

27 Abstract:

28 Floods are one of the most devastating natural hazards across the world, with India being one
29 of the worst affected countries in terms of fatalities and economic damage. In-depth research
30 is required in order to understand the complex hydrometeorological and geomorphic factors at
31 play and design solutions to minimize the impact of floods. But the existence of a historical
32 inventory of floods is imperative to promote such research endeavors. Though, a few global
33 inventories exist, they lack the spatio-temporal fidelity necessary to make them useful for
34 computational research due to reasons such as concentrating exclusively on large floods,
35 limited temporal scope, non-standard data formats etc. Therefore, there is an urgent need for
36 developing a new database that combines data from global and hitherto-underutilized local
37 datasets using an extensible and common schema. This paper describes the ongoing effort of
38 building the India Flood Inventory (IFI), which is the first freely-available, analysis-ready
39 geospatial dataset over the region with detailed qualitative and quantitative information
40 regarding floods, including spatial extents. The paper outlines the methodology that has been
41 adopted as well as some preliminary findings using the data contained in this inventory. This
42 dataset is expected to advance the understanding of flood processes in the worst affected region
43 of the world.

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51 Key words: Flood database, hazard assessment, impact assessment, hydrological modeling

52 **1. Introduction**

53 Floods continue to be one of the most devastating natural disasters across the world,
54 accounting for one-third of all global geophysical hazards (Smith and Ward, 1998). In India
55 alone, between 2010-2016, more than 10,000 people lost their lives and total damages of
56 around 16,500 crores were caused by floods, according to the Central Water Commission
57 (CWC, 2018). According to an Asian Development Bank report, floods have caused \$50
58 Billion of economic damage since 1990 (Patankar, 2019).

59 The existence of a comprehensive historical database of floods with adequate spatio-
60 temporal information is a key building block towards facilitating research into the causative
61 factors and impacts of floods. Several databases exist globally, such as the International
62 Disaster Database (EM-DAT), Relief Web (by United Nations), the International Flood
63 Network (IFNET), Global Active Archive of Large Flood Events by Dartmouth Flood
64 Observatory (DFO). The Global Flood Inventory (GFI) was one of the earliest efforts to
65 synthesize information from multiple sources and databases to create a continuous flooding
66 record (Adhikari et al. 2010). However, there are several limitations to GFI such as limited
67 time span from 1998-2008 as well as point locational information on floods with acknowledged
68 uncertainty. The global databases were also found to be of limited fidelity when it comes to
69 describing spatial extents of flooding impact as well as temporal coverage. The bigger
70 motivation behind the compilation of the India Flood Inventory (IFI) is the availability of large
71 amounts of valuable information currently stuck in printed documents published by various
72 government departments in India which have never found usage in furthering research due to
73 not being available as an easily accessible database. This data is ground-validated and can be
74 ascribed higher trustworthiness in terms of ascertaining damages, fatalities, as well as spatial
75 extents.

76 The IFI has been designed ground-up with careful consideration put into keeping it open,
77 standardized, and, extensible, with data recorded in a way that could be useful for quantitative
78 disaster modeling and analysis. The paper describes in detail the spatial and temporal coverage
79 of the India Flood Inventory, the augmentations made to existing datasets, incorporation of
80 new sources of information, and a summary of preliminary insights gained from this new
81 dataset.

82 **2. Existing flood databases and their biases**

83 Several multi-hazard databases catalogue flooding events with varying scope and
84 intended function. Two existing such databases are ReliefWeb (<http://www.reliefweb.int/>)
85 maintained by the United Nations Office for the Coordination of Humanitarian Affairs
86 (OCHA) and the International Flood Network (IFNET,
87 <http://www.internationalfloodnetwork.org/>) . ReliefWeb is more geared towards long-form
88 information about real-time events as they unfold and don't provide a historical database.
89 While the IFNET doesn't provide enough useful information over a long enough period to be
90 useful as a historical dataset. As such, both these databases were ignored during the creation
91 of the IFI.

92 A more widely-used international database is the The Emergency Disasters Database
93 (EM-DAT, <http://www.emdat.be/>) which is administered by the Center for Research on the
94 Epidemiology of Disasters (CRED) that collates natural and man-made disasters from 1900 to
95 present. The criteria for an event to be included is when 10+ people are killed, 100+ people are
96 affected, a state of emergency was declared, or a call for international assistance. This is the
97 longest readily available database available of disasters internationally. However, since the
98 inclusion criteria is impact-based, the data may be biased towards population centers like urban
99 areas.

100 The Dartmouth Flood Observatory (DFO, <http://floodobservatory.colorado.edu/>) is a
101 more comprehensive database exclusively focused on floods from 1985 to present. It's a simple
102 excel sheet titled Global Archive of Large Flood Events, where the data is sourced from news,
103 government sources, and satellite imagery. Though the data is richer than EM-DAT due to the
104 availability of flood start and end dates, country, details of affected locations, flooded river,
105 number of fatalities and damages, and spatial extent of flooding. The database also provides
106 both static images and analysis-ready imagery showing the flood-affected regions. Though it
107 has fairly good global coverage and higher data fidelity than EM-DAT, the database has lower
108 temporal coverage compared to other databases. The georeferenced record of flood event
109 locations is also only since 2006, limiting its viability in verification of long-term hydrologic
110 simulations, which is our primary objective behind creation of IFI.

