

Floods have become less deadly: an analysis of global flood fatalities 1975-2022

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1 **Floods have become less deadly: an analysis of global flood fatalities 1975-** 2 **2022**

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12 **Abstract**

13 Floods are amongst the most frequent disasters in terms of human and economic impacts. This study
14 provides new insights into the frequency of loss of life at the global scale, mortality fractions of the
15 population exposed to floods, and underlying trends. A dataset is compiled based on the EM-DAT
16 disaster database covering the period 1975 until 2022, extending previous studies on this topic. Flood
17 impact data is analysed over spatial, temporal and economic scales, decomposed in various flood
18 types and compared with other natural disasters. Floods are the most frequent natural disasters up to
19 1,000 fatalities, and flash floods lead to the highest mortality fractions per event, i.e. the number of
20 deaths in an event relative to the exposed population. Despite population growth and increasing flood
21 hazards, the average number of fatalities per event has declined over time. Mortality fractions per
22 event have decreased over time for middle and high-middle-income countries, but increased for low-
23 income countries. This highlights the importance of continuing and expanding risk reduction and
24 adaptation efforts.

25 **Keywords:** fatalities, floods, mortality, trends, disaster risk reduction.

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27 statistical methods.

28 **Author contributions:** S.N.J developed the concept and method, contributed to and reviewed the
29 analyses. A.C. created the database, contributed to the methods and performed the analyses. L.M.B.
30 has contributed to the methods, and reviewed the analyses. All contributed to the writing of this
31 manuscript and approved the final manuscript.

32 **1 Introduction**

33 The Global Risk Report by the World Economic Forum (2023) highlights extreme weather as one of
34 the most severe threats facing the world, alongside climate action failure and infectious diseases.
35 Analyses of the global trends in extreme weather events and their societal, economic and ecological
36 impacts are important for understanding local risks, and options for reducing these threats (Kreibich et
37 al. 2022). Life-threatening and existential extreme weather risks are being widely discussed in the
38 scientific literature because of uncertainties in impact trends and underlying causes, such as
39 anthropogenic climate change (Pielke 2021, Huggel et al. 2022). Moreover, fatalities from natural
40 hazards are the top priority for guiding policy on disaster risk reduction: the Global Target A of the
41 Sendai Framework aims to substantially reduce global disaster mortality by 2030 (UNDRR 2015).
42 Information and analysis on this indicator, and understanding underlying causes and changes over
43 time are therefore essential. The number of systematic studies into fatalities from disasters however is
44 relatively limited, and there are multiple interpretations of past trends and their causes.

45 Some studies have focussed on global scale analysis of flood fatalities. Notably, fatalities from river
46 floods have been analysed using insurance records (Jongman et al. 2014) and public databases
47 (Tanoue et al. 2016; Zhang et al. 2020). Studies focussed on global fatalities from inland (Jonkman
48 2005) and coastal floods (Bouwer and Jonkman 2018), or several aggregated flood types (Hu et al
49 2018). Others focussed on specific aspects such as seasonal patterns in mortality (Alfieri et al., 2020)
50 and benefits of flood loss prevention (Chen et al. 2020). Formetta and Feyen (2019) focus on
51 vulnerability of a broader group of climate-related disasters and showed that mortality has decreased
52 over time. These studies generally use global disaster databases such as EM-DAT which facilitates
53 analyses into impacts from extreme weather events (Guha-Sapir et al 2013).

54 A key question is if and how flood impacts have changed over the last decades. Past studies find that
55 while total global flood fatalities have increased (Tanoue et al. 2016, Jonkman 2005), the number of
56 fatalities per event have decreased (Bouwer and Jonkman 2018, Hu et al. 2019). However, for better
57 understanding of the relative impacts from flood events over time, not only the total fatality numbers,
58 or fatalities per event, but also the mortality fraction – that is the number of deaths relative to the
59 exposed population – is important. Existing studies, however, did not calculate mortality fractions per
60 event, and considered more limited time-periods. Mortality fractions per event are an indicator of
61 relative vulnerability, which is often neglected in literature. This indicator allows analysis and
62 comparison of vulnerability between countries and over time. Moreover, the frequency, scale,
63 impacting mechanisms and level of mortality vary between different flood event types, but these
64 aspects are often disregarded.

65 The objective of this study is to provide a comprehensive global analysis of flood fatalities over
66 spatial, temporal and economic scales, decomposed into various flood types. We go beyond previous

67 studies by a) including and distinguishing between fatalities from different flood types (riverine,
68 coastal and flash floods); b) providing a comparison against other climate-related hazards; c)
69 calculating and analysing changes in total and averaged event fatalities, as well as mortality fractions
70 per event and the relationship with income levels; and finally d) extending the previous global study
71 of Jonkman (2005) by more than 20 years of data, with recent years witnessing severe floods and
72 fatalities in several countries.

73 In order to meet the objective we developed a dataset of flood fatalities at the global scale, according
74 to flood type, country(ies) affected, and time of occurrence – for the period 1975 – 2022 (Curran et
75 al., 2023). The principal basis is data from the widely used EM-DAT International Disaster Database,
76 with quality checks, and additional information sources – particularly for coastal floods (see section
77 2). Global impacts in terms of flood fatalities are analysed by flood type and compared with those of
78 other disasters. We comprehensively analyse temporal trends and relate these to economic
79 development and other factors (section 3). The final section 4 summarizes the main findings and
80 discusses the limitations and implications of this work.

81 **2 Approach: EM-DAT and the global flood fatalities dataset**

82 This section outlines the data sources, data processing and methods used in the analyses in this study,
83 as well as associated limitations.

84 *2.1 The EM-DAT International Disaster Database*

85 The primary data source for our study is the EM-DAT International Disaster Database collected by
86 the Centre for Research on the Epidemiology of Disasters in Brussels – CRED at UC Louvain in
87 Brussels, Belgium (Guha-Sapir et al 2013). The information in this database is collected from various
88 international, governmental, commercial and NGO agencies. While the database stretches back to the
89 year 1900, CRED was established in 1973, and more recorded events with increased detail are
90 included in the database since that time. For this reason, the present study uses data starting from
91 1975.

92 For entry to the EM-DAT database (<https://www.emdat.be/database>), an event must fulfil at least one
93 of the following criteria; 10 or more deaths reported, 100 or more people affected reported, a state of
94 emergency is declared or international assistance is requested. This set of criteria ensures that also
95 relatively small (flood) events, e.g. an event with 100 people affected and 1 fatality, will be included
96 in the database. Each event is given a unique identifier related to the starting year and country of the
97 event. Each recorded event has 50 fields that detail information about the type, location, impact,
98 strength, and characteristics of the event, as well as the reason that it has been included (e.g. because
99 it surpassed a certain threshold in terms of economic loss or fatalities). Not all fields may be
100 completed.

