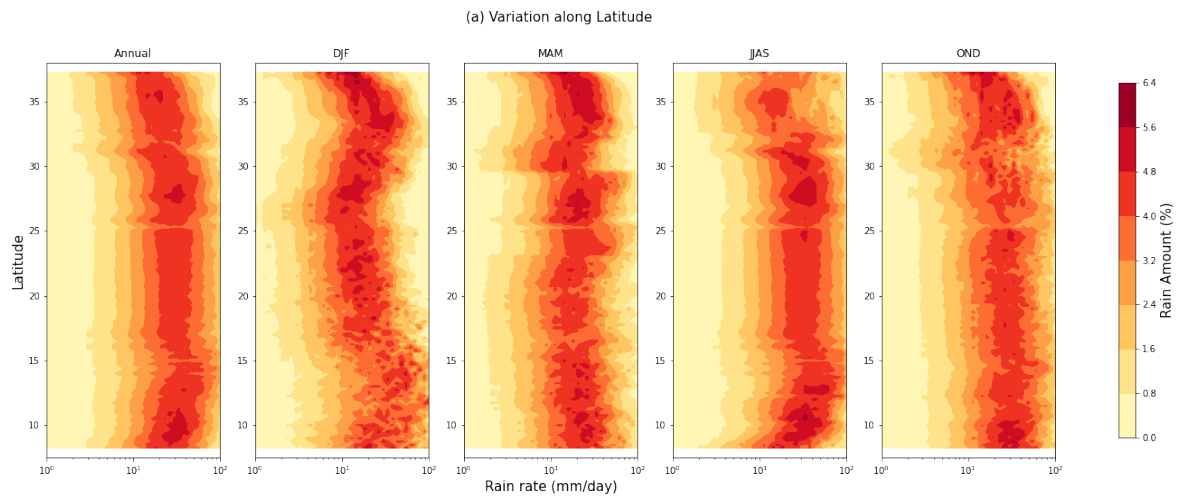
362 Similar to the variation of rain frequency distribution with latitude in Fig. 6a, the frequency  
363 percentage is highest in case of JJAS for the variation of rain frequency distribution with  
364 longitude as shown in Fig. 6b, which is evident as these months belong to the rainy season in  
365 India. When dividing India in longitudinal stripes, most of these stripes in the eastern part of  
366 India cover significantly less area. Thus, they are highly personalised to specific locations  
367 even if the rain frequency distribution is averaged over the region. Due to this, the frequency  
368 percentage is very high at eastern longitudes of India (90°E - 95°E) since the rain frequency  
369 distribution is mainly determined by the regions like Meghalaya that receive heavy rainfall,  
370 home to the area receiving the highest world's highest average annual precipitation of more  
371 than 11,000 mm. It is also because of this reason that the maximum frequency percentage is  
372 higher when averaged along longitude than in the case of averaged along latitude.

### 373 ***5.2 Seasonal variation of rain amount distribution:***

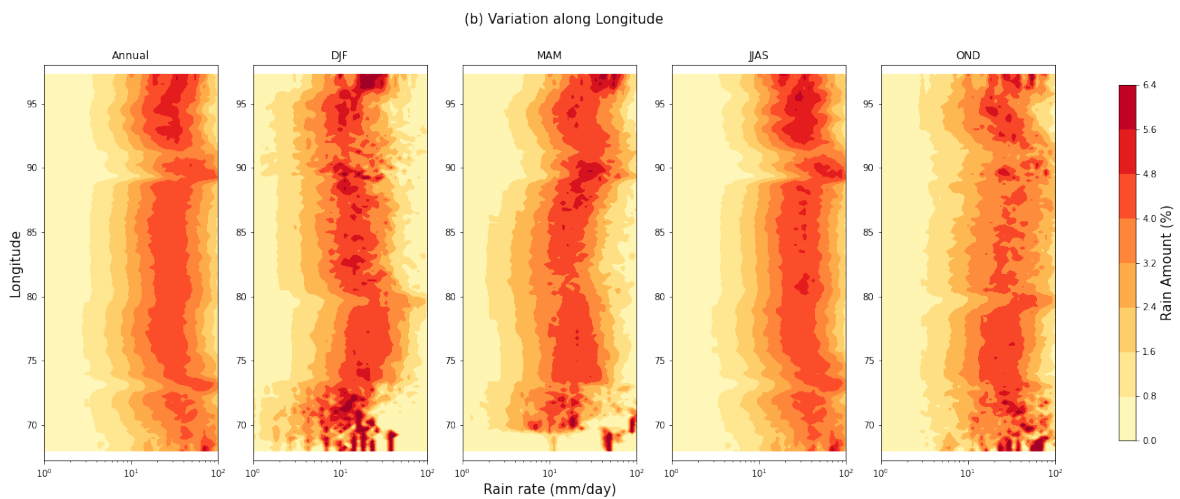
374 Similarly, the rain amount distribution is only plotted for the rain rates from 1 to 100 mm/day  
375 and while the rain amount distribution for the rain rates from 0.1 to 1 mm/day is cut off from  
376 these plots. It is done based on the observation made in Fig. 5b where it is evident that the  
377 rain amount percentage is significantly low for the rain rates of 0.1 to 1 mm/day and not  
378 crucial for our study.

