



1 **More frequent flooding? Changes in flood frequency in Pearl River**
2 **basin, China since 1951 and over the past 1000 years**

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17 **Abstract:** Flood risks across the Pearl River basin, China, were evaluated using peak
18 flood flow dataset covering a period of 1951-2014 from 78 stations and historical flood
19 records of recent 1000 years. The General Extreme Value (GEV) model and the kernel
20 estimation method were used to evaluate frequencies and risks of hazardous flood
21 events. Results indicated that: (1) no abrupt changes or significant trends could be
22 detected in peak flood flow series at most of the stations, and only 16 out of 78 stations
23 exhibited significant peak flood flow changes with change points around 1990. Peak
24 flood flow in the West River basin was increasing and significant increasing trends were
25 identified during 1981-2010; decreasing peak flood flow was found in coastal regions



26 and significant trends were observed during 1951-2014 and 1966-2014; (2) the largest
27 three flood events were found to cluster in both space and time. Generally, basin-scale
28 flood hazards can be expected in the West and North River basins; (3) the occurrence
29 rate of floods was increasing in middle Pearl River basin but decreasing in the lower
30 Pearl River basin. However, hazardous flood events were observed in the middle and
31 lower Pearl River basin, and it is particularly true in recent 100 years. However,
32 precipitation extremes were subject to moderate variations and human activities, such
33 as building of levees, channelization of river systems and rapid urbanizations were the
34 factors behind the amplification of floods in the middle and lower Pearl River basin,
35 posing serious challenges for developing measures of mitigation of flood hazards in the
36 lower Pearl River basin, particularly the Pearl River Delta (PRD) region.

37

38 **Key words:** Flood frequency; Flood risk; GEV model; Kernel estimation

39

40 **1. Introduction**

41 Climatic extremes are one of the crucial drivers of meteorological and hydrological
42 hazards, such as floods and droughts (IPCC, 2007; Li et al., 2016). Meanwhile, climate
43 change is expected to intensify the global hydrological cycle which would potentially
44 lead to a general increase in the intensity and frequency of extreme climatic events
45 (Ohmura and Wild, 2002; Alan et al., 2003; Zhang et al., 2013). This will, in turn, have
46 direct implications for hydrological extremes, such as floods and droughts (IPCC, 2013).
47 However, the impacts of climate change on hydrological extremes are expected to vary
48 across different regions over the globe due to the prevailing hydrometeorological
49 regimes and the nature of climate change in specific regions (Beniston and Stephenson,
50 2004; Burn et al., 2015).



51 Due to remarkable differences in the hydrometeorological processes that generate
52 floods, climate change can increase or decrease the magnitude, duration, frequency and
53 even nonstationarity of extreme hydrological events, such as floods considered in this
54 study (Gu et al., 2016; Vormoor et al., 2016; Zhang et al., 2016). A very recent study
55 by Zhang et al. (2015a) corroborated the changes in hydrological extremes across China
56 but also found the increasing impact of human activities on fluvial hydrological
57 processes. Changes in hydrometeorological triggers are believed to be the first step to
58 project likely future changes in flood generating processes (Hall et al., 2014; Vormoor
59 et al., 2016). It is particularly the case with flood processes in the Pearl River basin, the
60 second largest river basin in China in terms of flow volume with highly-developed
61 economy, dense population and important megacities, such as Guangdong, Macau and
62 Hong Kong. This constituted the motivation for this study. Hydrometeorological
63 extremes often have disastrous impacts on the society, water resources, agricultural
64 activities, urban infrastructure, and also ecosystems (Das et al., 2013; Li et al., 2016).
65 Floods in particular damage infrastructure, take away many lives and are one of the
66 costliest types of natural disaster in economic and human terms (Bouwer and Vellinga,
67 2003). It is also true for China where floods tend to have more significant impacts on
68 agriculture than droughts (Zhang et al., 2015b).

69 Increasing catastrophic losses due to natural hazards have aroused widespread
70 public awareness of extreme events in recent years (e.g. Beniston and Stephenson, 2004;
71 Zolina et al., 2004). By 2100, the mean annual global surface temperature would
72 increase by 1.4-5.8°C, and future climatic and hydrological extremes would tend to
73 increase and intensify correspondingly (Houghton et al., 2001; Beniston et al., 2007;
74 IPCC, 2007). Therefore, it is important to investigate the flood behavior, and related



75 studies can be of practical value in water resources management. It should be noted that
76 precipitation extremes have a predominant effect on floods (Jena et al., 2014). Studies
77 on precipitation extremes across the Pearl River basin have indicated that the amount
78 of rainfall has changed little but the variability has increased over the time interval
79 divided by change points (Zhang et al., 2009). Further, changes in the characteristics of
80 precipitation extremes across the Pearl River basin are similar to those over the globe
81 (Hirsch and Archfield, 2015), i.e. frequencies of precipitation extremes are increasing
82 but magnitudes have moderate changes. However, increasing precipitation extremes are
83 observed mainly in the lower Pearl River basin, including the Pearl River Delta (PRD)
84 region (Zhang et al., 2012), and also partly in the middle Pearl River basin. Therefore,
85 it can be expected that flood risk should be higher in the middle and lower Pearl River
86 basin, or coastal regions (Pino et al., 2016).

87 In general, extreme floods are rare and hence there is limited opportunity to collect
88 adequate samples of such events in order to make reliable predictions. Therefore, the
89 question is how best to extrapolate to extreme events, when no or only short series of
90 such events are available (Kjeldsen et al., 2014). High quality data and analyses of long
91 historical records of peak extreme events are important to determine whether climate is
92 becoming extreme or variable (Nicholls, 1995). To that end, flood records of 1000
93 years from Guangdong province (which covers the lower Pearl River basin) and
94 Guangxi province (which covers the middle Pearl River basin) were collected to
95 overcome the limitations of short gauge station-based flood records for analyzing floods,
96 and this is also the significance of this study.



97 The historical flood records were collected from two books compiled by Wen and
98 Song (2006) and Wen and Yang (2007). These two books included abundant records
99 relevant to various meteorological disasters, such as tropical cyclones, droughts, floods,
100 frosts, and so on. The historical flood records should be screened out from those
101 abundant records and it is a kind of time-consuming job. Therefore, few reports were
102 found concerning flooding changes over long period such as 1000 years in this study
103 based on historical records. Besides, historical flood records in other regions of the
104 globe have been used to analyze the changes in flood frequency (e.g. Mudelsee et al.,
105 2003). Mudelsee et al. (2003) collected flood records from A.D. 1000 to A.D. 2000,
106 and pointed out no upward trends in the occurrence of extreme floods in central Europe.
107 Mudelsee et al. (2004) indicated that the historical flood records can provide reliable
108 and unique information on heavy floods at least since A.D. 1500 in the Elbe and Oder
109 rivers, which further emphasized merits of historical records in the study of flooding
110 frequency from a long-term perspective.

111 Therefore, the objectives of this study are: (1) to quantify abrupt changes and trends
112 of flood events; (2) to characterize temporal changes of 10 year flood flow and spatial
113 distribution of flood magnitude > 10-year flood magnitude (the flood peak is expected
114 to occur, on average, once every 10 years); and (3) to determine frequency and
115 occurrence rate based on 1000 year flood records. Potential causes of spatiotemporal
116 patterns of floods across the Pearl River basin and related implications are also
117 discussed. This study would provide a clear picture showing the evolution of floods in
118 both space and time in a humid river basin and show the response of hydrological



119 extremes to climate change and human activities.