101 2.2 *The global flood fatalities dataset*

102 We have used EM-DAT to develop a database of flood events with one or more fatalities, also
103 distinguishing the various flood types (Curran et al. 2023). The dataset covers the period 1975 – 2022,
104 thus updating the previous global analysis for the period 1975 - 2001 by Jonkman (2005) with 20
105 years of data. While many of the factors influencing trends in fatalities such as population growth and
106 economic development are applicable to all natural disasters, anthropogenic climate change impacts
107 only floods considered to have hydrological and meteorological causes. As we are interested in
108 multiple drivers for changes in risk, including climate, we focus on flood types caused by extreme
109 weather, for which frequency and impacts are expected to be affected by anthropogenic climate
110 change through changes in extreme rainfall and changes in sea level, winds and extreme coastal surge
111 levels (IPCC, 2023). We have not included floods originating from tsunamis and dam failures in our
112 temporal analyses of floods as these have other causes. Statistics of flood fatalities are compared
113 against those from other disaster types that are associated with extreme weather, specifically:
114 (wind)storms, wildfires, extreme temperature, droughts and landslides.

115 Within EM-DAT, we have considered events from the following subgroups that are within the natural
116 disasters group: hydrological (floods and landslides), meteorological (extreme temperature and storm)
117 and climatological (drought and wildfire) - see (IRDR, 2014) for definitions. We have classified the
118 flood events by flood type, distinguishing riverine floods, coastal floods, and a joint category of
119 pluvial and flash floods caused by heavy local rainfall – see (Jonkman, 2005) for definitions. In
120 particular, we found that coastal flood events had to be selected within EM-DAT from two events
121 types: storm surge/ coastal flood and storm/cyclone (Bouwer and Jonkman, 2018). As coastal storms
122 are often characterized with a combination of extreme winds, coastal surge and heavy rainfall, these
123 events have been categorized in both the ‘flood’ and ‘storm’ categories within EM-DAT. For many
124 of the events designated as ‘storms’, the main cause of fatalities is often considered to be flooding
125 (Rappaport, 2014). Storm winds may play a large role in economic damage during such events. We
126 have reclassified a number of storm events as floods, either based on the event descriptions, or if they
127 were included in previous study on coastal floods (Bouwer and Jonkman, 2018). Also, other flood
128 events have been reviewed based on the available metadata to verify if they were in the appropriate
129 category and / or to assign a flood type. Figure 6 in the appendix includes a schematic of the
130 categorisation process.

131 Table 1 gives an overview of the events included in the present study. In total 5582 flood events with
132 1 or more fatalities are included. We reclassified 592 of these events from the ‘storm’ category to the
133 coastal or pluvial flood type using the process described above. Furthermore, our verifications and
134 reviews resulted in 3 extra coastal and 3 extra flash flood events being added to the database of flood
135 events (Table 3 in the appendix). (see supplementary information). The verification and classification

136 process also resulted in 34 events that could not be categorised as one of the flood types due to
 137 insufficient information.

138 **Table 1 Overview of number of records available within (adjusted) EMDAT database**

Subsets of total dataset	Number of records
Total number of singular natural disaster events in EMDAT	23,892
Natural disaster events between 1975-2022	21,570
Natural disaster events with fatalities (≥ 1) reported	17,459
Extreme weather events	9758
Storm	2514
Landslide	687
Drought	562
Wildfire	413
Extreme temperature	385
Flood	5582
Riverine	3913
Flash/pluvial	1508
Coastal	127
Uncategorised	34

139

140 *2.3 Data processing and analyses*

141 This section summarizes steps in our analysis as well as some important inputs and assumptions.

142 In this study we assumed that the number of affected people reported in EM-DAT equals the number
 143 of people exposed to the flood event as was done in previous studies (Jonkman 2005; Bouwer and
 144 Jonkman 2018). This could be a conservative estimate for estimating exposure; Some of the affected
 145 people may not be exposed directly to floodwaters, for example when they are evacuated before the
 146 flood or have been only exposed to wind effects in a coastal storm surge.

147 We have analyzed temporal trends in reported numbers of events (section 3.1) and the numbers of
 148 fatalities and affected (section 3.2). Both the absolute numbers of fatalities and affected, as well as the
 149 average numbers per event are analysed for a given year. The significance of the (logarithmic) trends
 150 reported in section 3.2 is tested with a Mann Kendall test.

151 We have analysed mortality fractions per event by flood type (section 3.3). Mortality fractions for an
 152 event ($F_{D,i}$) were determined based on the below formula.

153
$$F_{D,i} = N_i / N_{exp,i}$$

154 Where: N_i – number of fatalities for an event, $N_{exp,i}$ – number of exposed people for an event

155 The mortality fraction is a metric of the relative vulnerability for an event. It expresses the sensitivity
 156 of impacts to a given flood hazard, and it can be used to compare regions and patterns in time. Based

157 on the mortality fractions for single events we computed average mortality for a set of events (e.g.
158 floods in a single year, flood type and income group).

159 For the analysis of mortality by income group (section 3.4), World Bank income data
160 (<https://data.worldbank.org/>, accessed October 31 2023) has been appended to the database based on
161 the country and year of the event. The GDP for an event depends on both the year of the event and the
162 total GDP of the country affected. Note that the values are given in terms of ‘Adjusted Income’,
163 which is the average income of the country affected by a flood hazard in USD, adjusted for inflation.
164 The bands defining income classification (low, lower middle, higher middle, high) have also been
165 taken from the above World Bank data.

166 *2.4 Known issues and limitations*

167 Any attempt to evaluate and categorise major disaster events comes with a number of issues,
168 including biases in reporting of events over time (Jones et al. 2022).

169 In relation to time, there is more data available for more recent events (see also section 3.1). This is
170 mitigated by using (where appropriate) statistics averaged over all recorded events per year and by
171 reporting mortality fraction per event as a metric of the severity of impact. We have also analysed the
172 trends for the annual number of reported events and the average number of killed and affected per
173 event for various alternative thresholds (more than 10, 100, 1000 fatalities).

174 For event categorisation, even with the steps that were mentioned above, assigning a single category
175 to an event like a hurricane which can cause flooding, wind damage and landslides is not clear-cut,
176 and a known limitation of the study.

177 Some entries in EMDAT show the number for ‘Total Affected’ to be less than ‘Total Deaths’,
178 resulting in a mortality fraction above 1. For this reason, a minimum value for ‘Total affected’ has
179 been set to the sum of fatalities, injured and affected.