111 The mainstay of IFI is the hitherto under-explored “Disastrous Weather Events”
112 (DWE) database compiled by the Indian Meteorological Department (IMD). This is a printed
113 publication that has been published by IMD since 1979 till date and is extremely hard to access
114 due to not being available online readily. The publication covers a wide gamut of natural
115 hazards such as snowfall, cold wave, heat wave, squall, gale, dust storm, lightning,
116 thunderstorm, hailstorm, floods and heavy rains, and cyclonic storm. The database has been
117 used very few times in scientific research. For example, De et al. (2005) has used a small subset
118 of this archive along with other databases to provide broad highlights of extreme weather
119 events in India over 100 years (1901-2004). In another study, a more focused study on floods
120 was performed with data from 1978-2006 highlighting the flood events, fatalities, and damages
121 (2013). But the data currently remains underutilized as it is not available publicly in a
122 geospatial-analysis ready format.. The effort involves tremendous amounts of manual and
123 automation work as well as careful verification, which the present study has sought to embark
124 upon, the details of which are explained in the next section. While designing the IFI, we have

125 been motivated by our desire to create a schema and database that is suitable for use in big data
126 modeling studies in the future.

127 **3. Compilation of the Flood Inventory**

128 **3.1 Sources of Information**

129 The IFI currently incorporates information from the following sources, which then undergoes
130 multiple levels of augmentation:

- 131 a. An annual printed publication named “Disastrous Weather Events” (DWE) by the
132 Indian Meteorological Department from 1979 till date. The database covers a wide
133 number of geophysical hazards, of which only floods were digitized for IFI
- 134 b. Dartmouth Flood Observatory (DFO)
- 135 c. Emergency Events Database (EM-DAT)

136 **3.2 Description**

137 The flood inventory has been structured into 2 parts: textual attributes and a spatial database.
138 In order to capture the qualitative and quantitative aspects of floods, we have defined several
139 terms for the database:

- 140 a. Unique Event Identifier (UEI)

141 Each flood event is assigned a unique identifier in an extensible format such as UEI-IMD-FL-
142 2015-0001, where IMD is the source dataset name, FL is for flood, 2015 is for year, and 0001
143 is for the serial event number of that year. This schema is flexible enough for us to incorporate
144 different disaster database within a common framework. It will also facilitate incorporation of
145 other geospatial disasters in the future and maintain interoperability, which may facilitate
146 research into compound disasters such as floods and landslides.

- 147 b. Start date

148 This is the start date of the flooding event. The IMD DWE contains more granular information
149 about the start and end of the event while databases EM-DAT often only indicate the months.

150 Often, the times provided are generic such as 3rd week of the month, which were transformed
151 to exact calendar dates. In order to maintain interoperability between various formats, all dates
152 conform to ISO 8601 (YYYY-MM-DD), which is the international standard for the
153 representation of dates and times.

154 c. End date

155 This is the end date of the flooding event which also conforms to ISO 8601 standards.

156 d. Duration

157 The number of days that have elapsed between the estimated start and end date of the event.

158 e. Main Cause

159 The primary cause of the flooding event, as recorded in the databases.

160 f. Location

161 This is only available for information incorporated from IMD. It indicates the names of
162 districts, states, and regions.

163 g. Districts

164 This information had to undergo lots of standardization and quality control as many district
165 names are wrongly entered in the original databases.

166 h. State

167 Substantial amount of data did not come with state information and only with region or district
168 information. These had to entered manually after consulting national geospatial databases. A
169 few states have undergone changes in their official names, which have also been corrected.
170 They conform to

171 i. Latitude and Longitude

172 A major lacuna of the existing databases is the non-availability of latitude and longitude of the
173 events, which is required for computational studies. For data sourced from DFO, the
174 coordinates were recorded as available. For IMD, based on the district and state information

175 provided, shapefiles were generated for the thousands of events, the centroid for which was
176 recorded in the database.

177 j. Severity

178 Only events sourced from DFO contains severity information.

179 k. Area affected

180 Only events sourced from DFO contains area affected information.

181 l. Human fatality, injured, and/or displaced

182 While DFO contained human fatalities and displaced for certain locations, IMD DWE contains
183 far more granular information regarding these. However, they were available in a verbose
184 textual style, which have now been recorded separately in the database.

185 m. Animal fatality

186 IMD DWE also contains information on animal casualties which are not unavailable in global
187 databases.

188 n. Description of casualties

189 IMD DWE contains textual description of casualties which have been kept in their original
190 format to provide more context.

191 o. Extent of damage

192 IMD DWE contains granular information on how the flood damage happened (E.g. Houses
193 and bridges collapsed, low lying areas flooded etc.). This is expected to provide more
194 contextual information to individual events.

195 p. Event Source ID

196 Wherever available, the original source IDs have been preserved in IFI in order to facilitate
197 cross-checking.

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200 **3.3 Methodology**

201 A systematic methodology was adopted to build the India Flood Inventory with the goal of
 202 conforming to modern interoperable standards and promoting computational hydrology
 203 research and applications. Different challenges were encountered with different datasets.
 204 EM-DAT and DFO were the two global datasets that were incorporated. DFO was a simple
 205 excel sheet and the attribute names were standardized for our dataset and provided the Unique
 206 Event Identifiers (UEI). For EM-DAT, the same operation was performed after accessing the
 207 global database.