120

121 **2. Study region and data**

122 2.1 Study region

123 The Pearl River (97°39'E- 117°18'E; 3°41'N- 29°15'N) (Fig. 1), with a drainage
124 area of $4.42 \times 10^5 \text{ km}^2$, is the second largest river in China in terms of flow volume
125 (PRWRC, 1991). It involves three major tributaries: West River, North River and East
126 River. The West River (Region I) is the largest tributary, accounting for 77.8% of the
127 total drainage area of the basin. The North River (Region II) is the second largest one
128 with a drainage area of 46710 km². The East River (Region III) accounts for 6.6% of
129 the total area of the Pearl River. And the Region IV, which is beyond the three major
130 tributaries, locates in the west of the Guangdong province (Fig. 1). The annual mean
131 temperature ranges between 14-22°C and the precipitation mainly occurs during April-
132 September (Zhang et al., 2009), accounting for 72%-88% of the annual precipitation
133 (PRWRC, 1991). The Pearl River basin is covered mainly by two provinces, i.e.
134 Guangdong and Guangxi (Fig. 1b). Numerous water reservoirs have been built in
135 northern, eastern and western Guangdong and also central and southern Guangxi (Fig.
136 1b). In addition, widespread urbanization can be observed in the PRD, eastern
137 Guangdong and coastal regions of Guangdong (Chen et al., 2008) (Fig. 1b) that have a
138 highly-developed economy (Fig. 1c) and dense population settlements (Fig. 1d).
139 Central and southern Guangxi is dominated by croplands (Fig. 1b). The streamflow
140 variations of the Pearl River basin have a considerable influence on the hydrological



141 processes of the PRD, one of the most complicated deltaic drainage systems in the
142 world (Chen et al., 2008). Flat terrain at low-lying altitude and the downstream location,
143 together with rapid economic development and population growth over the past three
144 decades have made the PRD region more and more vulnerable to natural hazards, such
145 as flood, salinity intrusion, and storm surge. In recent years, engineering facilities and
146 other modifications of the Pearl River network have been designed to strengthen flood
147 protection and to cater for huge requirements of building materials.

148

149 2.2 Data

150 The annual largest 1 day streamflow data (i.e. annual maxima) were collected from
151 78 hydrological stations across the Pearl River basin. Locations of these hydrological
152 stations are shown in Fig. 1a. Besides, daily precipitation data were also collected from
153 74 stations across the Pearl River basin and their locations are shown in Fig. 1c. All the
154 precipitation and hydrological data cover the period of 1951-2014. Detailed information
155 of these hydrological and precipitation data can be found in Fig. 2. The hydrological
156 data were provided by the Hydrological Bureau of Guangdong province and the
157 precipitation data were collected from National Climate Center. The quality of these
158 data is firmly controlled before release.

159 Mudelsee et al. (2003) classified floods into three types, based on inundation area
160 and flood-induced losses: (1) floods that occur locally with short duration and small
161 damages; (2) regional floods that have relative longer duration and cause damages to
162 hydraulic infrastructure and also casualties; and (3) fluvial disastrous floods with long



163 lasting duration (usually days or weeks) causing serious and even disastrous damages
164 to hydraulic infrastructure and massive casualties. In this study, historical flood records
165 were collected from documented flood records compiled by Wen and Song (2006) and
166 by Wen and Yang (2007). The documented flood records for Guangdong and Guangxi
167 provinces covered a period of 383-2000 and 107-2000, respectively. The disasters were
168 recorded in history books, local chronicles, water conservancy archives, documents,
169 and so on. For the sake of the study on relations between climate change and disasters,
170 the group was developed to compile the documented nature disaster records spanning
171 almost 2000 years for each province in China based on multisource information. The
172 group selected the recorded flood events with mutual confirmation in different
173 documents as far as possible. In addition, the flood event with more relevant
174 information, such as magnitude, casualty rates, flood-damaged and flood-affected
175 farmland areas, flood-induced damaged water conservancy facilities, is more likely
176 selected. The director of the group is Wen who served as director of China
177 meteorological administration, and one of the group is Ding who is an academician of
178 the Chinese academy of sciences. The members of the group coming from senior
179 government authority and famous scientists, can largely ensure the quality of the data.
180 Based on flood types defined by Mudelsee et al. (2003), only disastrous flood events
181 were singled out, since floods occurred almost annually. Meanwhile, flood records
182 before 1000 AD were not complete and had missing information, thus disastrous flood
183 records during a period of 1000-2000 were singled out and analyzed in this study. One
184 flood event, which caused life losses or submersing more than 10 thousands areas of



185 farmland or destroying important water conservancy facilities, will be classified as
186 disastrous flood events.

187

188 **3. Methods**

189 3.1 Detection of change points and trends

190 The Pettitt method (Pettitt, 1979) is a nonparametric test and enables the detection
191 of change in the mean (median) when the change point time is unknown. This method
192 has been using widely in detection of change points (Villarini et al., 2009) and also used
193 in this study. The test is based on the Mann-Whitney statistic for testing whether the
194 two samples X_1, \dots, X_m and X_{m+1}, \dots, X_n come from the same population. The p value
195 of test statistic is computed using the limiting distribution approximated by Pettitt
196 (1979), which is valid for continuous variables (e.g. Villarini et al., 2009). The 95%
197 confidence level was used to evaluate the significance of change point in the study.

198 Trends were tested by non-parametric trend detection methods which are less
199 sensitive to outliers than are parametric statistics. In this study, the modified version of
200 the Mann-Kendall (MMK) trend test method was used which was proposed by Hamed
201 and Rao (1998) based on effective or Equivalent Sample Size (ESS) to eliminate the
202 effect of autocorrelation. MMK has been used in analyzing the effect of global warming
203 on small aquatic ecosystems (Daufresne et al., 2009). In this study MMK was employed
204 to explore trends in flood series, with the significance level set at 5%. For the
205 computation procedure one can refer to Daufresne et al. (2009).

206

207 3.2 Generalized Extreme Value (GEV) model

208 The GEV distribution has been widely used in the analysis of hydrometeorological



209 extremes (e.g. Gu et al., 2016) and has three parameters, i.e. the location μ , the scale, α
 210 ($\alpha > 0$), and the shape, κ . In this paper, GEV is used to calculate the return period of
 211 flood events. The cumulative density function (cdf) of a random variable y drawn from
 212 a GEV distribution is given as (Cannon, 2010):

$$213 \quad F(y; \mu, \alpha, \kappa) = \exp \left[- \left\{ 1 - \kappa \frac{y - \mu}{\alpha} \right\}^{1/\kappa} \right], \quad \kappa \neq 0, \quad 1 - \kappa \frac{y - \mu}{\alpha} > 0 \quad (1)$$

$$F(y; \mu, \alpha, \kappa) = \exp \left[- \exp \left\{ 1 - \frac{y - \mu}{\alpha} \right\} \right], \quad \kappa = 0$$

214

215

216

217 3.3 Kernel density estimation of occurrence rates of floods

218 The Kernel density estimation method is used to estimate the occurrence rates of
 219 historical floods. The estimation of occurrence rates of time-dependent extreme events
 220 can be computed as (Mudelsee et al., 2003; Mudelsee et al., 2004):

$$221 \quad \lambda(t) = h^{-1} \sum_{i=1}^m K \left(\frac{t - T_i}{h} \right) \quad (2)$$

222 where T_i is the timing of the i th flood event with unit in day; m is the number of floods;
 223 $K(\cdot)$ is the kernel function; and h is the width of the kernel function. The Gaussian
 224 kernel function is the widely-used kernel function, which can use the Fourier space and
 225 produce a smoothed estimation of the occurrence rates of extreme events (Mudelsee et
 226 al., 2003; Mudelsee et al., 2004):

$$227 \quad K(y) = \frac{1}{\sqrt{2\pi}} \exp \left(- \frac{y^2}{2} \right) \quad (3)$$