379





380



381

382 **Fig. 7.** Climatological zonal, annual-mean rain amount distribution for India from IMD daily dataset  
 383 for the years 1979 – 2016 and also stratified by seasons. (a) Seasonal variation with latitude and (b)  
 384 Seasonal variation with longitude. (DJF is December, January and February, MAM is March, April  
 385 and May, JJAS is June, July, August and September, and OND is October, November and December)

386

387 It can be observed that the rain amount distribution is dominated by the heavy rain rates in  
 388 the plots for JJAS and OND, indicating a high amount of total precipitation which is apparent  
 389 as these months belong to the monsoon and post-monsoon seasons in India. The plots for  
 390 MAM and OND show a similar rain amount distribution and rain amount peak for all  
 391 latitudes throughout India during the summer and autumn seasons with a slight variation at  
 392 around 30°N. The annual rain amount peak at around 8°N is relatively low during the  
 393 monsoon season, i.e. in JJAS plot but still has high rain amount percentage. This indicates  
 394 that the total amount of precipitation for this region is lower than other latitudes which is also

395 evident by the comparatively lower frequency of rainfall observed in the rain frequency  
396 distribution plot of JJAS in Fig. 6a. The annual mean plot for the rain amount distribution,  
397 however, is quite uniform but has relatively lower rain amount peak at very low and very  
398 high latitudes. The high latitudes belong to the Himalayan region while the low latitudes  
399 observed in the JJAS plot receive relatively lower total precipitation, and thus, the rain  
400 amount peak is lower.

401 Similar to the variation of rain amount distribution with latitude in Fig. 7a, the rain amount  
402 peak is higher in case of JJAS and OND for the variation of rain amount distribution with  
403 longitude as shown in Fig. 7b, which is evident as these months belong to the monsoon and  
404 post-monsoon season in India. As explained earlier, when looking at India longitudinally,  
405 most of these stripes in the eastern part of India cover significantly less area. Thus, they are  
406 highly personalised to specific locations even if the rain amount distribution is averaged over  
407 the region. It is also because of this reason that the rain amount peak is relatively higher in  
408 the annual mean plot of rain amount distribution along the longitude as well.