208 However, majority of the work required the digitizing and processing of the IMD
 209 Disastrous Weather Events that are available only as paper publications. The IMD DWE
 210 dataset is the most detailed official dataset of flooding in India, but records are available in a
 211 format not readily amenable for computational work and a geospatial database (See, Figure 1).
 212 The dates were conformed to ISO 8601 standards and the human and animal casualties/injury
 213 numbers were extracted into separate columns. The most valuable part of this dataset was the
 214 information regarding districts that were affected. In order to generate GIS-friendly spatial
 215 extents of flood-affected areas, these district names were reverse-matched with a national
 216 district shapefile database (<http://projects.datameet.org/maps/districts/>) and a consolidated
 217 shapefile was generated for each event. Based on this event-based shapefile, the centroid of
 218 latitude and longitude was extracted and recorded. Each event was assigned a unique identifier
 219 like the global databases.

S.N.	Date / Period	Area affected	Intensity [River(s)]	Casualties	Extent of damage
9	17 & 18 Aug.	Shimla, Sirmaur, Solan, Una	Heavy rains	2 persons died & 5 others injured as wall collapsed.	iii) Vehicular traffic disrupted due to damage of roads.

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221 *Figure 1: Example of how IMD Disastrous Weather Events records information*

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223 **3.4 Uncertainty and limitations of the database**

224 Compiling a hazard database of this nature is crucial for developing future hydrologic studies
225 but requires painstaking work that is both scientifically and logistically challenging. The data
226 itself is inconsistent as different agencies record it in different ways, but without the data in a
227 common usable format, it remains a source of information rather than promoting further
228 research. The obvious bias in the global databases such as EM-DAT and DFO is concentrating
229 on only events with large impacts and covered by international media, while smaller events
230 and more granular information is better recorded in local databases such as IMD DWE. There
231 are several uncertainties inherent in the IMD DWE database which may be noted. Firstly,
232 administrative factors that may impact the information in these databases, for example over-
233 reporting when flood assistance from federal government is tied to damage reported by local
234 disaster management offices. Secondly, under-reporting of events may happen for locations
235 that have experienced fewer damages or casualties or located in more geographically distant
236 locations instead of the bigger cities. Reporting bias is especially true in developing countries
237 such as India where data collection is constrained due to budgetary reasons. This bias can be
238 reasonably expected to have reduced over the years and hence an obvious increase in the
239 number of flooding events may simply be due to better observational capabilities.

240 Finally, the other main source of uncertainty is the locational information. For example, the
241 IMD DWE dataset is often inconsistent in what it is recording as the location, using districts,
242 states, and regions interchangeably. The geographic centroid has been painstakingly recorded
243 by building shapefiles for every event but is likely being biased due to insufficient granularity
244 in the original database. But since no dataset is currently available for India, such information
245 is expected to provide a certain bound in terms of understanding these natural hazards.

246 **4 Preliminary analysis of hazards, fatalities, and damages**

247 **4.1 National and regional patterns**

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249 After the digitization, standardization, and augmentation, the India Flood Inventory was

250 analyzed for spatio-temporal patterns to understand the frequency and severity of the events,

251 the human and animal fatalities caused, and the causative factors. The IMD DWE dataset
 252 yielded the largest number of events (4176) with the highest spatio-temporal data fidelity.
 253 Collected manually from government records, it can also be regarded as the best available
 254 lower-bound of ground reality. EM-DAT contains 276 events but since the criteria for
 255 inclusion in the dataset is 10 or more fatalities and 100 or more injuries, it is inherently biased
 256 towards larger flood events. Additionally, DFO contained 262 events, but for a much shorter
 257 period. The summary of these databases is provided in Table 1 with the global databases, EM-
 258 DAT and DFO, contributing 6% and 5% to the IFI respectively, while the national database of
 259 IMD DWE is contributing 89%, which substantially increases the sample size and
 260 consequentially the robustness of studies based on this dataset. For the common period of
 261 1985-2016 between the three databases, the number of recorded events is 206 for EM-DAT,
 262 235 for DFO, and 3487 for IMD DWE.

263 *Table 1: Summary of databases incorporated into the India Flood Inventory*

Summary of information	Global		National
	EM-DAT	DFO	IMD DWE
Date	1926-2019	1985-2019	1967-2016
Number of records	276	262	4176
Percentage of India Flood Inventory	6%	5%	89%
Number of records (1985-2016)	206	235	3487

264
 265 Figure 2 shows the evolution of the number of floods events in India for different time periods
 266 since 1926, as available in the 3 data sources. The increasing trend is clearly visible in all
 267 three, though some of it may be attributed to better data collection over the years as well.

268 With a slope of -2.45 , the IMD data shows the sharpest trend, followed by EM-DAT with
269 a slope of -0.16 , and DFO with a slope of -0.13 .