228 where $y = (t - T_i)/h$. The occurrence rate of an extreme event, $\lambda(t)$, denotes the number
 229 of an extreme event exceeding threshold values given a certain time interval, t . The time



230 interval of time series is $[t_1, t_m]$. Since no data are available outside of the time interval,
 231 i.e., $[t_1, t_m]$, $\lambda(t)$ near the boundaries of the time interval is usually underestimated. In
 232 this case, a kind of pseudodata is used to reduce the error as a result of underestimated
 233 $\lambda(t)$. A mapping technique is used to produce the pseudodata (Mudelsee et al., 2004).
 234 pT is the pseudodata outside of the time interval of $[t_1, t_m]$ for the flood series. For $t <$
 235 t_1 , $pT[i] = t_1 - [T_i - t_1]$; and the same procedure was done for $t > t_m$. The extended series
 236 is 1.5 times longer than the original one. The computation of $\lambda(t)$ based on the extended
 237 data series was based on:

$$238 \quad \lambda(t) = h^{-1} \sum_{i=1}^{m^*} K \left(\frac{t - T_i^*}{h} \right) \quad (4)$$

239 where T_i^* is the timing of the i th flood event based on the extended data series with a
 240 unit of day; m^* is the sample size of the extended data series. Also, the selection of
 241 window width, h , is important for the estimation of $\lambda(t)$. Too small window width, h ,
 242 selected for computation of $\lambda(t)$ will substantially influence the randomness on $\lambda(t)$; too
 243 large window width, h , may cause over smoothing of the data series and hence details
 244 in information may be excluded. The cross validation method was used to determine
 245 the width of window (Mudelsee et al., 2003).

246

247 3.4 Confidence interval by Bootstrap technique

248 The bootstrap technique and Equation (4) can be combined to enable uncertainty
 249 analysis of the occurrence rate of floods, $\lambda(t)$, using the following procedure (Mudelsee
 250 et al., 2004):

- 251 (1) Based on the extended data series, T_i^* , the simulated T^+ of the same series length
 252 can be obtained using the bootstrap technique;
- 253 (2) The occurrence rates, $\lambda^+(t)$, of sample extreme events, T^+ , can be computed using



254 Equation (5);
255 (3) Steps (2) and (3) above will be repeated for 2000 times, and $\lambda^+(t)$ of 2000 samples
256 can be obtained;
257 (4) The 90% confidence interval for $\lambda^+(t)$ will be obtained using the quantile method.

258

259 **4. Results**

260 4.1 Change points and trends of peak flood flow

261 Analyses of change points and trends were only applied in observed flood events
262 (i.e. annual maxima of period of 1951-2014). Fig. 3 illustrates spatial patterns of
263 stations with different change points of peak flood flow. It can be observed from the
264 figure that only 16 out of 78 stations, accounting for 20.5% of the total stations, were
265 characterized by significant change points of peak flood flow changes and most of these
266 stations are found in the middle and lower Pearl River basin. In the coastal regions of
267 the lower Pearl River basin, 10 out of 16 stations with significant change points were
268 observed, accounting for 62.5% of the total stations characterized by significant change
269 points of peak flood flow. Generally, flooding in the Pearl River basin is mainly
270 attributed to precipitation extremes which were observed mainly in the middle and
271 lower Pearl River basin and particularly in the lower Pearl River basin (Zhang et al.,
272 2012). Results of change points of precipitation maxima by Zhang et al. (2009)
273 indicated that precipitation maxima were dominated by significant change points during
274 1980-1993, and significant change points of peak flood flow series detected in this study
275 were during 1986-1995, showing significant impacts of precipitation extremes on
276 flooding. These results showed that significant change points of flood processes were



277 found mainly in the middle Pearl River basin and particularly in the lower Pearl River
278 basin. Besides, human impacts on flood processes cannot be ignored and it is
279 particularly the case for the East River basin (Zhang et al., 2015c), where 3 large water
280 reservoirs were built that controlled 11700 km² of drainage area.

281 Significant increasing peak flood flow was observed mainly in northeastern West
282 River basin, mainstream of the Pearl River basin, northern North River basin,
283 southeastern West River basin, and southern North River. Significant decreasing peak
284 flood flow was observed mainly in southeastern West River, southern North River basin,
285 and also parts of rivers along the coastal regions of the lower Pearl River basin (Fig.
286 4a). Significant increasing precipitation extremes were found in northeastern West
287 River basin and northern North River basin, and significant decreasing precipitation
288 extremes were detected in East River basin. Hence, spatial patterns of peak flood flow
289 matched those of precipitation extremes (Figs. 4a, 4b), implying that floods in these
290 regions were impacted mainly by precipitation extremes. The southeastern West River
291 basin and rivers in the west parts of region III were dominated by significant increasing
292 precipitation extremes but significant decreasing peak flood flow (Fig. 4). Human
293 activities exerted considerable impacts on flood processes in these regions. Crop land
294 in the Guangxi province was found mainly in the southeastern Jiangxi province (Fig.
295 1b) and irrigated cropland had a significant increase (Zhang et al., 2015b). The volume
296 of water withdrawal for agricultural irrigation during 2014 only reached $2.09 \times 10^{10} \text{m}^3$.
297 Meanwhile, the total water storage capacity of water reservoirs of the Guangxi Province
298 reached $6.74 \times 10^{10} \text{m}^3$, and more than half of the reservoirs were built in the southeastern



299 West River basin (Fig. 1b). The western parts of the region III were also dominated by
300 croplands and large-scale reservoirs (Fig. 1b), and agricultural water consumption
301 reached $2.24 \times 10^{10} \text{ m}^3$, and the total storage capacity of reservoirs reached $4.48 \times 10^{10} \text{ m}^3$
302 during 2014. These human activities greatly decreased peak flood flow volume in these
303 regions. Therefore, increasing human impacts on flood processes should arouse
304 considerable concerns for the management of water resources and mitigation of flood
305 hazards (Zhang et al., 2015a).

306 To determine trends of peak flood flow during specific time intervals, multi-scale
307 trend analysis was done (Fig. 5). Trends were identified by changing the time interval
308 by shifting the beginning and ending time of the interval with a time step of 5 years.
309 The shortest time interval was 15 years to ensure the validity of statistical analysis.
310 Besides, the percentages of stations with significant trends were also analyzed (Fig. 5).
311 The percentage of stations with significant trends in almost all time intervals, was
312 relatively low, being about 15% and even lower. However, stations with significant
313 decreasing trends of peak flood flow during 1966-2005 accounted for 20-25% of the
314 total stations considered in the study (Figs. 5). Significant increasing peak flood flow
315 was identified during 1981-2010 in the West River basin, and stations with significant
316 increasing peak flood flow during 1981-2010 accounted for 25-35% of the total stations.
317 Stations with significant decreasing peak flood flow during 1966-1990 accounted for
318 25-30% of the total stations (Figs. 5). Peak flows in the North River basin had moderate
319 changes without statistically detectable trends (Figs. 5). Significant decreasing peak
320 flood flow can be observed at the stations in the East River basin or eastern parts of the
321 region III, and stations with significant decreasing peak flood flow during 1951-1980
322 and 1951-2014 accounted for 20-25% and 25-30% of the total stations (Figs. 5).



323 Stations with significant decreasing peak flood flow were fewer after 1981 (Fig. 5),
324 implying amplifying flooding regimes after 1981 in the eastern parts of the region III.
325 Larger changing variability of peak flood flow in the western parts of the region III
326 were observed. Stations with significant increasing peak flood flow during 1951-1975
327 accounted for 30-35% and less stations were characterized by significant increasing
328 peak flood flow after 1966, and peak flood flow after 1966 turned out to be significantly
329 decreasing after 1966 (Figs. 5), stations with significant decreasing peak flood flow
330 accounted for 35-50% of the total stations.

