



1	The relative importance of antecedent soil moisture and precipitation
2	in flood generation in the middle and lower Yangtze River basin
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4	Sheng Ye ¹ , Jin Wang ¹ , Qihua Ran ² *, Xiuxiu Chen ¹ , Lin Liu ¹
5	
6	¹ Institute of Hydrology and Water Resources, School of Civil Engineering, Zhejiang
7	University, Hangzhou 310058, China
8	² State Key Laboratory of Hydrology-Water Resources and Hydraulic Engineering,
9 10	Hohai University, Nanjing 210098, China
	* ~
11	* Corresponding author: Qihua Ran
12	
13	Email address of the corresponding author: ranqihua@zju.edu.cn
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19 Abstract

20	Floods have caused severe environmental and social economic losses worldwide in
21	human history, and are projected to exacerbate due to climate change. Many floods are
22	caused by heavy rainfall with highly saturated soil, however, the relative importance of
23	rainfall and antecedent soil moisture and how it changes from place to place has not
24	been fully understood. Here we examined annual floods from more than 200
25	hydrological stations in the middle and lower Yangtze River basin. Our results indicate
26	that the dominant factor of flood generation shifts from rainfall to antecedent soil
27	moisture with the increase of watershed area. The ratio of the relative importance of
28	antecedent soil moisture and daily rainfall (SPR) is positively correlated with
29	topographic wetness index and has a negative correlation with the magnitude of annual
30	floods. This linkage between watershed characteristics that are easy to measure and the
31	dominant flood generation mechanism provides a quantitative method for flood control
32	and early warnings in ungauged watersheds in the middle and lower Yangtze River
33	basin.

- 34 Key words: flood generation, scaling effect, topographic wetness index
- 35





37 1. Introduction

38	Flooding is one of the most destructive and costly natural hazards in the world, resulting
39	in considerable fatalities and property losses (Suresh et al., 2013). River floods have
40	affected nearly 2.5 billion people between 1994 and 2013 worldwide (CRED, 2015),
41	and caused 104 billion dollars losses every year (Desai et al 2015). The damages may
42	be further exacerbated by increasing frequency of extreme rainfall events according to
43	climate change projection (IPCC 2012; Ohmura and Wild 2002). Flood control
44	infrastructures and more accurate predictions are needed to reduce flood damages,
45	which requires better understanding of the underlying mechanism of flood generation.

46 Numerous studies have been conducted to investigate the cause of floods across 47 the world (Bloschl et al 2013; Munoz et al 2018; Zhang et al 2018). Many studies 48 focused on examining the environmental and social characteristics that lead to specific 49 catastrophic flood events (Bloschl et al 2013; Liu et al 2020; Zhang et al., 2018). Others 50 concentrated on single locations, usually catchment outlets, to explore the influential 51 factors of floods and the future trends (Brunner et al., 2016; Munoz et al 2018). Yet 52 given the amount of data and time required, it is not practical to apply these detailed 53 studies to hundreds of catchments to generate an overview of the flood generation 54 mechanism at large scale.

Recently, researchers started to investigate the dominant flood generation mechanisms at regional scales (Berghuijs et al 2019b; Do et al 2020; Garg & Mishra 2019; Smith et al 2018; Tramblay et al 2021; Ye et al 2017). Most of these studies are conducted in North America and Europe with well-documented long-term records (Berghuijs et al 2016; Bloschl et al 2019; Do et al 2020; Musselman et al 2018; Rottler et al 2020). Little work has been conducted on the flood generation mechanisms in





61 China (except Yang et al 2019).

62 As the largest river in China, Yangtze River basin has long suffered from floods. In 63 summer 2020, 378 tributaries of the Yangtze River had floods exceeding the alarm level, 64 causing billions of dollars damage (Xia et al., 2021). With the increasing public 65 awareness, more accurate prediction is needed, which relies on better understanding. 66 However, due to the limitation of observations, there are only a few regional studies of 67 the flood generation mechanism in China, even little in the Yangtze River basin (Zhang 68 et al 2018; Yang et al 2019; Yang et al 2020). The large number of dams and reservoirs 69 built along the river further complicated the situation (Feng et al., 2017; Qian et al 2011; 70 Yang et al 2019).

71 Because of the relatively warm temperature, snowmelt has little impact on flood 72 generation in the Yangtze River basin (Yang et al 2020). Floods in the Yangtze River 73 basin usually occur during summer with relatively wet soil and high rainfall (Wang et 74 al 2021). Heavy rainfall with high antecedent soil moisture has also been identified as 75 dominant driver of floods across world (Beighuijs et al 2019b; Garg et al 2019; 76 Tramblay et al 2021; Wasko et al 2020). However, little study examines the relative 77 importance of rainfall and antecedent soil moisture in flood generation. A quantitative 78 overview of how the combination of rainfall and antecedent soil moisture change across 79 watersheds is currently unavailable in China (Liu et al., 2021; Wu et al., 2015).