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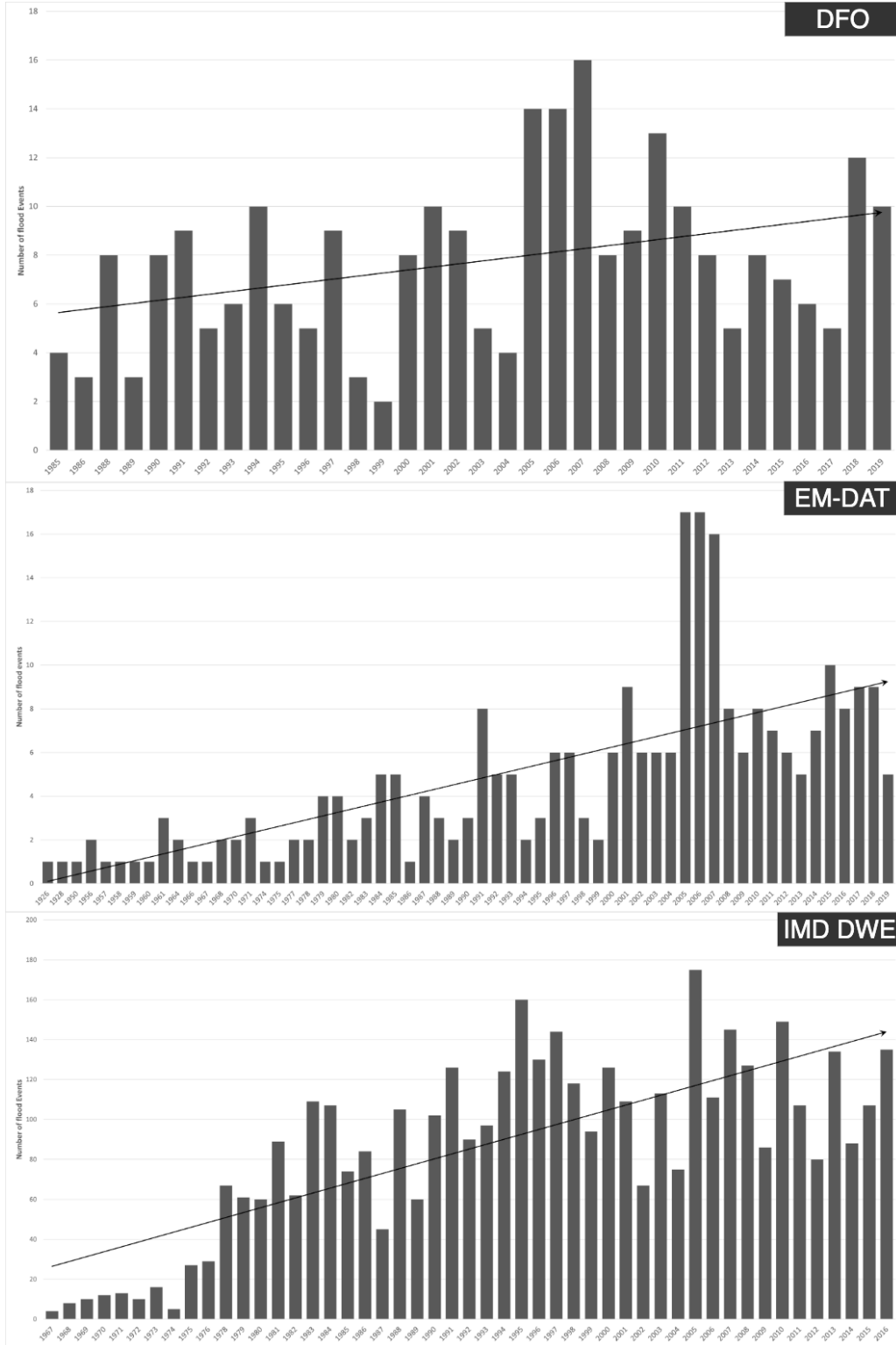


Figure 2: Temporal evolution of the number of floods at national scale

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277 Figure 3 shows the temporal evolution of the number of flood fatalities over India since the
278 1970s. The number of fatalities per year vary widely, with the lowest being 67 in 1974,
279 highest being 5473 in 2013, and an average of 1387 fatalities per year in the IMD DWE. The
280 single worst [Database code: UEI-IMD-FL-2013-0131/UEI-DFO-FL-2013-0001] event in
281 terms of fatalities is the 2013 June 14-18 cloudburst-induced heavy rainfall and flash flood
282 event in Uttarakhand that caused more than 5000 human fatalities. Multiple landslides and
283 avalanches were reported at several locations with 12 out of the 13 districts badly affected,
284 with the worst affected being Chamoli, Pithoragarh, Rudraprayag, and Uttarkashi districts.
285 Similarly, events with extraordinarily large number of fatalities were also observed in 1979,
286 1980, and 2000. In a well-known event known as Machchhu dam failure of Morbi disaster,
287 the dam breach inundated the town of Morbi killing approximately 1500 people according to
288 official estimates. In 1980, around 1300 people perished due to severe flooding and ensuing
289 drowning, house collapses, landslides, and boat tragedies in large swathes of Uttar Pradesh.
290 Overall, there is a definite increasing trend in flood fatalities across India.