331

332 4.2 GEV-based flood frequency

333 The GEV model was used to fit peak flood flow series (i.e. annual maxima of period
334 of 1951-2014) and the Kolmogorov Smirnov (K-S) statistic D was used to evaluate the
335 goodness-of-fit of GEV-based fitting performance (Fig. 6). Fig. 6 indicated that peak
336 flood flow series at almost all stations, except two stations, was well modelled by GEV
337 at 0.05 significance level. The peak flood flow series at these two stations were also
338 modelled by GEV at the 0.1 significance level. Therefore, GEV was used for flood
339 frequency analysis across the Pearl River basin. Return periods of floods at all
340 hydrological stations were estimated and spatial patterns were also characterized across
341 the basin. It can be observed from Fig. 7 that floods of > 10-year flood magnitude
342 occurred with high frequency. However, large floods occurred in a clustering manner at
343 the annual time scale. The three largest floods occurred mainly during two time
344 intervals, i.e. 1965-1970 and 1993-2002. There occurred 38 out of 66 three largest
345 floods at 22 stations during these two time intervals, i.e. 1965-1970 and 1993-2002
346 (region I in Fig. 7). There was no temporal clustering observed for the three largest
347 floods that occurred in the North River basin, but the spatial concentration was



348 identified (region II in Fig. 7). Taking the great floods that occurred during May 1982
349 across the entire North River basin as an example, flood events of > 10 year flood were
350 observed at 9 out of 20 stations and the measured largest three floods were observed at
351 7 stations.

352 Survey of floods across the West and North River basins indicated that higher
353 probability was expected for the simultaneous occurrence of floods in both North and
354 West River basins. For example, 1994 was a serious flooding year and floods that
355 occurred in the West River basin at 11 stations were larger than 10 year flood and 7 out
356 of 11 stations were dominated by the largest three floods of the recorded floods.
357 Meanwhile, peak flood flows at 8 stations in the North River basin were larger than 10
358 year flood and 6 out of 8 stations were dominated by the recorded largest three floods.
359 The occurrence of large floods in the east parts of region III was evidently uneven in
360 time and the measured largest three floods occurred mainly during 1958-1969, and
361 decreased occurrence rates was observed for large floods after 2005. However, floods
362 of > 10 year flood were amplifying after 2005. Besides, the occurrence of large floods
363 in the east parts of the region III was also subject to spatial clustering, and floods of >
364 10 year flood were observed usually at numerous stations at the same time (region III
365 in Fig. 7). Moreover, large floods occurred in a clustering manner during 1966-1974
366 and occurrence rates of large floods after 1980 exhibited moderate changes (region IV
367 in Fig. 7).

368 The percentage of stations with flood regimes of > 10 year flood to the total stations
369 for each region was counted and trends were evaluated by the 11-year moving average
370 method (Fig. 8), with the aim to determine the occurrences of large floods in both space
371 and time. It can be seen from Fig. 8a that the percentage of stations dominated by the
372 occurrence of large floods in West River basin had moderate changes with a slight



373 increasing tendency (Fig. 8a), and particularly after 1990. The percentage of stations
374 dominated by the occurrence of large floods had an increasing tendency and this
375 increasing tendency was maintained during the entire time interval considered in this
376 study (Fig. 8b). The percentage of stations with the occurrence of large floods followed
377 similar changing patterns, i.e. increase and then decrease, implying enhancing risks of
378 floods across the entire region (Figs. 8c and 8d).

379

380 4.3 Flood risks based on historical flood records

381 Based on historical flood records, the occurrence rates of floods during the last 1000
382 years in the Guangdong and Guangxi provinces were analyzed and local polynomial
383 regression fitting technique was used to smooth the series. It can be observed from Fig.
384 9a that the occurrences of floods had increasing trends before 1600 AD and reached the
385 peak value during about 1600 AD. The occurrences of floods had moderate variations
386 during 1600-1900 with moderate variability. Recent 100 years, i.e. 1900-2000, however,
387 witnessed decreasing occurrence rates of floods (Fig. 9a). These results would be
388 further evaluated using the kernel density estimation method in the next section.
389 However, the occurrence of floods in Guangxi province told another story when
390 compared to those in Guangdong province (Fig. 9b). A moderate increasing tendency
391 of occurrence rates of floods was observed before 1800 AD and the time interval after
392 1800 AD witnessed abruptly elevating occurrence rates of floods and it is particularly
393 the case during recent 100 years, i.e. 1900-2000 (Fig. 9b).

394 Basin-scale hazardous flood events were identified based on flood criterion defined
395 by Mudelsee et al. (2003, 2004). Meanwhile, flood risks of recent 1000 years were
396 evaluated using the kernel estimation method (Fig. 10). The width of the time window
397 was 56 years and 41 years for hazardous floods in the Guangdong and Guangxi



398 provinces based on the cross validation method (Figs. 10b and 10d). Moreover, the time
399 window of 30 years was used to have a closer look at the occurrence rates of hazardous
400 floods (Figs. 10a and 10c). It can be seen from Fig. 10 that hazardous floods had an
401 increasing tendency in general, except the time interval of 1400-1800 which was
402 characterized by decreasing occurrence rates. Recent 200 years witnessed a sharp
403 amplification of floods in the Guangxi and Guangdong provinces. Remarkable
404 amplification of floods in the middle and lower Pearl River basin, and particularly in
405 recent 200 years, should arouse a considerable concern.

406

407 **5. Discussions**

408 Change points and trends analyses showed that only a few stations showed a
409 change point or significant trend in the flood peaks. In other words, the flood peaks are
410 stationary in most of the stations considered in this study. Our previous study has also
411 detected the trends in flood peaks before and after the change points, indicating that no
412 significant trends have been found (e.g. Zhang et al., 2014). Taking change points in
413 the West River as example, Change points of flood peaks in the mainstream of the West
414 River occurred approximately in 1990 in spite of a few difference. The flood peaks of
415 the West River basin are heavily influenced by the confluences of tributaries on the
416 upstream of the West River and the factors causing abrupt changes in mean are
417 complicated and blurry. The influence of hydraulic facilities is considerable. However,
418 after the 1990s a few hydraulic facilities have been constructed and their influence can
419 be ignored. Analysis of precipitation extremes in the Pearl River basin indicated that
420 the amount of rainfall had changed little but its variability had increased over the time
421 interval divided by change points. Abrupt changes of precipitation maxima were
422 shifting in different seasons. However, change points of precipitation maxima in



423 summer occurred in 1990, 1988 and 1991, which are in line with changes points of
424 flood peaks of the West River basin. It should be noted that floods occur mainly during
425 the summer season. Therefore, it can be tentatively stated that abrupt changes of flood
426 peaks of the West River basin are mainly the result of abrupt behavior of precipitation
427 maxima. However, due to spatiotemporal patterns of precipitation maxima in the Pearl
428 River basin and the production and confluence of flood streamflow, the abrupt behavior
429 of flood peaks usually does not match that of precipitation maxima. Moreover, human
430 interferences also introduce considerable uncertainty and cause obscure relations
431 between abrupt changes of flood peaks and precipitation maxima. This analysis implies
432 abrupt changes of flood peaks due to various influencing factors.