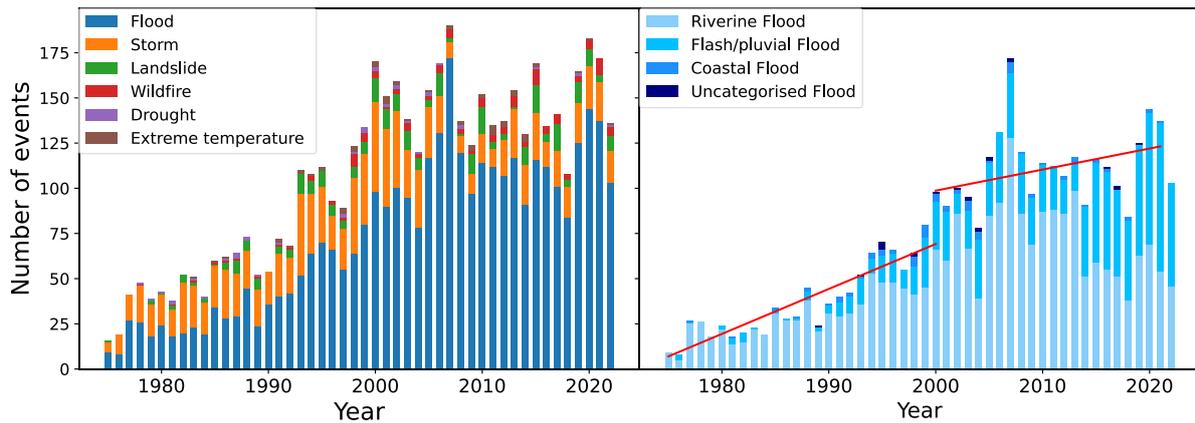
180 An practical issue specific to the EM-DAT database is that no distinction is made between events
181 where the numeric value of an attribute (e.g. fatalities or total damage) is not known and one in which
182 the value is 0. In both cases the field is left blank. So it is difficult to filter out flood events with 0
183 fatalities. Also for this reason only events with one or more reported fatalities are included in the
184 analysis in this paper.

185 **3 Results**

186 *3.1 Global flood fatalities and comparison to other disasters*

187 Floods are generally the most frequent event type associated with one or more fatalities, compared to
188 other natural disasters (Fig 1, left panel). The most frequently occurring flood types are riverine and

189 flash floods (Fig. 1, right panel).



190

191 **Fig. 1 Number of disaster events (left) and flood events (right) with one or more fatalities as recorded in**
192 **EM-DAT for the period 1975-2022**

193 Figure 1 (right panel) displays the number of reported flood events per year. Trendlines are included
194 to split the data into periods before and after the time that digital information became more widely
195 available. Assuming two linear trends, it has been found that the year 2000 corresponds to a splitting
196 point in which the combined regression error of those two trend lines is minimised. There is a strong
197 increase in the number of reported events until the year 2000 and increase in the yearly reported
198 number of events slows down afterwards (Fig. 1, right panel)

199 We have also analysed the trends for the annual number of reported events for various alternative
200 thresholds (more than 10, 100, 1000 fatalities) and results are reported in Fig. 7 in the appendix. For
201 the period after the year 2000, this shows that there is an increase in the smaller events (≥ 1 fatalities),
202 a stabilization for medium events (≥ 10) fatalities and a decrease of the annual number of large events
203 (≥ 100 fatalities).

204 The initial increase until the year 2000 is most likely due to extended and improved reporting and
205 availability of information online in the digital age (Tanoue et al. 2016; Jonkman 2005; Jones et al
206 2022). Here, we account for this trend by reporting average number of fatalities as well as mortality
207 fractions per event, besides the total annual numbers of fatalities and affected.

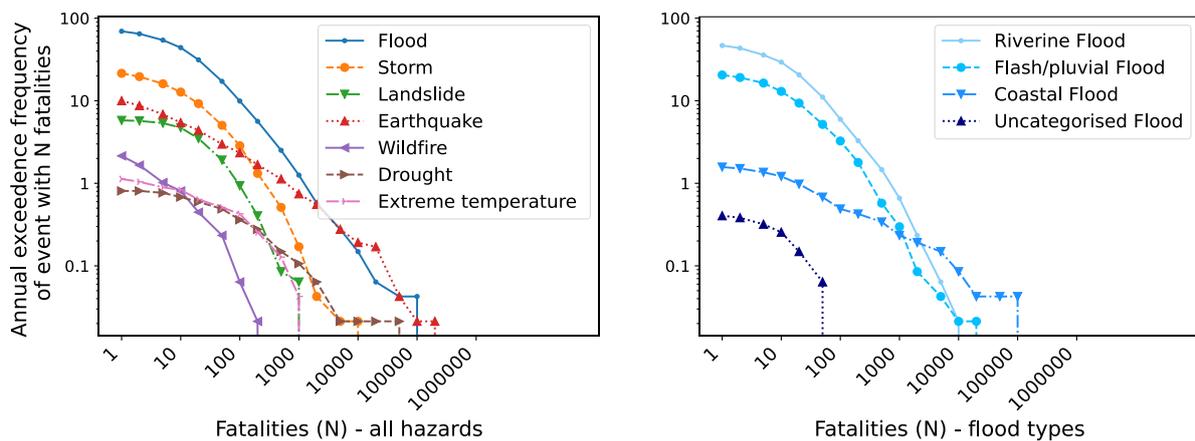
208 Table 2 summarizes the largest events in terms of fatalities in the database for various categories of
209 disasters. Note that some events fall into multiple categories. The events with largest numbers of
210 fatalities include the 1983 Sudan drought, the 1976 China earthquake, the Indian Ocean tsunami of
211 2004, and the Haiti earthquake in 2010. Some very deadly floods occurred in the form of coastal flood
212 events in Bangladesh (1991) and Myanmar (2008), and catastrophic river floods in China (1975).

213 Particularly due to these large events, it is found that most flood fatalities (around 85%) occurred in
214 Asia.

215 **Table 2 List of largest disaster events in database for various categories, ranked by total number of**
 216 **fatalities**

Year	Disaster Category	Disaster Subcategory	Country / Region	Fatalities	People Affected (mlns)	Direct Damage (10 ⁶ US\$)	Ranks		
							All events	Extreme Weather events	Flood events
1983	Drought	-	Sudan et al.	450,000	1.7	-	1	1	
1976	Earthquake		China	242,000	0.41	5,600	2		
2004	Earthquake	Geophysical (Tsunami)	Sri Lanka et al.	226,408	0.24	500	3		1
2010	Earthquake		Haiti	222,570	0.37	8,000	4		
1991	Flood	Coastal	Bangladesh	138,865	1.54	-	5	2	2
2008	Flood	Coastal	Myanmar	138,375	0.24	-		3	3
1981	Drought	-	Mozambique	100,000	0.47	-		4	
1999	Flood	Flash/pluvial	Venezuela	30,000	0.05	3,160		5	4
1975	Flood	Riverine Flood	China	20,000	1.1	-			5