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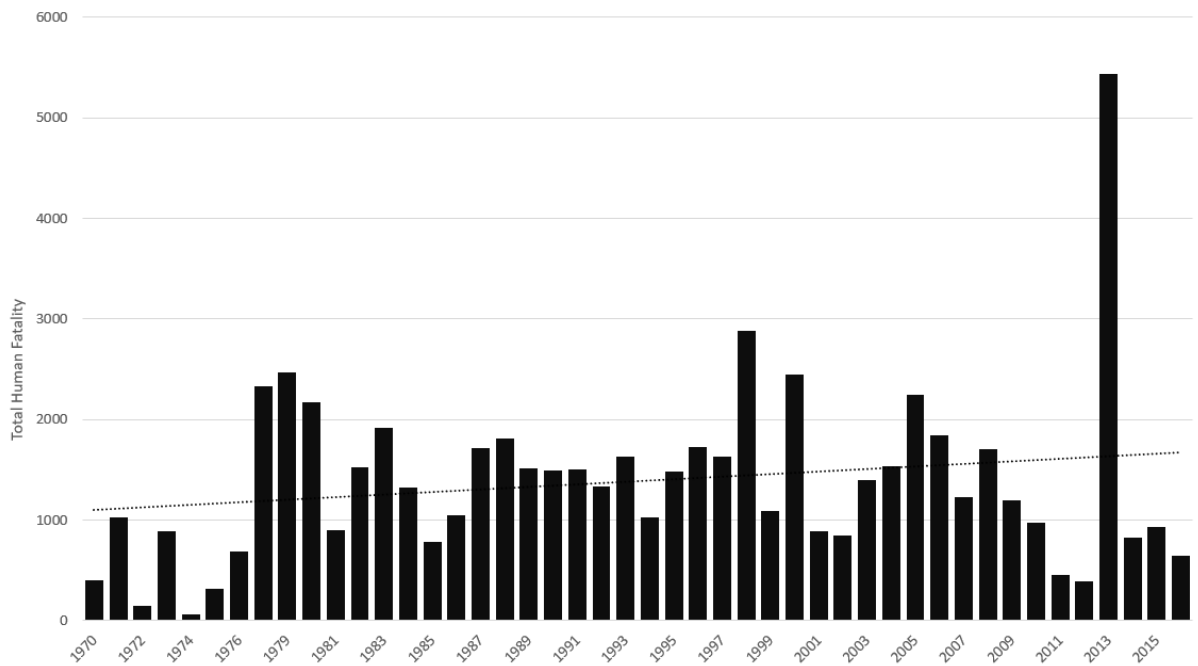


Figure 3: Temporal evolution of the number of flood fatalities at national level

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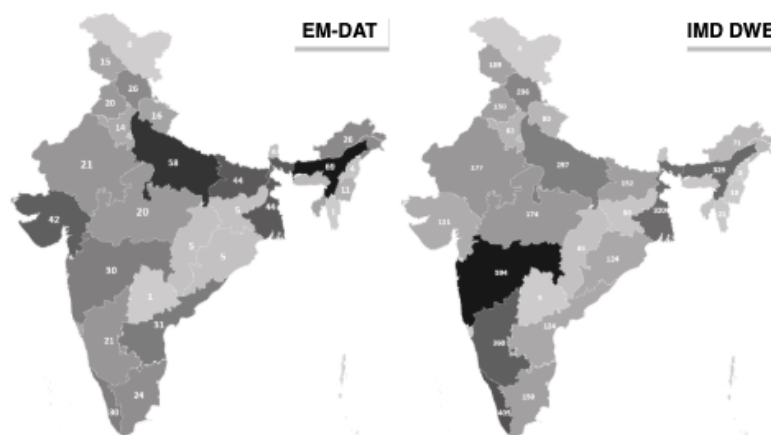
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298 This progressive increase in flood events and human fatalities in India has also been reported
299 by other studies. Shreshtha (2008) analyzed a global dataset to conclude that there is a
300 dramatic increase in flood disasters from 1976-2005 in South Asia. This is in line with has
301 been observed in Europe and attributed to factors that have appeared or gained influence
302 over the years, with higher sensitivity to smaller disasters, and a consequent increase in
303 the reporting of such disasters (Hoyois and Guha Sapir, 2011). In a study of disasters globally,
304 Jonkman (2005) found that Asian rivers are the most significant in terms of number of people
305 killed and affected, with flash floods resulting in highest average mortality per event. There is
306 a threefold rise in widespread extreme rain events over Central India (Roxy et al., 2017),
307 increasing spatial variability in observed Indian rainfall extremes (Ghosh et al., 2012), and the
308 increasing frequency of heavy rainfall events in peninsular, east and north east India that is
309 correlated with flood risk (Guhathakurta et al., 2011).

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311 India is divided into 28 states and 8 union territories, consisting of total 36 administrative
312 entities. For the sake of simplicity, all of them have been referred to as states here. The
313 number of flooding events for each state and database type has been shown in Figure 4. Since,
314 DFO only records the latitude and longitude, state-wise statistics were not reported. The top 5
315 states have been reported in Table 2, with Assam experiencing 25% of the national totals in
316 EM-DAT. While IMD DWE mentions Maharashtra as the highest with roughly 14% of total
317 number of floods. Similarly, IMD DWE shows Uttar Pradesh experiencing the highest
318 number of flood fatalities at roughly 17%. There could be several possible reasons for this
319 difference between the two databases at the state level. Since, EM-DAT limits itself to events
320 that have caused death of 10 or more people or affected 100 or more people, it is inherently
321 biased towards more destructive events at a larger scale. The Brahmaputra causes longer-
322 duration riverine floods throughout the state of Assam, causing enormous damage to life and
323 property on a yearly basis. On the other hand, IMD has field offices across the country from
324 where they collect the data and is also not limited by death and damage criteria like EM-DAT
325 is. Thus, states like Maharashtra record the highest number of floods in this database, which
326 may be due to more frequent flooding events at a shorter scale.