433 The Guangdong province is dominated by high urbanization, highly-developed
434 socio-economy and dense population density and it is particularly the case for the PRD
435 region (Figs. 1b, 1c and 1d). Intensifying human activities, such as in-channel sand
436 dredging, building of levees and fast urbanization, have greatly altered physical and
437 geographical features of underlying surfaces and hence modified the flooding processes.
438 The volume of sand dredged during the 1990s in the North and East River basins was
439 respectively 3.38×10^6 m³/year and 1.50×10^6 m³/year, causing deepening of river
440 channel (e.g. Luo et al., 2007). Massive building of levees and simplification of river
441 channel systems have caused wide-spread gathering of flood waters and hence
442 amplification of floods.

443 Taking the PRD as an example, during recent 60 years, more than 20000 levees
444 were combined with 400 levees and the length of river channel was reduced from 10000
445 km to 5000 km. Besides, the construction of large-scale reservoirs greatly reduced the
446 occurrence rates and magnitude of floods (Figs. 11a, 11b and 11c). However, fast and
447 massive urbanization, such as the urbanization rate of the Guangdong province reaching



448 67.67% caused fast production of floods and hence enhanced flood risk (Fig. 10a).
449 Increasing flood magnitudes in the recent 60 years (Fig. 3a) and increasing occurrence
450 rates of floods in the recent 100 years (Fig. 9b) caused increasing losses of agricultural
451 production and increasing casualty (Figs. 11c, 11d, and 11f). However, when compared
452 to Guangdong province, Guangxi province was dominated by lower urbanization and
453 less dense population density (Fig. 1d), human activities did not exert significant
454 impacts on floods. Increasing precipitation extremes and particularly increasing
455 precipitation concentration (Zhang et al., 2012, 2013) triggered discernable
456 amplification of floods in the Guangxi Province. Therefore, recent 200 years also
457 witnessed the intensification of hazardous floods which undoubtedly posed a challenge
458 for mitigation of flood hazards in the lower Pearl River basin, particularly the PRD
459 region. Although the fluvial disastrous floods may be ignored in the early time, the
460 increasing numbers of extreme floods are significant and sharp. The no reported flood
461 events in the early time may be one of the reasons of sharp increases. However, we
462 think it is not enough to explain this. Taking Guangxi province as an example, the
463 significant increase is continual, especially for recent 200 years (Fig. 9b). In recent 200
464 years, no reported extreme floods did not have so much difference that number of floods
465 is still significantly increasing. When historical flood records are tried to use, the no
466 reported flood events are the problem that the users must face not only for us but also
467 for Mudelsee et al. (2003). In addition, the spanning time is larger, the problem is more
468 difficult to solve. Nevertheless, the historical flood information can definitely provide
469 valuable information to improve our understanding of the changes in flood frequency.

470

471 **6. Conclusions**

472 Evaluation of flood risks was done in both space and time across the Pearl River



473 basin, China, based on peak flood flow data from 78 hydrological stations during the
474 period of 1951-2014 and 1000-year flood hazard records. The following conclusions
475 can be drawn from this study:

476 (1) No statistically significant changes can be detected in the peak flood flow series at
477 most of the stations, but significant changes were observed at 16 out of 78 stations.
478 Stations with significant peak flood flow changes were found in mainstream of the
479 West River basin, the East River basin and rivers in western parts of the coastal
480 regions and the change points were mainly during the 1990s. Abrupt changes of
481 peak flood flow in the West River basin were attributed to the abrupt behavior of
482 precipitation extremes. Construction of large-scale hydraulic facilities and
483 reservoirs was the major cause behind abrupt behavior of peak flood flow in the
484 coastal region.

485 (2) The northern parts and mainstream of the West River basin and northern North
486 River basin were dominated by significant increasing peak flood flow, implying
487 amplification of floods. Peak flood flow in the East River basin however had
488 significant decreasing trends which were attributed to the changes in precipitation
489 extremes. It should be emphasized that precipitation extremes were increasing in
490 southeastern West River basin and west parts of the coastal regions, and peak flood
491 flow in these regions was decreasing. Expanding agricultural irrigation and
492 hydrological regulation of reservoirs were the causes of decreasing peak flood flow
493 in these regions. A closer look at the abrupt behavior of peak flood flow indicated
494 that significant increasing peak flood flow was identified during 1981-2010 at 25-



495 35% of the stations in the West River basin; significant decreasing peak flood flow
496 was observed during 1951-2014 at 25-30% of the stations in the East River basin;
497 and 30-35% of the stations in the western parts of the coastal region were dominated
498 by significant increasing peak flood flow during 1951-1975, and 35-50% of the
499 stations were dominated by significant decreasing peak flood flow.

500 (3) The largest three flood events were concentrated during two time intervals, i.e.
501 1965-1970 and 1993-2002. The percentage of stations characterized by floods of >
502 10-year flood was increasing after 1975. The East River basin was dominated by
503 the concentrated occurrence of the three largest flood events during 1958-1969 and
504 the percentage of stations with > 10-year flood was increasing after 2000. Results
505 indicated temporal and spatial clustering of flood hazards. This point should arouse
506 considerable concern for the mitigation to flood hazards.

507 (4) Historical flood records of recent 1000 years told an interesting story about flood
508 risks from a long term perspective. Flood risks of the middle and lower Pearl River
509 basin were enhancing and it is particularly the case in the recent 100 years.
510 Particularly, the flood risks in the middle and lower Pearl River basin in terms of
511 disastrous flood regimes were increasing, posing serious challenges for mitigation
512 of flood hazards in the PRD region.

513

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519

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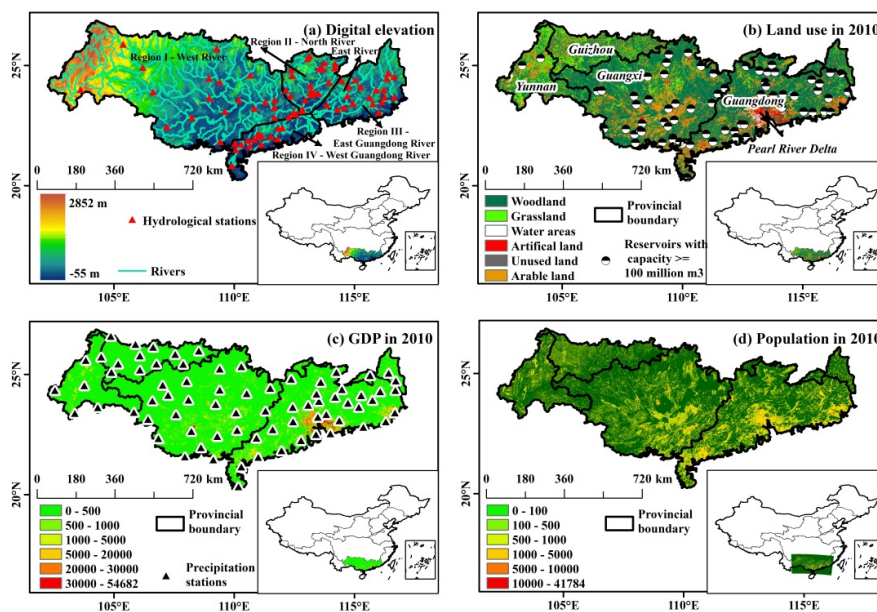
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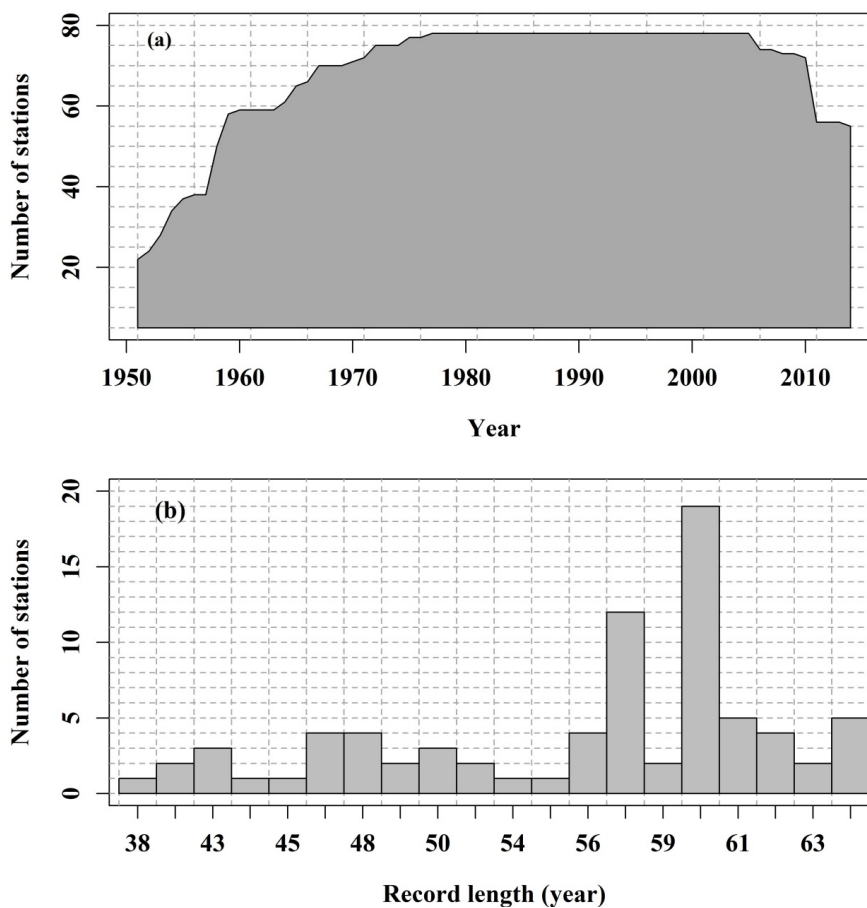