217



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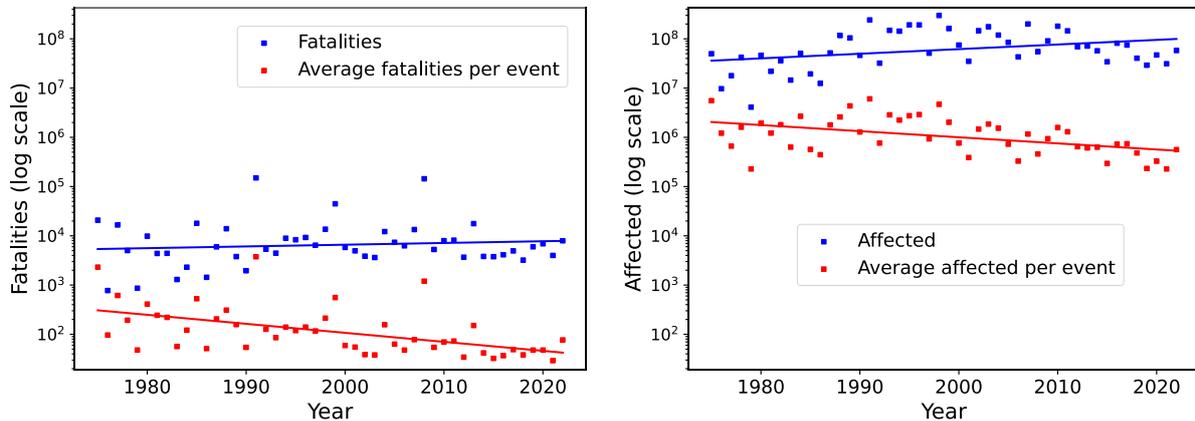
219 **Fig. 2 Frequency curves for disasters (left) and flood types (right) indicating the annual exceedance**
 220 **frequency of events with a certain number of fatalities**

221 Figure 2 (left panel) shows the average annual frequency of events with N or more fatalities for
 222 various natural hazards. What we observe from the left side of the figure is that floods are the most
 223 likely disaster amongst the geophysical and meteorological hazards to cause any given number of
 224 deaths up to 5000. Flood events with >100 deaths happen globally on average 10 times a year.

225 Riverine and flash floods are the most frequent flood type for flood events up to 1,000 fatalities (Fig.
 226 2, right panel). These flood types combine to cause many relatively smaller events per year: around 60
 227 events occur each year with 5 or more fatalities. Coastal floods from storms and tropical cyclones
 228 occur less frequently, but are dominant for events over 1,000 fatalities (see also Table 1). In the
 229 subsequent analysis, we focus on coastal and inland flood events with meteorological causes, as
 230 explained in the introduction.

231 3.2 Trends in flood fatalities and people affected

232 The trends for both the total global annual number of reported flood fatalities (left) and people
233 affected (right) are shown in blue in Figure 3, and appear to be static or slightly increasing. Using a
234 Mann Kendall test (alpha value of 0.05) it is shown that neither of these trends are statistically robust,
235 with the test giving p-values of 0.75 and 0.1, respectively.



236

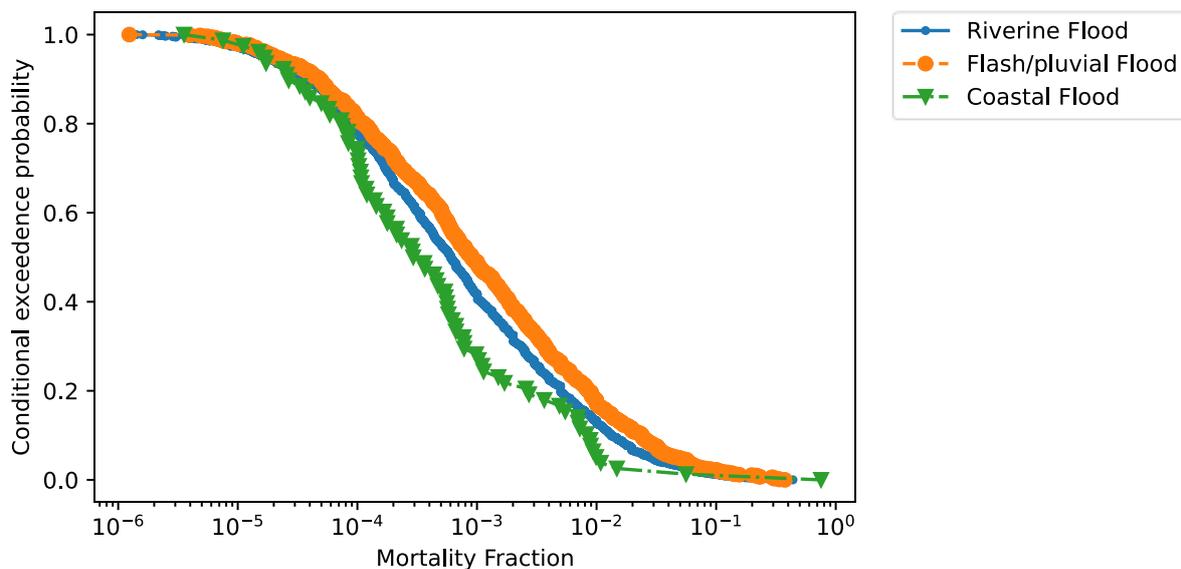
237 **Fig. 3 Trends in flood fatalities (left) and affected (right): yearly totals (blue) and averages per event (red)**

238 The average number of fatalities per event (left) and the average number of people affected per event
239 (right) in each year are shown in red in Figure 3, showing a statistically significant downward trend
240 (Mann-Kendall test p-values of 0.001 and 0.025, respectively). This reduction in the average impact
241 (both fatalities and people affected) per event can be attributed to a) changes in reporting over time
242 and b) improved flood risk management – as is further explained below.

243 A potential change in reporting is that more smaller events are getting reported for more recent
244 decades, leading to an overall decline over time in the average number of fatalities (and affected
245 people) per event. We have analysed the annual number of reported events (see above) and the
246 average number of killed and affected per event for various thresholds (more than 10, 100, 1000
247 fatalities). From the analysis it appears that the average number of fatalities and affected per event
248 decreases over time for all thresholds (figure 8 in the appendix). This shows that the downward trend
249 is robust. It is therefore expected that improved flood risk management practices over the last decades
250 will have contributed to the downward trend in fatalities per event. This includes improved
251 forecasting and flood early warning, increased protection in several places, and other forms of risk
252 reduction. These are global trends measured in terms of reduced average impacts. Reports from
253 individual events may clearly point out failures in risk management, e.g. lack of flood warning (also
254 see section 4).