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328 *Figure 4 shows the spatial distribution of the number of flooding events in IFI from IMD DWE and EM-DAT.*

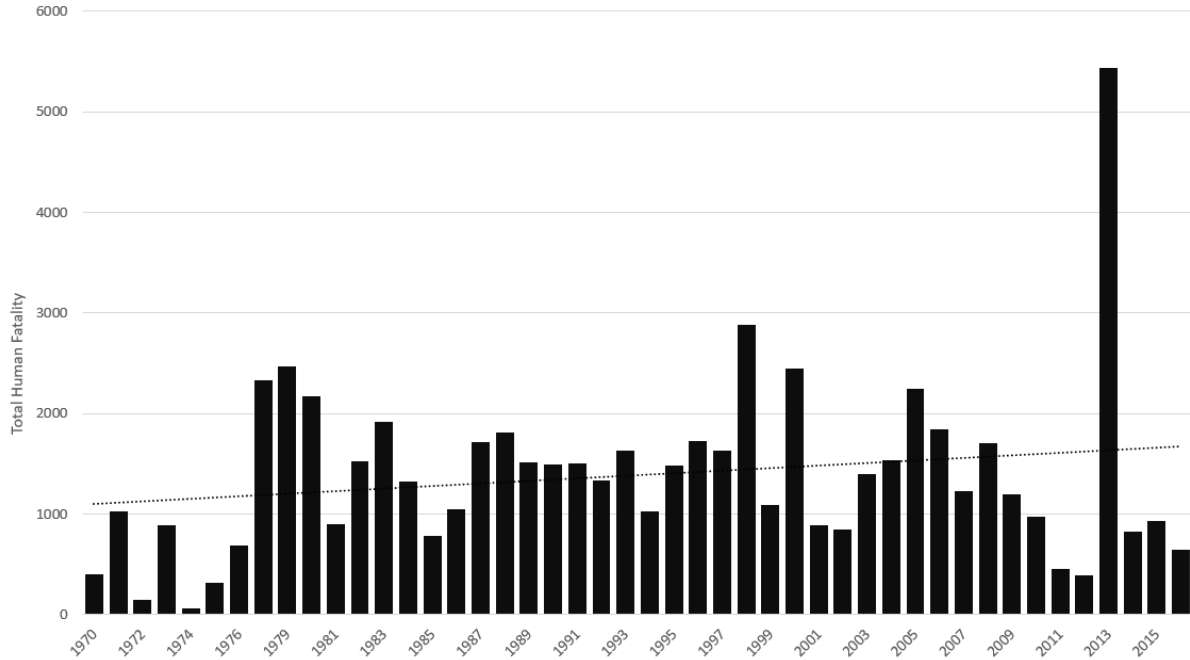
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Table 2 Top 5 states with the highest number of floods and fatalities

EM-DAT			IMD DWE			IMD DWE		
States	Number of Floods	National %	States	Number of Floods	National %	States	Number of Fatalities	National %
Assam	69	25.00	Maharashtra	594	14.22%	Uttar Pradesh	12158	16.78%
Uttar Pradesh	58	21.01	Kerala	405	9.70%	Maharashtra	6943	9.58%
Bihar	44	15.94	Karnataka	360	8.62%	Uttarakhand	6725	9.28%
West Bengal	44	15.94	Assam	329	7.88%	Bihar	6366	8.79%
Gujarat	42	15.22	West Bengal	320	7.66%	West Bengal	6081	8.39%

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Figure 5: Temporal evolution of the number of flood fatalities at national level

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Figure 6 and Figure 7 shows the temporal evolution of the causative factors according to DFO

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and IMD DWE. As expected, monsoonal rains dominate the cause behind floods in India. A

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substantial number of cloudbursts have been recorded in IMD DWE, which is a cause of

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major concern due to short but devastating nature of its impact. It is to be noted that the

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causative factors in IMD DWE are not encoded systematically and often don't record any

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causative factors. Hence, they need to be approached with caution.

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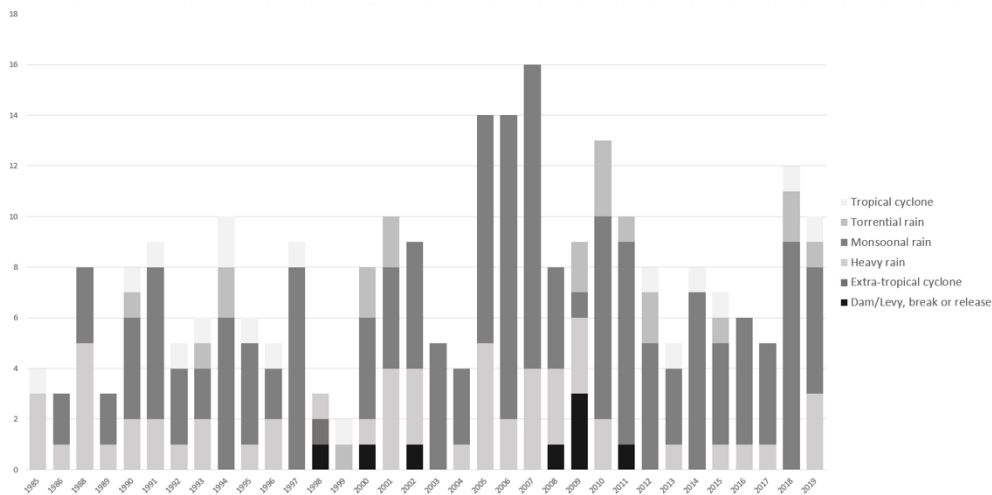