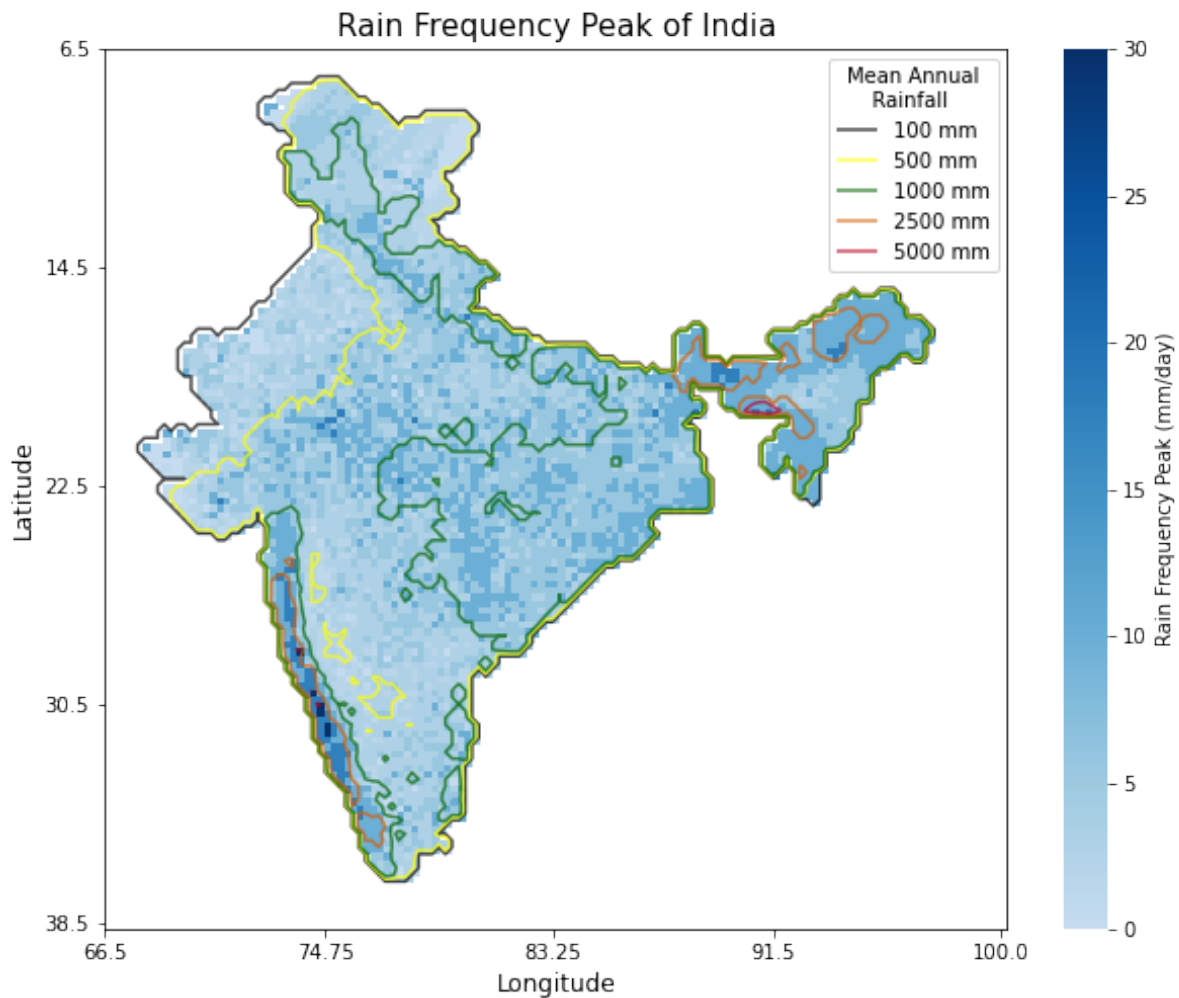
409 Similarly, the rain amount peak is also high for very low longitude. These longitudinal stripes  
410 belong to the western regions of Gujrat, which has a lower frequency of rainfall as observed  
411 in the rain frequency distribution plots (Fig. 6b). As most of the rain in this region falls at  
412 lighter rain rates, the heavy rain rate even having low frequency can still result in high rain  
413 amount peak since heavy rain rates contribute disproportionately towards total precipitation.  
414 Overall, the annual mean rain amount distribution plot for India is relatively similar along the  
415 longitudes, except for some minor spikes at around  $90^{\circ}\text{E}$ ,  $72^{\circ}\text{E}$  and very low longitudes. The  
416 reason for this significant spike at very low longitudes and  $90^{\circ}\text{E}$  has already been described.  
417 The increase in rain amount peak at around  $72^{\circ}\text{E}$  is because a larger area of these longitudinal  
418 stripes belongs to India's west coastal region, which receives high amounts of heavy rainfall.

## 419 **6. Maps of climatological annual mean precipitation metrics**

### 420 ***6.1 Rain frequency peak:***

421 To illustrate how the rain frequency peak metric varies throughout India, the rain frequency  
422 peak corresponding to each pair of latitude and longitude for India is plotted in Fig. 8. This  
423 plot is prepared using the IMD daily dataset with the latitude and longitude varying from  
424  $6.5^{\circ}\text{N}$  to  $38.5^{\circ}\text{N}$  and  $66.5^{\circ}\text{E}$  to  $100^{\circ}\text{E}$  respectively. Note that the rain rate in terms of mm/day

425 corresponding to the rain frequency peak metric is represented using a colour bar shown on  
426 the right side of Fig. 8.