255 3.3 Mortality fractions by flood type

256 Mortality fractions are used to indicate the severity in terms of lethality of natural hazard events and
257 are important indicators of vulnerability and risk. These fractions allow comparing how lethal
258 different events are in one location, or similar events between locations. These fractions are calculated
259 by dividing the number of fatalities by the exposed population for each flood event (see section 2.3).
260 Fig. 4 shows the conditional exceedance probability of the mortality fraction given that a certain event
261 type occurs. The data is lognormally distributed using a Kolmogorov-Smirnov test (p-value of 2.3e-
262 238). Flash floods generally lead to the highest event mortality fractions in any flood event, followed
263 by riverine floods (for values of the mortality fraction below 10^{-2}). This order roughly follows the
264 available early warning lead times, as flash floods flash floods generally occur with limited or no
265 warning and limited time to respond. Conversely, windstorms leading to coastal flooding, and
266 flooding in large river basins, can usually be forecasted multiple days in advance, with possibilities
267 for warning and evacuations. Although the average event mortality fraction is the lowest for coastal
268 floods, these events generally affect larger areas and populations than other floods, thus leading to
269 higher overall impact. This is also illustrated by the fact that the right tail in the frequency – fatalities
270 diagram in Fig. 2 (right panel) is dominated by coastal floods.



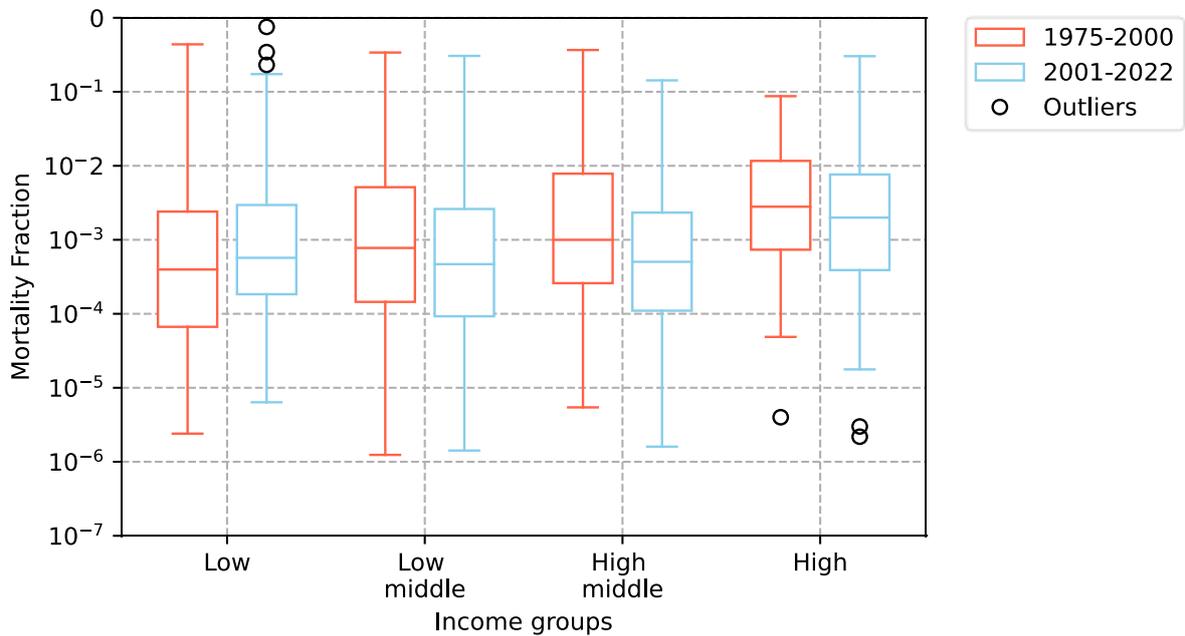
271

272 **Fig. 4 Conditional exceedance probability of mortality fractions given the occurrence of a certain flood**
273 **type**

274 3.4 Trend in mortality fractions by income group

275 We analyse temporal changes in mortality fractions, by comparing the periods 1975 – 2000 and 2001
276 – 2022, for different countries grouped by income. There is no significant reduction in total mortality
277 between both periods for the entire population of events. However, using one-way ANOVA tests, we
278 observe that a statistically significant reduction in mortality fractions is experienced by middle and

279 high middle-income countries (Fig. 5). Strikingly, low-income countries show slightly increased
 280 mortality fraction values. The change in event mortality for the high-income group is not statistically
 281 significant.



282

283 **Fig. 5** Boxplots showing flood mortality fraction distribution by income groups (according to World Bank
 284 classification) and temporal period. The boxplots show the median as a straight line, interquartile ranges
 285 as boxes, and ‘whiskers’ showing the limit of what is considered to be within the estimated distribution
 286 (i.e. not an outlier). Outliers are events that fall at a distance of more than twice the inter-quartile range
 287 from the median

288 4 Discussion and conclusions

289 In our study, we find that the number of flood events has increased over time with a slower overall
 290 increase since 2000, and a decrease of the annual number of large events (≥ 100 fatalities) since that
 291 year. The average number of people killed and affected per event has decreased over time. Since
 292 1975, year-on year declines of roughly 6 fatalities and over 30,000 persons affected per event are
 293 observed.

294 A key question is which drivers have caused these trends. Increased reporting and effects of climate
 295 change would lead to increasing numbers of flood events. Climate change would lead to more
 296 frequent flood events and higher intensities (IPCC, 2023). Population and exposure growth would
 297 mainly result in increasing event impacts. Measures to reduce vulnerability, e.g. better warning and
 298 shelters, will lead to a reduction in the mortality fraction. Protective measures, such as flood defences,
 299 will prevent flood events or lower exposure by protecting parts of the population. Since all these
 300 factors change over time, mortality fractions and average impacts per event are considered key
 301 indicators for tracking relative impact of flood events. These indicators can be used to compare trends

302 in different countries, regions and income groups, and are more time-consistent and robust metrics
303 than total annual fatalities.

304 Note that the above effects are not fully independent: climate change could lead to stronger floods and
305 higher impacts at the event level, but local adaptation through increased dike height and strength
306 could prevent disasters in those locations.

307 It is striking that the total number of annual flood fatalities is rather constant over time, as there have
308 been large changes in exposure in the past decades. Over the considered period the global population
309 has grown substantially from 4 to almost 8 billion people (source:
310 <https://data.worldbank.org/indicator/SP.POP.TOTL>, accessed October 31 2023). Substantial part of
311 this growth has taken place in flood prone areas along coasts and in river basins, leading to increases
312 in the exposed population (Tellman et al. 2021; Rentschler et al. 2023).