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639 Fig. 1 Locations of hydrological stations, precipitation gauging stations, and water
640 reservoirs, and spatial patterns of land use, socio-economy and population across the
641 Pearl River basin.

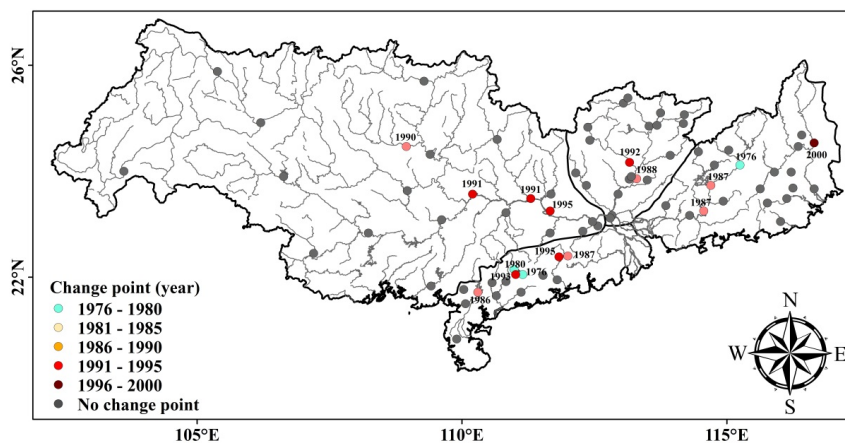
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Fig. 2 Information on peak flood flow dataset.

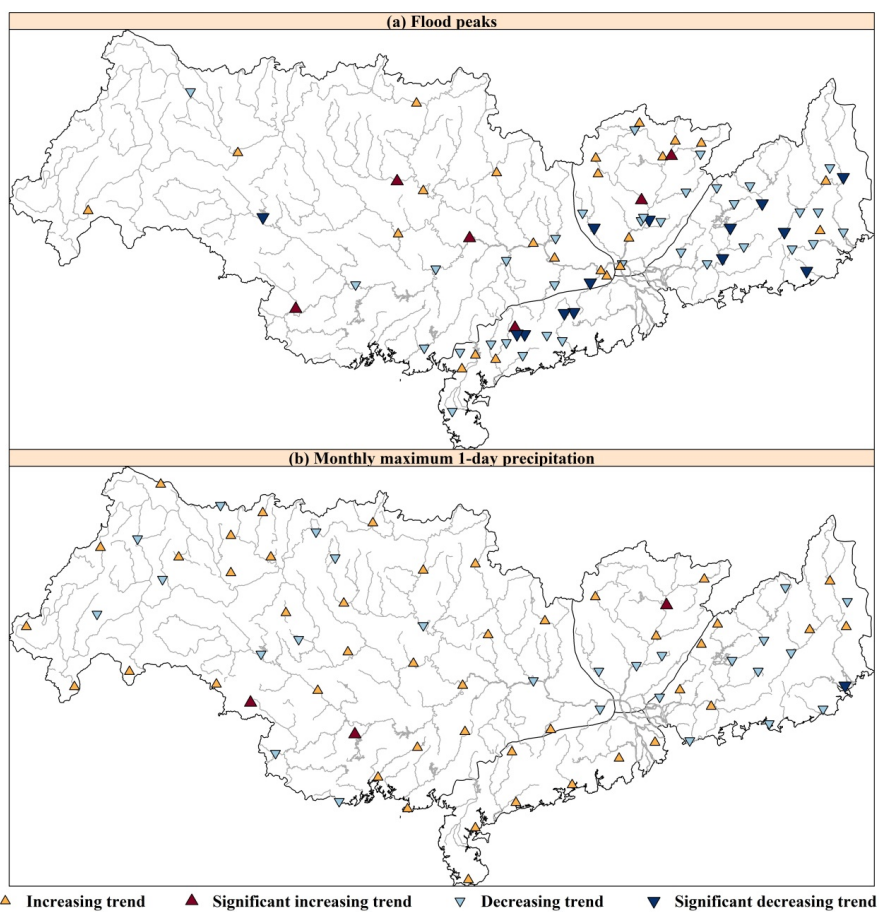


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Fig. 3 Spatial distribution of change points by Pettitt test for peak flood flow changes