427

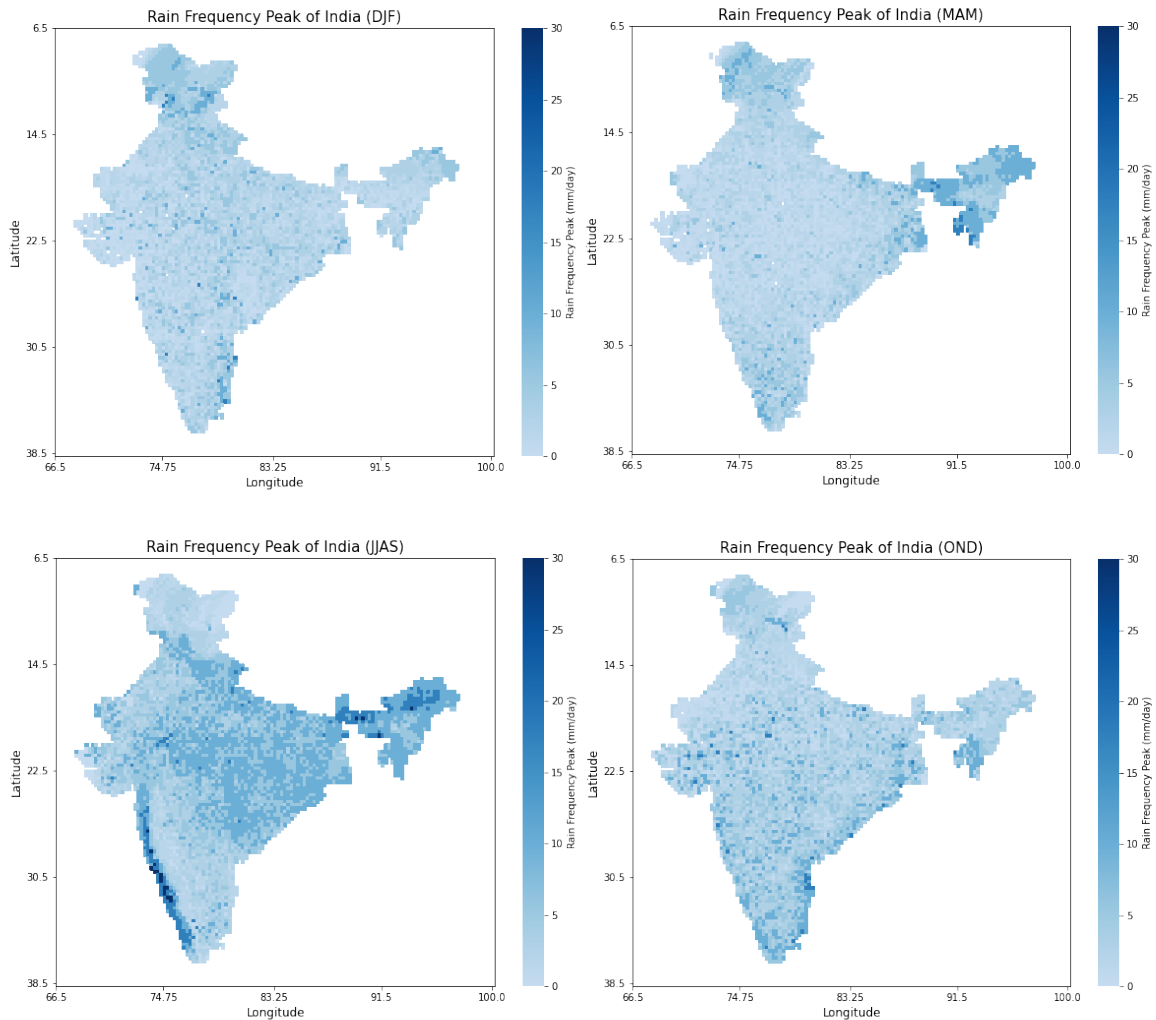
428 **Fig. 8.** Map of climatological annual mean rain frequency peak for India from IMD daily dataset for  
429 the years 1979 – 2016. (Rain frequency peak is represented in terms of mm/day, defined by the colour  
430 map on the right)

431

432 The western coast and the eastern parts of India, however, have higher values of rain  
433 frequency peak, especially the western coast. The rainfall in India's western coast is frequent  
434 due to the presence of the Western Ghats that result in higher rainfall because of the  
435 prevailing rain-bearing winds. The eastern parts of India also receive rain frequently, which  
436 was also observed in the plots for rain frequency distribution variation with the longitude  
437 (Fig. 6b). It must be remembered that this plot shows the most frequent non-zero rain rate  
438 throughout India. The maximum rain frequency peak for any region in India reaches up to a

439 value of 35 mm/day compared to 5.7 mm/day rain rate for annual mean rain frequency peak  
 440 averaged spatially throughout India.