Figure 6. Temporal evolution of causative factors according to DFO

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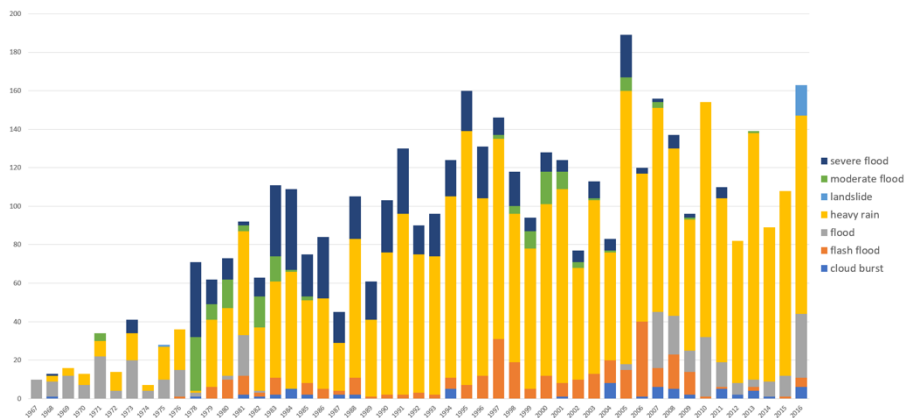


Figure 7. Temporal evolution of causative factors according to IMD DWE

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4.2 Seasonal patterns of hazard and fatalities

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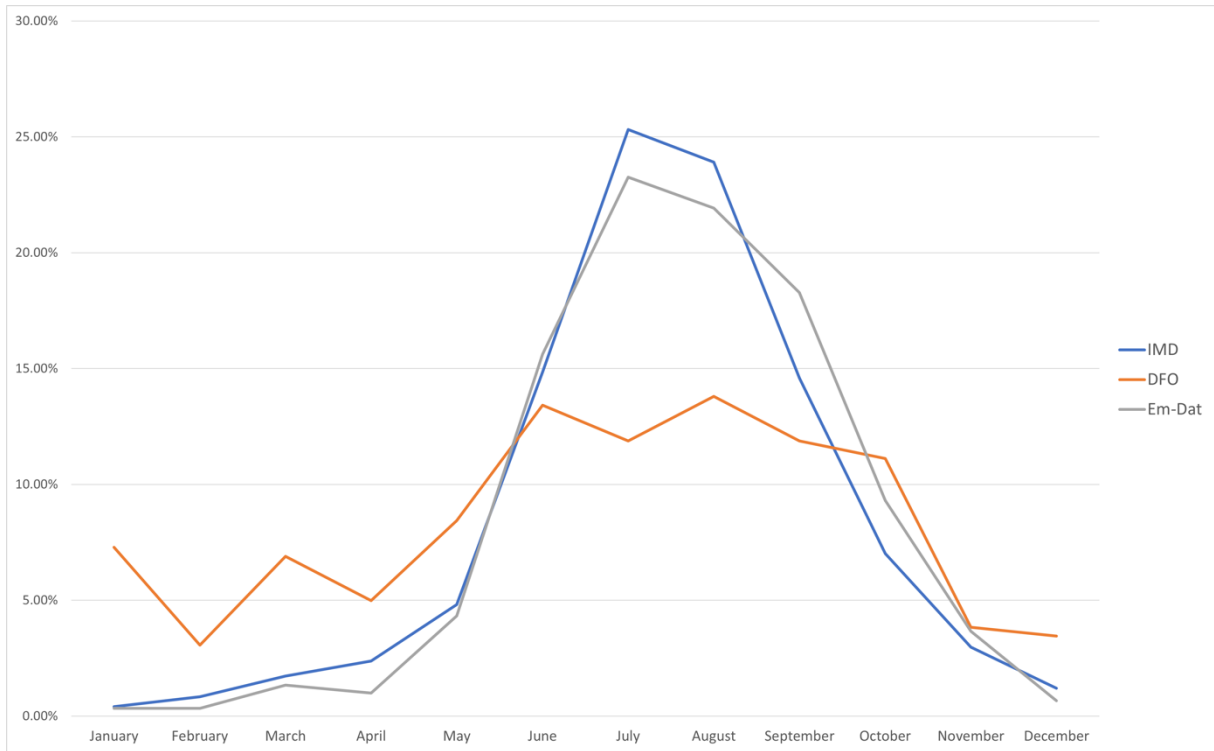
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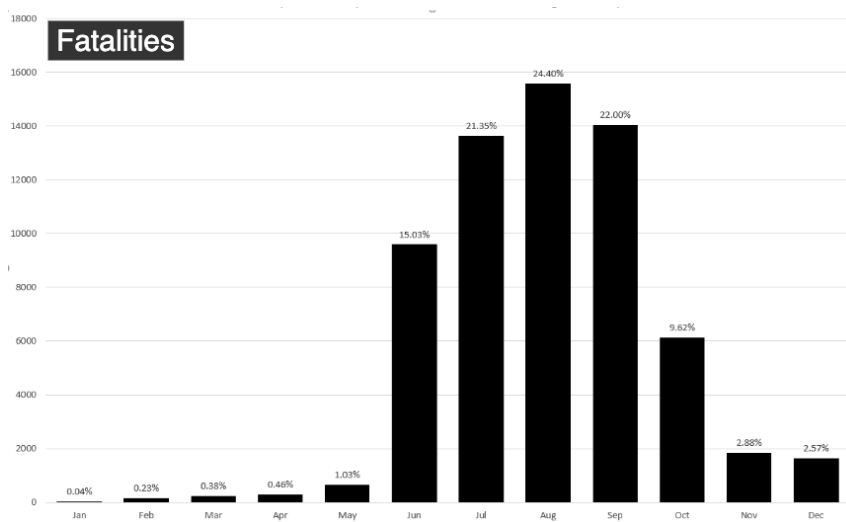
Flooding in India varies according to monsoon activity and tropical cyclone patterns. Both the number of flooding events and human fatalities are dominated by the monsoon and post-monsoon season, which can be attributed to outbreak of monsoonal rain activity across the country. Water levels in rivers rise while reservoirs are running at capacity during this period, causing widespread floods in the country. Averaging globally, the flood season starts in May and peaks in August (Adhikari et al., 2010), but in India, it peaks in July and August as shown in Figure 8 and Figure 9. Monsoons from June to September record the majority of the flood events at 79% of the total, according to the IMD DWE. It also accounts for 83% of the total fatalities year-round. Figure 8 shows the comparison of the number of floods per month as