313 A key cause for the resulting downward temporal trend in the average number of people killed per
314 event concerns improved flood risk management practices over the last decades. This includes
315 increased protection, better warning, forecasting and early warning communication and other forms of
316 risk reduction. These are general trends, reports from recent individual events clearly point out
317 failures in risk management. For example for the deadly summer 2021 floods that affected multiple
318 countries in Northwestern Europe most fatalities (more than 180) occurred in Germany, but flood
319 warnings failed to reach most of the affected residents, and many were not prepared to act on the
320 warning information (Thieken et al. 2023). At the same time there are successful examples: in
321 Bangladesh, improved forecasting, early warning and cyclone shelters, as well as improved coastal
322 protection has led to progressively decreasing fatalities in similar cyclone-related coastal flooding
323 (Paul 2009; Bouwer and Jonkman 2018). Also in the western world, improved flood protection for
324 instance in the city of Hamburg after the 1962 flood disaster has avoided loss of life in more severe
325 storm surge events in the years since (Kron and Müller 2019). Also, the New Orleans hurricane
326 protection system that was built after the catastrophic flooding due to hurricane Katrina (2005) has
327 prevented flooding and loss of life during subsequent hurricanes, such as Isaac in 2012.

328 Our results confirm the finding that mortality from flood events has declined over time (Jongman et
329 al. 2015; Tanoue et al. 2016; Formetta and Feyen 2019) and also show that average fatalities per event
330 have declined for all flood types, and not just for large-scale coastal flood events (Bouwer and
331 Jonkman 2018). However, we did not find statistically significant declining trends in the total annual
332 number of people killed or affected per year on a global scale. The reduction of average impacts per
333 event seems to be “compensated” by an increase over time of the number of reported events (Fig. 1)
334 and the increased exposure due to population growth (see above). The trends in flood fatalities (a
335 stabilization of the annual total, and a decline in average fatalities per event) are opposed to trends in
336 damages from weather-related extreme events (including floods) which have increased (Coronese et

337 al. 2019). However, this damage increase is largely caused by increasing exposure of capital in urban
338 areas (Bouwer 2019; Geiger and Stomper 2020).

339 Floods have become less lethal over time in many regions. Especially middle-income countries have
340 succeeded in reducing mortality from flooding, while low-income countries have witnessed an
341 increase after the year 2000 (Fig. 5). For high income groups we find no trend. In both cases, potential
342 explanations may be the increased exposure in the floodplains. In low-income countries, people are
343 increasingly moving into floodplains due to lack of available options without the implementation of
344 substantial risk reduction. In middle-income countries, increasing financial and other resources to
345 protect and warn populations would lead to reduced vulnerabilities. This is different from the inverted
346 U-shape in vulnerability, found in other studies for total mortality (Tanoue et al. 2016; Kellenberg and
347 Mobarak 2018). Here we demonstrate the relationship between income level and mortality for average
348 mortality fractions per event, which is a more robust indicator. The difference with other studies could
349 be due to the fact that here we have recorded the mortality for the income group at the time of the
350 event, so countries move from one income group to another over time. This accurately reflects the risk
351 of the income level group at that time.

352 This study complements an earlier study (Jonkman 2005) on global patterns in loss of life, which
353 considered the period 1975-2002. It extends the dataset beyond 2002 with 20 years of data, and adds
354 information on coastal floods for all years. While the 2005 study suggested a slight increase in the
355 number of killed and affected per event, the present study shows a declining trend when the longer
356 period 1975-2022 is considered. The 2005 study also highlighted a growth in the number of reported
357 events up until 2002. The analyses in the present article show that an increase in reporting through
358 wider availability of media and internet resources has played a role in the initial increase in reported
359 events.

360 Our analysis is based the publicly available global disaster database EM-DAT. A number of main
361 issues included the consistency in reporting, the classification of combined coastal windstorm and
362 flood events and the effects of data collection and reporting efforts on observed upward global
363 temporal trends. Tracking global trends in vulnerability and disaster mortality – particularly in line
364 with the goals of the Sendai Framework – requires a consistent data basis over time. It is thus
365 important to minimise and correct for temporal trends in the analysis of disaster mortality statistics.
366 Also, the use of the time-consistent indicators from this study (event mortality and average number of
367 fatalities) can be utilized to monitor progress towards reducing risk and impacts from flooding and
368 other disasters.

369 **5 Data availability**

370 The dataset used for the analyses in this paper is made available in the 4TU Research Data repository
371 and can be found under the citation (Curran et al. 2023).

372

373 **6 References**

- 374 Alfieri L, Dottori F, Salamon P, Wu H, Feyen L (2020) Global Modeling of Seasonal Mortality
375 Rates From River Floods. *Earth's Future*, <https://doi.org/10.1029/2020EF001541>
- 376 Bouwer LM (2019) Observed and projected impacts from extreme weather events: implications for
377 Loss and Damage. In: Mechler, R., Bouwer, L.M., Schinko, T., Surminski, S. & Linnerooth-Bayer,
378 J. (eds.) *Loss and Damage from Climate Change: Concepts, Principles and Policy Options*,
379 Springer, 63-82
- 380 Bouwer LM, Jonkman SN (2018) Global mortality from storm surges is decreasing. *Environmental*
381 *Research Letters*, 13(1), 014008
- 382 Chen B, Shi F, Lin T, Shi P, Zheng J (2020) Intensive Versus Extensive Events? Insights from
383 Cumulative Flood-Induced Mortality Over the Globe, 1976–2016. *Int J Disaster Risk Sci* 11, 441–
384 451 <https://doi.org/10.1007/s13753-020-00288-5>
- 385 Coronese M, Lamperti F, Keller K, Roventini A (2019) Evidence for sharp increase in the economic
386 damages of extreme natural disasters. *Proceedings of the National Academy of Sciences* 116(43),
387 21450-21455
- 388 Curran A, Bouwer LM, Jonkman SN (2023) EMDAT disaster data used for article 'Floods have
389 become less deadly: an analysis of global flood fatalities 1975-2022. 4TU.ResearchData. doi:
390 10.4121/eba1143f-adbe-4038-bac9-dae804a8b65a.v1
- 391 Formetta G, Feyen L (2019) Empirical evidence of declining global vulnerability to climate-related
392 hazards, *Global Environmental Change*, Volume 57, 101920,
393 <https://doi.org/10.1016/j.gloenvcha.2019.05.004>.
- 394 Geiger T, Stomper A (2020) Rising economic damages of natural disasters: Trends in event intensity
395 or capital intensity?. *Proceedings of the National Academy of Sciences* 117(12), 6312-6313
- 396 Guha-Sapir D, D'Aoust O, Hoyois P (2013) The frequency and impact of natural disasters. In:
397 Guha-Sapir, D., Santos, I., Borde A. *The economics of natural disasters*. Oxford University Press
- 398 Hu P, Zhang Q, Shi P, Chen B, Fang J (2018) Flood-induced mortality across the globe:
399 Spatiotemporal pattern and influencing factors. *Science of the Total Environment* 643, pp.171-182