441



442

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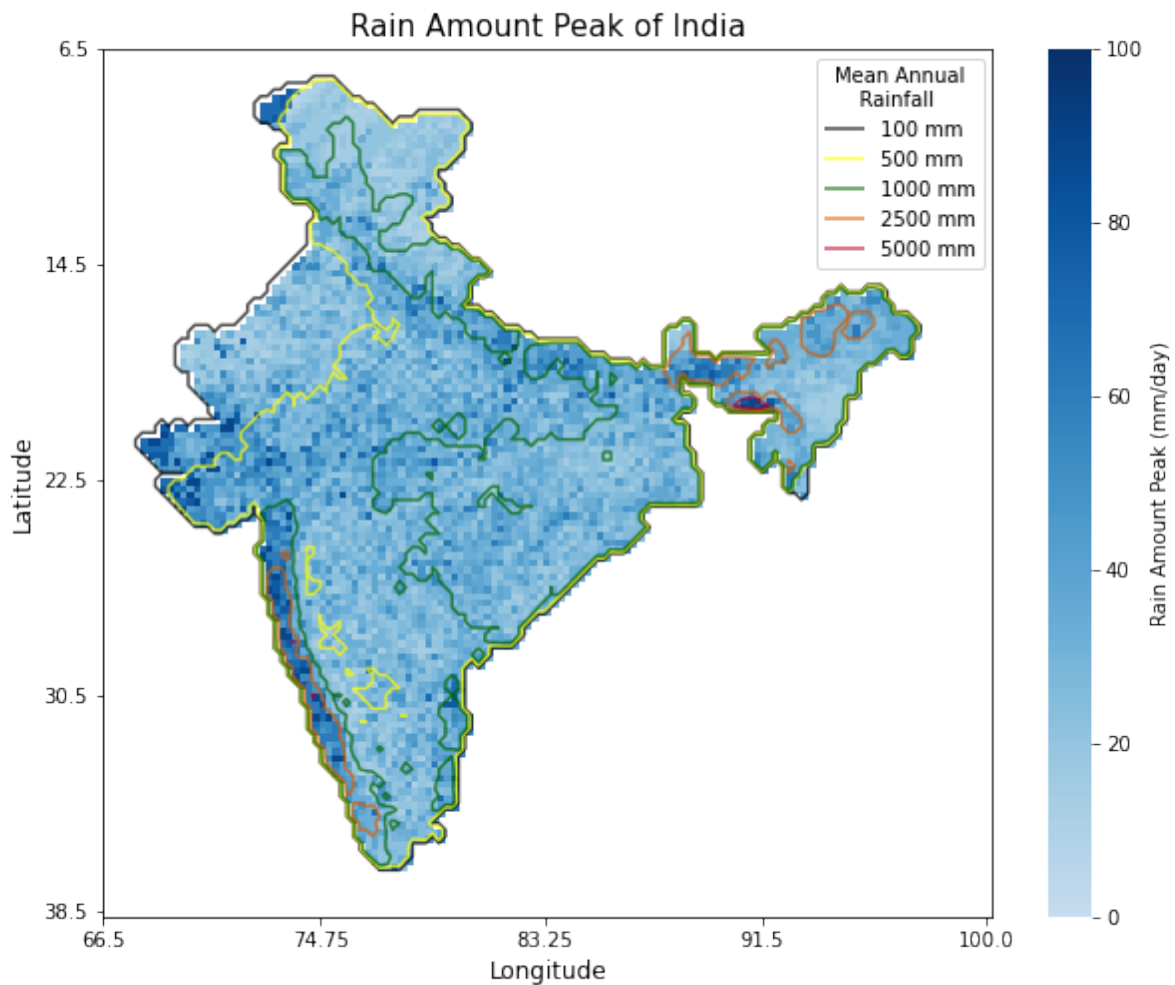
444 **Fig. 9.** Climatological zonal rain frequency peak for India from IMD daily dataset for the years 1979  
 445 – 2016 stratified by seasons. (DJF is December, January and February, MAM is March, April and  
 446 May, JJAS is June, July, August and September, and OND is October, November and December)