362 a percentage of the yearly totals. IMD and DFO records the highest number of floods in
363 July, while EM-DAT records the highest late monsoon in August. It may be noted that
364 EM-DAT concentrates on floods causing large number of fatalities and injuries,
365 compared with IMD, which represents floods without regard to damage.



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367 *Figure 8. Monthly variation of the number of floods at national scale as a percentage of yearly totals*

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369
370 *Figure 9: Monthly variation of the number of flooding fatalities according to the IMD DWE*

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4.3 Digitization and possible Applications

375 Another uniqueness of this dataset is the availability of flooding extents in modern formats
376 such as Shapefile (.shp), GeoPackage (.gpkg), and KML (.kml file). These extents have been
377 calculated for each event by matching the district/state level information available in these
378 datasets. Since these extents come with temporal information, remote sensing data such as
379 Landsat/Sentinel/MODIS etc. could be used to develop inundation imagery for specific
380 flooding events. This would be very helpful in validating hydrologic modeling simulations in
381 various locations.

382

383 **5 Conclusions and Future Work**

384 The India Flood Inventory (IFI) is India's most comprehensive database of flooding events
385 that is a) multi-source, b) standardized to international data specifications, and c) freely
386 available in modern geospatial formats. Currently, IFI includes 49 years [1967-2016] of
387 flood data digitized from the IMD Disastrous Weather Events. It also includes 34 years
388 [1985-2019] of data from the Dartmouth Flood Observatory (DFO) and 93 years [1926-
389 2019] of data from the International Disaster Database (EM-DAT). Best possible effort has
390 been made to augment and standardize them to a common schema, which makes IFI an
391 analysis-ready dataset for a wide-variety of applications related to flood hazard, risk, and
392 exposure.

393 The majority of floods in the country happens in the monsoon season, which is 79% of the
394 yearly total, with a peak in July. On the other hand, the number of flood fatalities during the
395 same period is 83% of the yearly total, with a peak in August. The seasonality of flooding is
396 well indicated in the country, which can guide flood management and disaster reduction
397 efforts in the country. The large flood plains of the country such as Uttar Pradesh, Assam,
398 Maharashtra, Bihar, and West Bengal experiences the highest number of floods and fatalities.

399 While, the hill states such as Uttarakhand have experienced catastrophic events, with some of
400 the highest per capita death rates in the country.

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402 This study has only begun a preliminary investigation into the spatio-temporal variations of
403 flooding in India. Further investigation into the causative factors will be necessary to
404 determine the structural and non-structural flood mitigation measures that may be necessary.

405 This dataset is expected to contribute towards encouraging such diagnostic and prognostic
406 efforts. One of the goals is this study was to propose a standard specification for recording
407 natural disaster information which will aid future data collection efforts. The extensible
408 framework proposed for India Flood Inventory can be used to integrate data from large
409 number of disparate databases for any number of natural hazards. An on-going upgradation
410 to the inventory is to use a cloud-based platform to derive the spatial inundation extents for
411 the events using satellite imagery. This compilation is designed to be a massive ongoing
412 effort going forward as we digitize and incorporate sources of information from other federal
413 and state disaster management agencies, most of whom maintain independent datasets and
414 are expected to be of even higher fidelity.

415 **Compliance with Ethical Standards**

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

418

419 **Author Contributions**

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421 MS proposed the idea. SH and RB digitized the dataset. MS and AJ analyzed the dataset. MS
422 and AR wrote the manuscript. SOP and DSP provided the dataset and comments on the
423 manuscript.

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425 **Data Availability**

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427 The dataset and shapefiles are freely available from this repository:

428 <https://github.com/hydrosenselab/India-Flood-Inventory>

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