400 Huggel C, Bouwer LM, Juhola S, Mechler R, Muccione V, Orlove B, Wallimann-Helmer I (2022)
401 The existential risk space of climate change. *Climatic Change* 174, 8

402 Intergovernmental Panel on Climate Change (IPCC) (2023) *Climate Change 2023: Synthesis*
403 *Report*. In: *Climate Change 2023: Synthesis Report. Contribution of Working Groups I, II and III to*
404 *the Sixth Assessment Report of the Intergovernmental Panel on Climate Change [Core Writing*
405 *Team, H. Lee and J. Romero (eds.)*. IPCC, Geneva, Switzerland, pp. 1-34

406 IRDR (Integrated Research on Disaster Risk) (2014) *Peril Classification and Hazard Glossary*
407 (IRDR DATA Publication No. 1). Beijing: Integrated Research on Disaster Risk

408 Jones RL, Guha-Sapir D, Tubeuf S (2022) Human and economic impacts of natural disasters: can
409 we trust the global data?. *Scientific Data* 9, 572

410 Jongman B, Winsemius HC, Aerts JC, Coughlan de Perez E, Van Aalst MK, Kron W, Ward, PJ
411 (2015) Declining vulnerability to river floods and the global benefits of adaptation. *Proceedings of*
412 *the National Academy of Sciences*, 112(18), pp.E2271-E2280

413 Jonkman SN (2005) Global Perspectives on Loss of Human Life Caused by Floods. *Natural Hazards*
414 34, 151–175

415 Kellenberg DK, Mobarak AM (2008) Does rising income increase or decrease damage risk from
416 natural disasters? *Journal of urban economics* 63(3), pp.788-802

417 Kreibich H, Van Loon AF, Schröter K, Ward PJ, Mazzoleni M, Sairam N, ... & Di Baldassarre G
418 (2022) The challenge of unprecedented floods and droughts in risk management. *Nature* 608(7921),
419 80-86

420 Kron W, Müller O (2019) Efficiency of flood protection measures: selected examples. *Water Policy*,
421 21(3), pp.449-467

422 Paul BK (2009) Why relatively fewer people died? The case of Bangladesh's Cyclone Sidr. *Natural*
423 *Hazards* 50, 289–304 (2009). <https://doi.org/10.1007/s11069-008-9340-5>

424 Pielke R (2021) Economic 'normalisation' of disaster losses 1998–2020: a literature review and
425 assessment. *Environmental Hazards* 20(2), 93-111

426 Rappaport EN (2014) Fatalities in the United States from Atlantic tropical cyclones: new data and
427 interpretation. *Bulletin of the American Meteorological Society* 95, 341–346

428 Rentschler J, Avner P, Marconcini M et al. (2023) Global evidence of rapid urban growth in flood
429 zones since 1985. *Nature* 622, 87–92 <https://doi.org/10.1038/s41586-023-06468-9>

430 Tanoue M, Hirabayashi Y, Ikeuchi H (2016) Global-scale river flood vulnerability in the last 50
431 years. *Scientific Reports* 6, 36021

432 Tellman B, Sullivan JA, Kuhn C, Kettner AJ, Doyle CS, Brakenridge GR, Erickson TA, Slayback
433 DA (2021) Satellite imaging reveals increased proportion of population exposed to floods. *Nature*
434 596, 80–86

435 Thielen AH, Bubeck P, Heidenreich A, von Keyserlingk J, Dillenardt L, Otto A (2023)
436 Performance of the flood warning system in Germany in July 2021 – insights from affected
437 residents. *Natural Hazards Earth System Sciences* 23, 973–990

438 United Nations Office for Disaster Risk Reduction (UNDRR) (2015). Sendai framework for disaster
439 risk reduction 2015–2030. In: *Proceedings of the 3rd United Nations World Conference on Disaster*
440 *Risk Reduction (WCDRR)*, Sendai, Japan, 14–18 March 2015; pp. 14–18

441 World Economic Forum (2023) *The Global Risks Report 2023 18th Edition*. World Economic
442 Forum

443 Zhang J, Xu W, Liao X, Zong S, Liu B (2021) Global mortality risk assessment from river flooding
444 under climate change. *Environmental Research Letters*, Volume 16, Number 6; DOI 10.1088/1748-
445 9326/abff87

446 **7 Statements and declarations**

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452 **8 Competing interests**

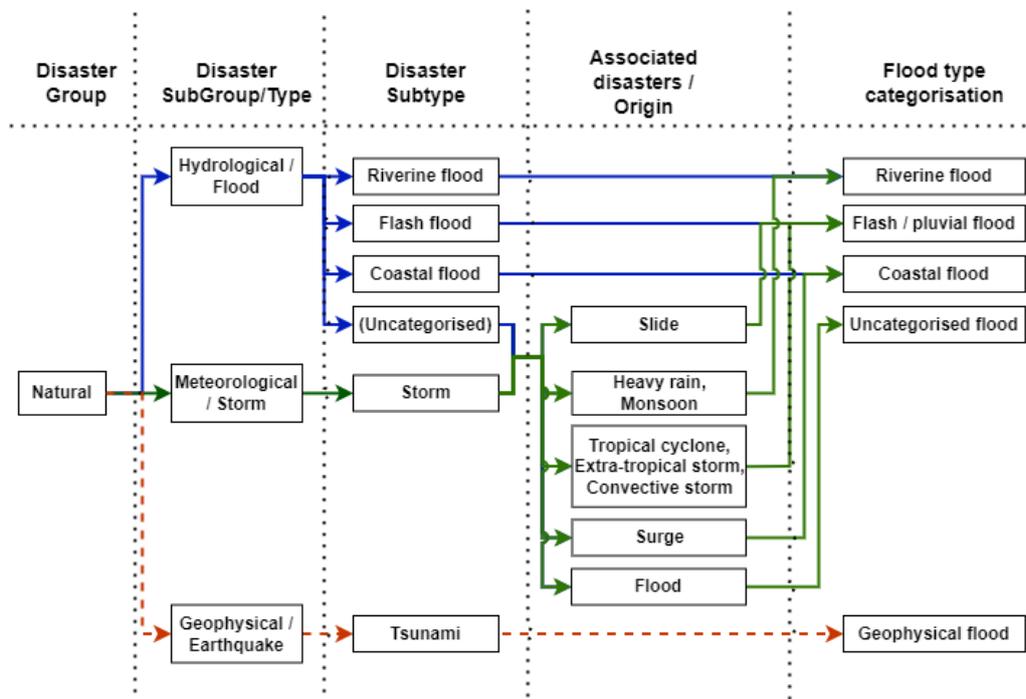
453 The authors have no relevant financial or non-financial interests to disclose.