447 **6.2 Rain amount peak :**

448 To illustrate how the rain amount peak metric varies throughout India, the rain amount peak  
 449 corresponding to each pair of latitude and longitude for India is plotted in Fig. 10. This plot is  
 450 prepared using the IMD daily dataset with the latitude and longitude varying from 6.5°N to  
 451 38.5°N and 66.5°E to 100°E respectively. The points lying outside India's boundary are not  
 452 crucial for our study and hence not present in the figure. Note that the rain rate in terms of

453 mm/day corresponding to the rain amount peak metric is represented using a colour bar  
454 shown on the right side of Fig. 10.

455



456

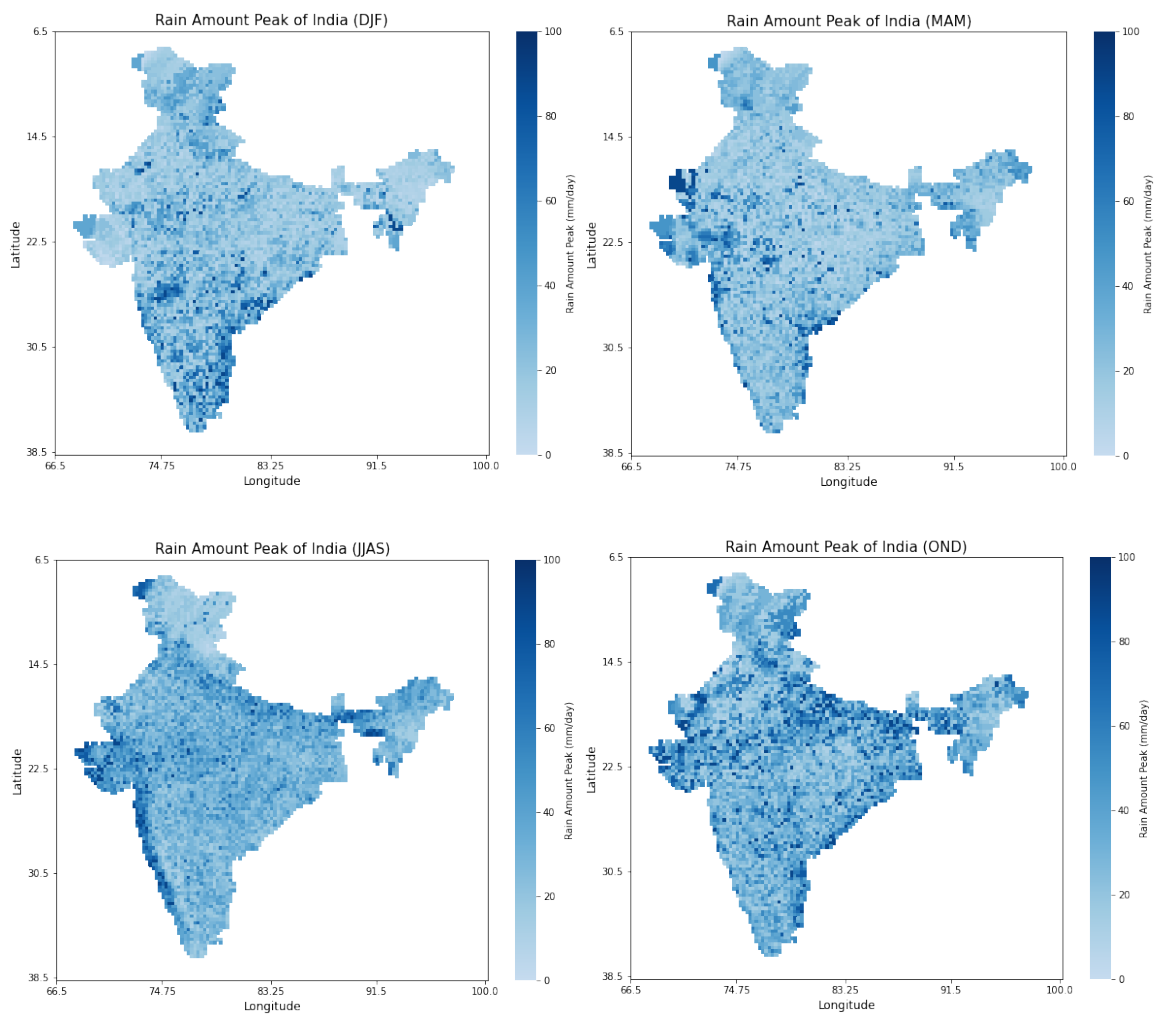
457 **Fig. 10.** Map of climatological annual mean rain amount peak for India from IMD daily dataset for  
458 the years 1979 – 2016. (Rain amount peak is represented in terms of mm/day, defined by the colour  
459 map on the right)