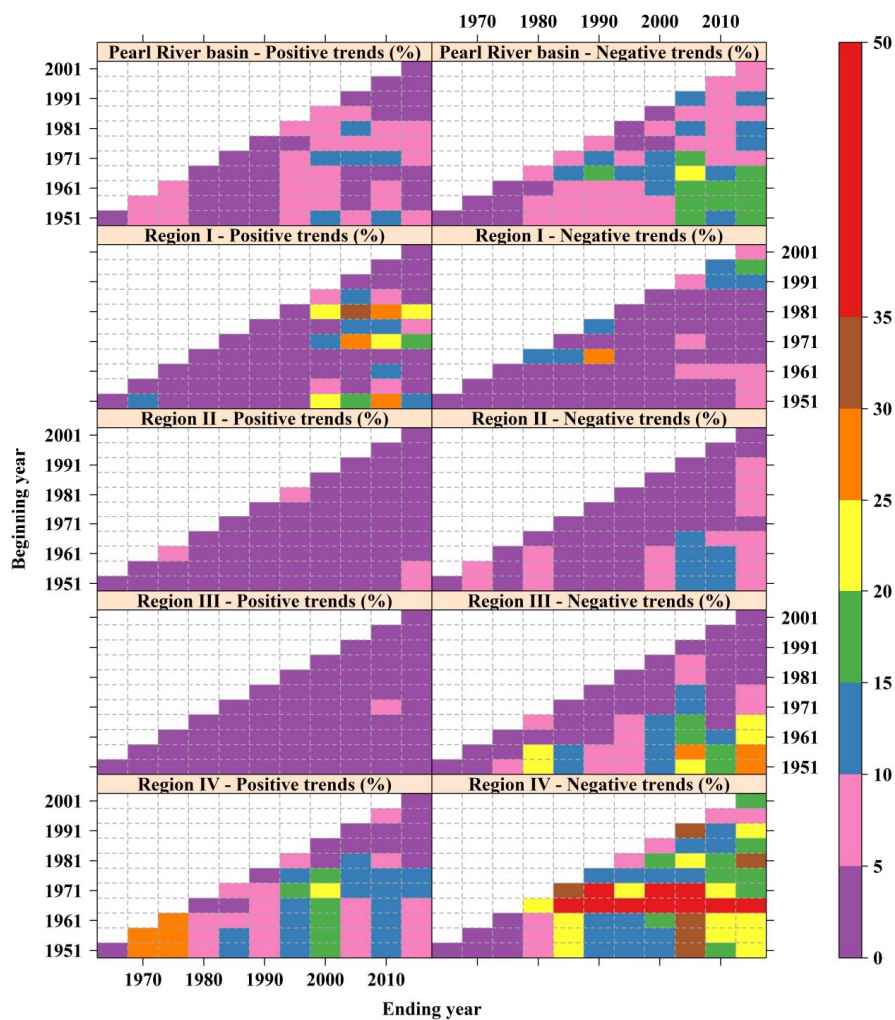
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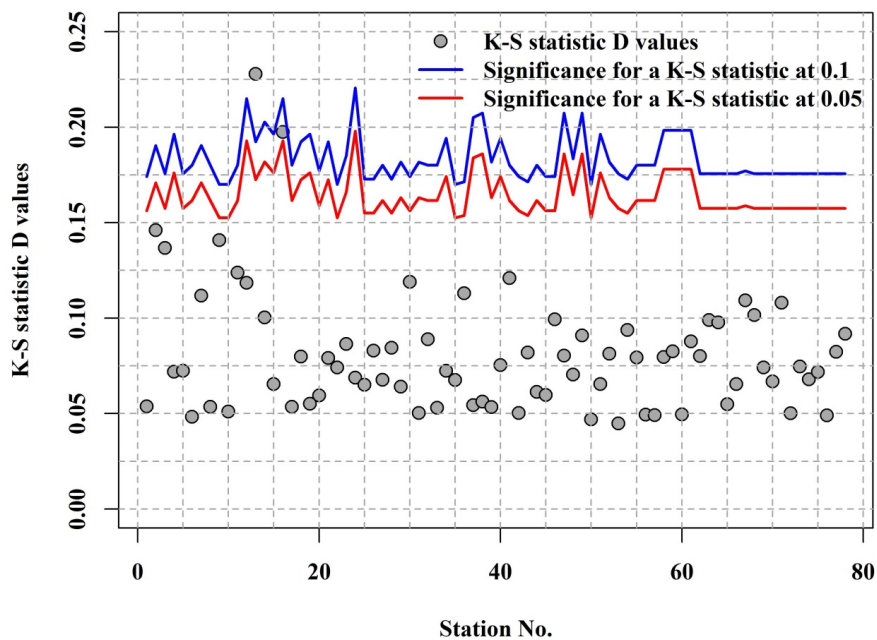
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Fig. 4 Trends in flood peak flow and precipitation extremes

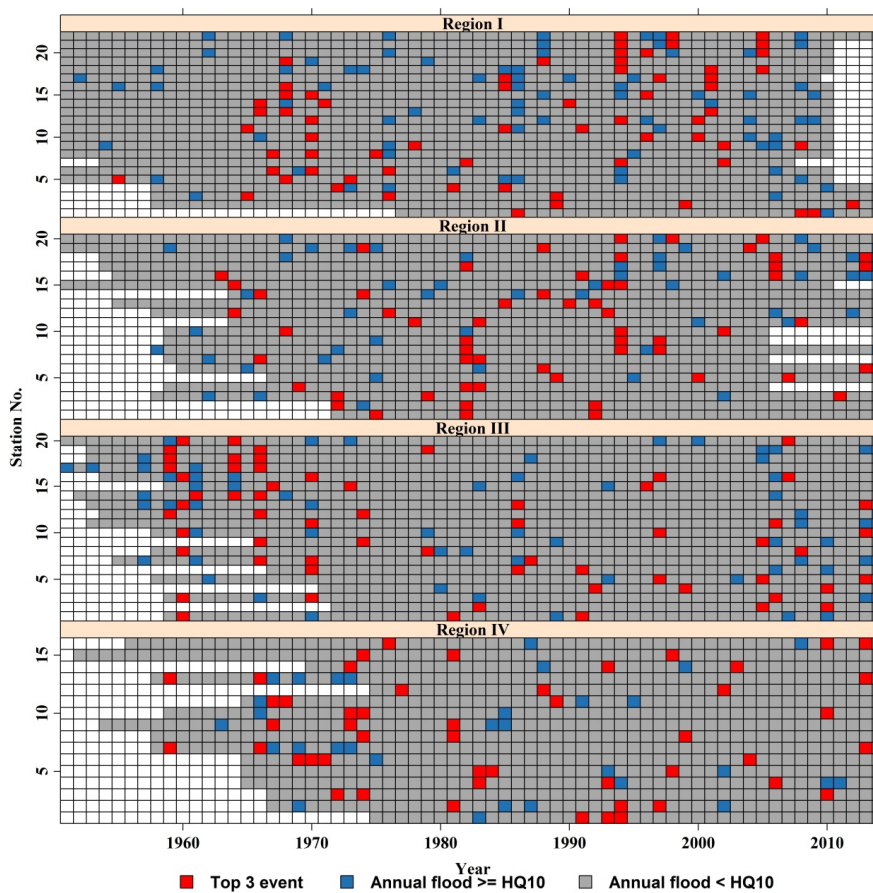


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Fig. 5 Percentage of stations with significant trends in peak flood flow