454 **9 Appendix**

455 **Table 3 Manually reclassified events that have been added to the flood fatalities database**

EMDAT code (incl. year)	Countr(ies)	Storm name	Reclassified as:	Fatalities	Affected (mln)
1979-0070	Dominican Republic	David / Frederick	Flash	1400	1.51
1984-0105	Philippines	Agnes / Undang	Coastal	1079	2.26
1984-0185	Philippines	June / Maring	Flash	1399	1.78
1997-0267	Vietnam / Thailand	Linda	Coastal	3859	1.08
1998-0345	Nicaragua	Mitch	Flash	14609	2.12
2005-0567	Guatemala	Stan	Coastal	1074	2.52

456



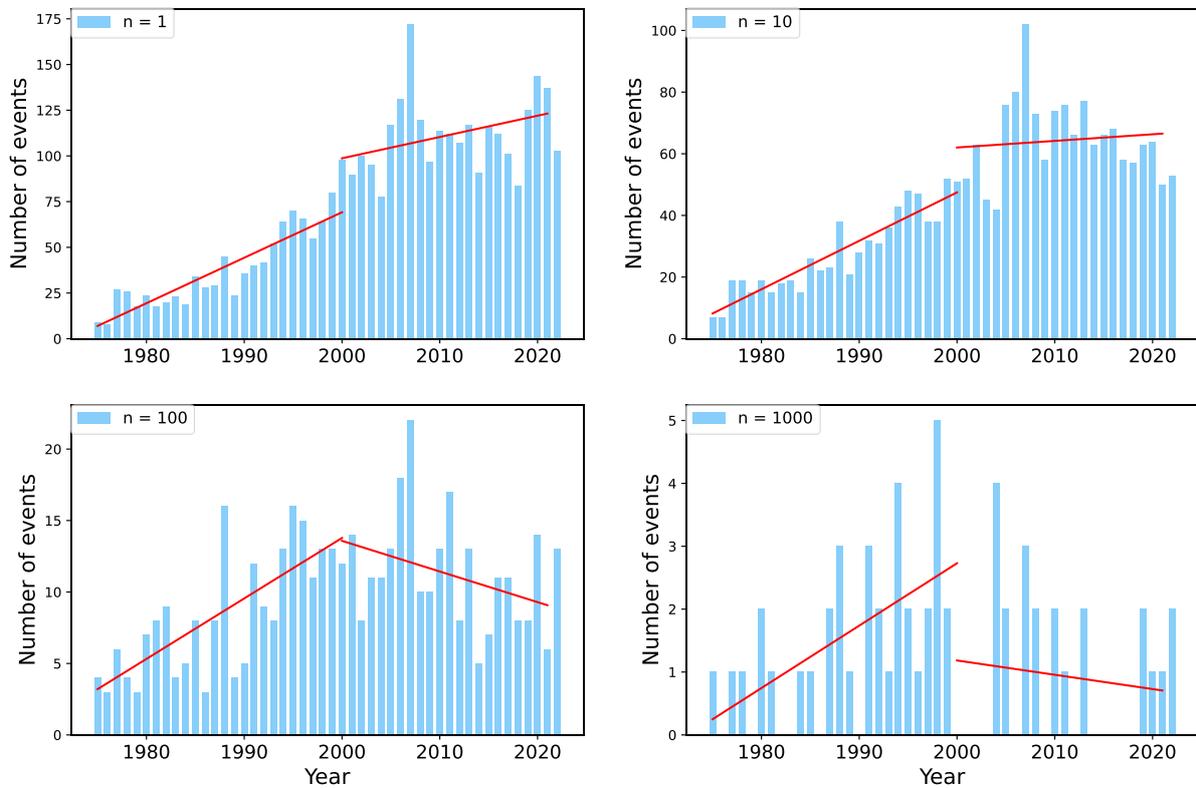
457

458 **Fig. 6 Categorisation procedure for flood events based on EMDAT attributes. The first four columns**
 459 **show the categories from EM-DAT, the last column shows the flood type categories used in this paper.**
 460 **Tsunami events can also be retrieved from the database, but are not included in the present study**

461 We have analysed the total number of events per year for various thresholds ($\geq 1, 10, 100, 1000$ fatalities
 462 respectively) (fig. 7). For all categories there is a strong increase until the year 2000, with the most
 463 significant absolute increase for events with 1 or more fatalities. After the year 2000, there is still an
 464 increase in the smaller events (≥ 1 fatalities), a stabilization for medium events (≥ 10) fatalities and a

465 decrease of the annual number of large events (≥ 100 fatalities).

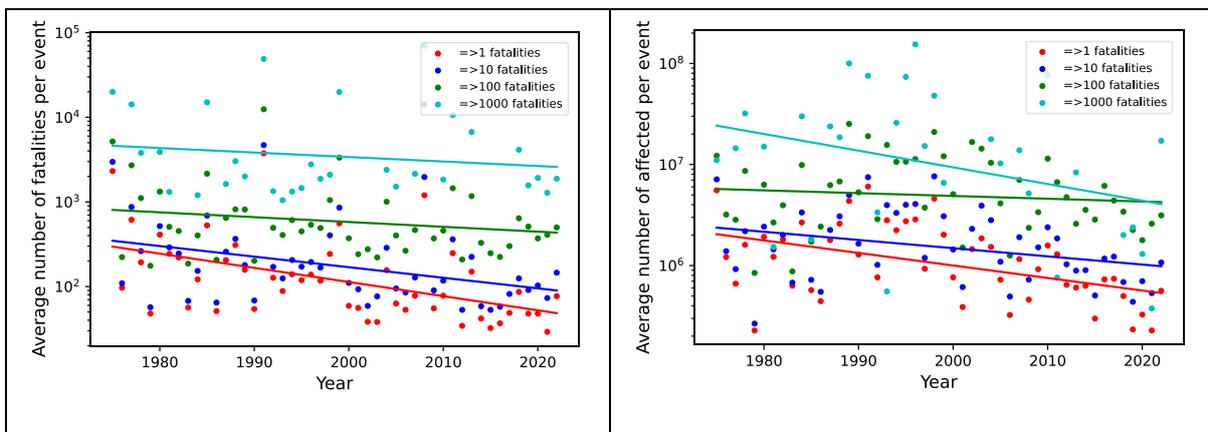
Number of flood events with n or more deaths



466

467 **Fig. 7** Reported number of fatal flood events per year and trendlines for various thresholds. Note that the
 468 vertical axes are not at the same scale.

469 We have analysed temporal trends in average event impact for various thresholds (1,10,100, 1000
 470 fatalities). Figure 8 shows the average number of fatalities per event (left) and affected (right) for a
 471 given year. For all categories a downward trend is found.



472 **Fig. 8** Average number of fatalities per event (left) and average number of affected per event (right) and
 473 trendlines