460

461 The western coast of India, as expected, has high rain amount peak compared to other regions  
462 of India. Meghalaya has the maximum rain amount peak, which is evident due to the fact that  
463 it experiences a high amount of rainfall at heavy rain rates from the south-west monsoon  
464 winds. It is quite intriguing to notice that while the rain frequency peak in the north-eastern  
465 part of Jammu & Kashmir is quite low, it has a significantly high rain amount peak.  
466 Similarly, the western parts of Gujrat also have high rain amount peak but a very low value of

467 rain frequency peak. This is because of the high contribution of heavy rain rates towards total  
468 precipitation compared to very light rain rates resulting in an increased value of rain amount  
469 peak for specific regions. The state of Rajasthan, which receives very low rainfall throughout  
470 the year has both low rain frequency peak and low rain amount peak, which is apparent  
471 because of the existence of Thar desert. The maximum rain amount peak for any region in  
472 India reaches up to a value as high as 90 mm/day compared to 33.6 mm/day rain rate for  
473 annual mean rain amount peak averaged spatially throughout India.

474



477 **Fig. 11.** Climatological zonal rain amount peak for India from IMD daily dataset for the years 1979 –  
478 2016 stratified by seasons. (DJF is December, January and February, MAM is March, April and May,  
479 JJAS is June, July, August and September, and OND is October, November and December)

## 480 7. Conclusions

481 As we have shown in this study, the two-precipitation metrics, i.e. the rain frequency peak  
482 and the rain amount peak can be crucial in defining the typical daily precipitation of any  
483 region. Two different datasets are analysed to compute the rain frequency peak and rain  
484 amount peak. The comparison between the rain frequency distribution and the rain amount  
485 distribution obtained for IMD dataset and CHIRPS dataset showed how these precipitation  
486 metrics may vary quantitatively (having its own uncertainties) but not qualitatively for  
487 different observational datasets. This comparison indicated that the distinctions between the  
488 rain amount distribution curves for both the datasets are less compared to the rain frequency  
489 distribution curves. This reveals that there are critical gaps that exist in our understanding of  
490 the rain frequency distribution, including the scope of accurate representation of precipitation  
491 in the observational datasets.

492 The rain amount distribution concentrates on the larger part of the total rainfall, thus making  
493 it a simple and conservative target since it is relatively easy to record them for models and  
494 somewhat hard to miss them from observations. Hence, more emphasis should be given to  
495 rain amount distribution than rain frequency distribution when analysing the typical  
496 precipitation characteristics. Studying the typical rainfall characteristics rather than just  
497 extreme rain events that quite often lead to floods is significantly essential for many reasons.  
498 Such a case is of the farmers for whom it does not matter how heavy rain can fall on a single  
499 random day but knowing how often it rains is crucial for them. It would be a worthy  
500 endeavour to enhance the state of our knowledge of the absolute magnitude of the total  
501 precipitation and the frequency distribution of rain.

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## 512 **Data Availability Standards**

513 CHIRPS precipitation data (<https://www.chc.ucsb.edu/data/chirps>) and IMD precipitation  
514 data ([https://www.imdpune.gov.in/Clim\\_Pred\\_LRF\\_New/Grided\\_Data\\_Download.html](https://www.imdpune.gov.in/Clim_Pred_LRF_New/Grided_Data_Download.html)) are  
515 freely available online.

516

517 **Compliance with Ethical Standards:**

518 The authors declare that they have no conflict of interest.

519 **Author Contributions:**

520 MS conceived and designed the analysis. YG performed the analysis. YG, MS, and BG wrote  
521 the paper.

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