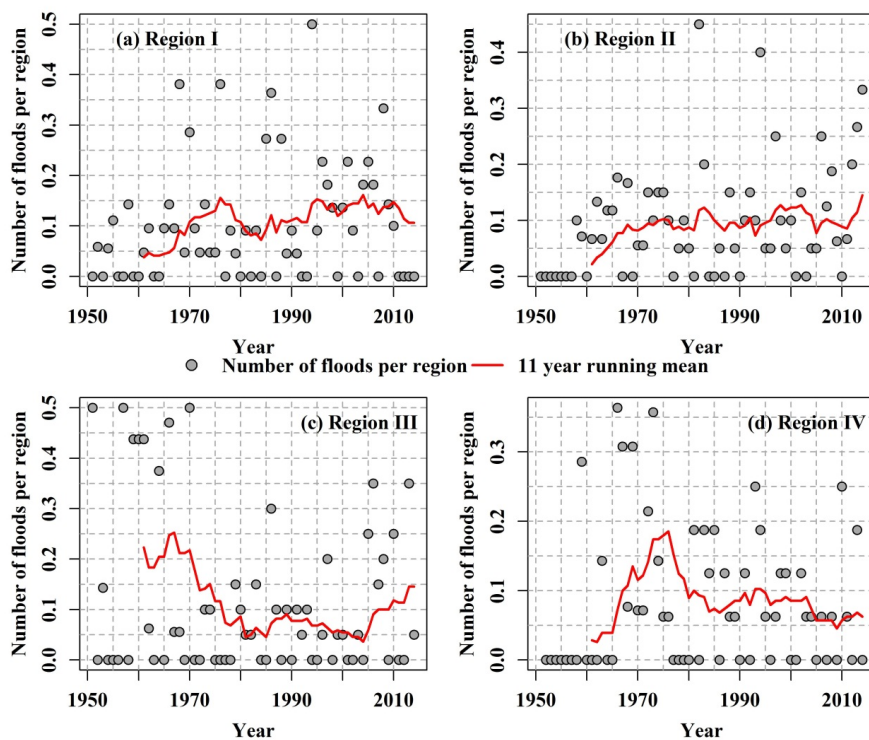


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659 Fig. 6 K-S test results of performance of GEV fitting of peak flood flow of the Pearl
660 River basin
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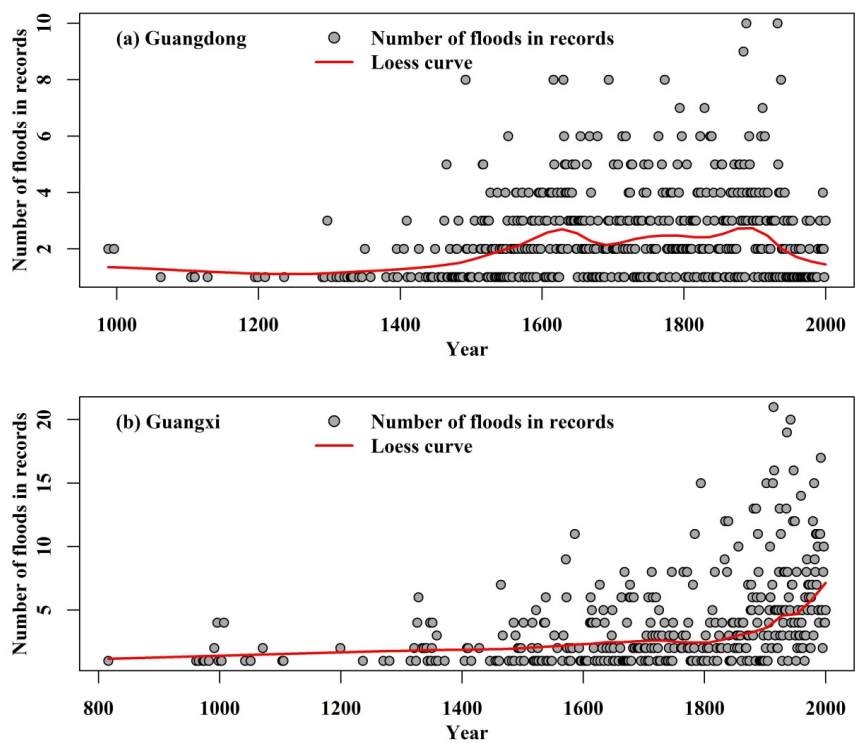
Fig. 7 GEV model based estimated return periods of peak flood flow at hydrological stations considered in this study across the Pearl River basin.



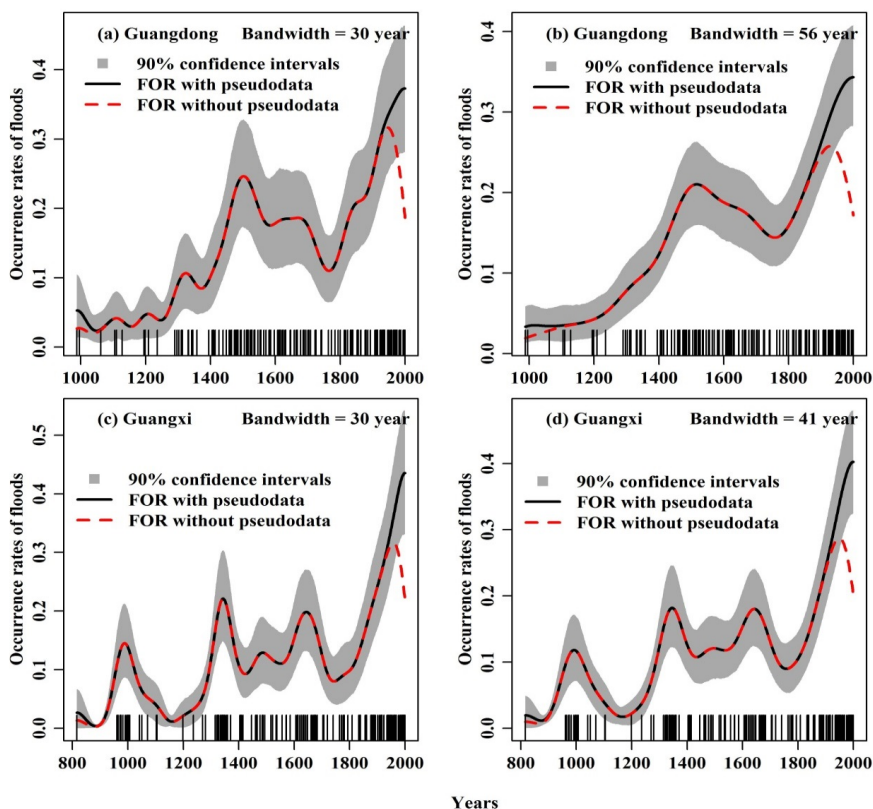
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667 Fig. 8 Temporal changes of percentage of stations with flood events of magnitude of >
668 10-year flood magnitude

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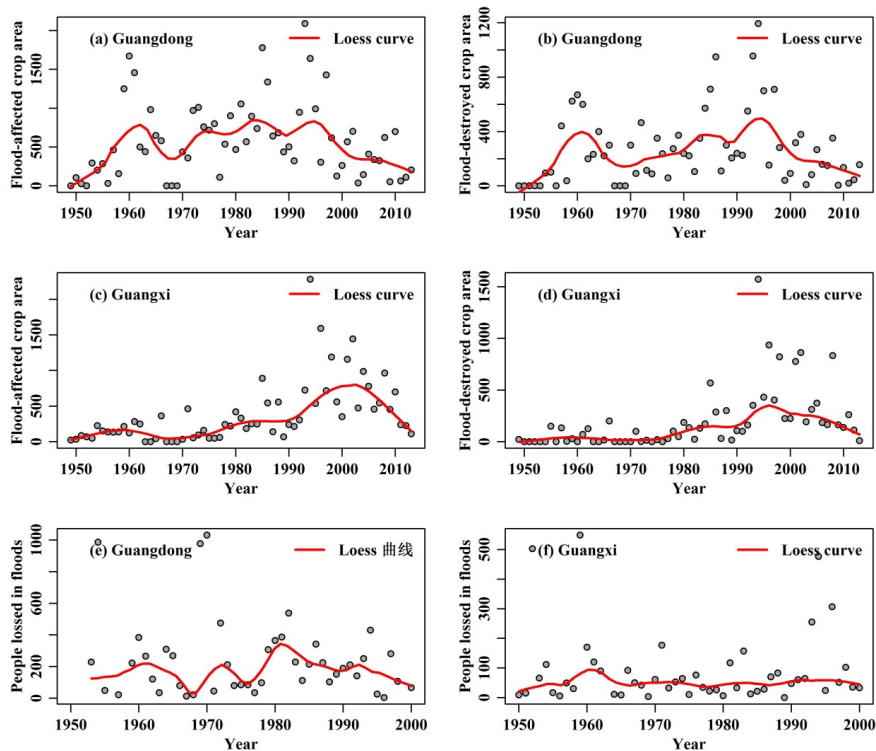
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671 Fig. 9 Temporal changes in occurrence rates of flood hazards of past 1000 years in
672 Guangdong (a) and Guangxi provinces (b).
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675 Fig. 10 Temporal variations of frequency of flood hazards during past 1000 years in
676 Guangdong and Guangxi provinces.

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679 Fig. 11 Temporal changes of flood hazard-induced agricultural losses and mortality in
680 Guangdong and Guangxi provinces.