Based on the watersheds in the middle and lower Yangtze River basin, this study attempts to explore the following questions: 1) is there a way to quantitatively describe the relative importance of antecedent soil moisture and rainfall on flood generation; and) how would this combination of flood-generation rainfall and soil moisture vary across watersheds, and what are the influential factors. Based on the observations and





85	model estimation (Section 2), the spatial distribution patterns of antecedent soil
86	moisture and rainfall were obtained and analyzed to investigate their individual
87	contribution to flood generation and the influential factors (Section3). This allows for
88	further examination of the relative importance of antecedent soil moisture and rainfall
89	on flood generation and its linkage to watershed characteristics as well as its
90	implications to flood prediction (Section 4), all the results are summarized in Section 5.

- 91 2 Methods
- 92 2.1 Study area

93 Yangtze River is the largest river in China, with a total length of 6,300 kilometers and 94 annual discharge of 920km³ at the outlet (Yang et al., 2018). It drains through an area 95 of 1.8*10⁶ km², lying between 90°33'and 122°25'E and 24°30'and 35°45'N, and is 96 home to over 400 million people, most of which live in the middle and lower Yangtze 97 River basin (YZRB) (Cai et al., 2020). The elevation of the YZRB declines from west 98 to east: from over 3000m in Qinghai-Tibet Plateau, to around 1000m in the central 99 mountain region, and the 100m in Eastern China Plain (Wang et al., 2013). The 100 vegetation types in the YZRB are forests, shrubs, grassland and agricultural land, 101 accounting for 11.85%, 12.65%, 32.26% and 42.88% respectively. Grassland and 102 shrubs are the dominant vegetation in the middle and upper YZRB, while the 103 downstream YZRB is dominated by forests and agricultural land (Miao et al., 2010). 104 There are more than 51,000 reservoirs of different sizes in the whole basin, including 105 280 large ones (Peng et al., 2020).

Most of the YZRB is semi-humid and humid, with s a typical subtropical monsoon
 climate. The mean annual temperature is approximately 13.0 °C, varying from -4 °C





108	to 18°C downstream. The mean annual precipitation of the whole basin is about 1200
109	mm, increasing from 300mm in the western headwaters to 2400 mm downstream. (Li
110	et al., 2021). Most of the precipitation comes between June and September, the premise
111	of persistent heavy rain in the Yangtze River basin is the frequent activity of weak cold
112	air in the north (Tao et al., 1980) and the intersection of mid-latitude air mass and
113	monsoon air mass (Kato et al., 1985). Studies have found that both annual precipitation
114	and the frequency of extreme precipitation events have increased in the middle and
115	lower reaches of the Yangtze River (Qian et al., 2020; Fu et al., 2013). As a result, floods
116	have occurred frequently in the middle and lower reaches of the Yangtze River, where
117	most of the population in the YZRB live (Liu et al., 2018).

118 2.2 Data

119 In this work, we focus on the middle and lower reaches of the Yangtze River for the 120 high population density and increasing flood risk. The 30-meter digital elevation model 121 (DEM) was downloaded from Geospatial Data Cloud (http://www.gscloud.cn/), from 122 which the drainage area corresponding to the hydrological station was extracted by 123 ArcGIS. Daily precipitation data and temperature data between 1970 and 2016 from 124 247 meteorological stations within and near the YZRB were downloaded from China 125 Meteorological Data Network (https://data.cma.cn/). The temperature data was used to 126 estimate potential evaporation. The observed precipitation and estimated potential 127 evaporation were interpolated into the whole YZRB using Thiessen polygon method 128 (Meena et al., 2013). The interpolated precipitation and potential evaporation were then 129 averaged for the drainage area corresponding to each hydrological station.

130 The daily streamflow data was collected from 267 stations from Annual131 Hydrological Report of the People's Republic of China. Among which, 224 stations





- 132 with at least 20 years records from 1970 to 1990 and from 2007 to 2016 were selected.
- 133 Information of 361 reservoirs in the middle and lower YZRB, including capacity and
- 134 controlling area was downloaded and extracted from the Global Reservoir and Dam
- 135 database (GRanD) (Lehner et al 2011). Previous study showed that this database
- 136 provides reliable information of middle and large reservoirs in China (Yang et al 2021).
- 137 Watersheds with more than 80% of the drainage area under control reservoirs according
- 138 to GRanD database and/or located right downstream of reservoirs and water gates were
- 139 considered as watersheds under strong regulation (regulated watersheds).

140 2.3 Calculation of hydrological and topographic characteristics

141 Potential evaporation estimation

- 142 The temperature data was used to estimate potential evaporation following the
- Hargreaves method (Allen et al., 1998; Vicente et al., 2014; Berti et al., 2014).

144
$$ET_0 = 0.0023 \times (Tmax - Tmin)^{0.5} \times (Tmean + 17.8) \times Ra$$
 (1)

145 where ET_0 is potential evaporation (mm/d), *Tmax* is the highest temperature (°C), *Tmin*

146 is the lowest temperature (°C), *Tmean* is the mean temperature (°C), and *Ra* is the outer

- 147 space radiation $[MJ/(m^2 d)]$, which can be calculated as follows:
- 148 $Ra = 37.6 \times d_r \times (\omega_s \sin \varphi \sin \delta + \cos \varphi \sin \delta \sin \omega_s), \qquad (2)$

149 where d_r is the reciprocal of the relative distance between the sun and the earth, ω_s is 150 the angle of sunshine hours, δ is the inclination of the sun (rad), φ is geographic 151 latitude (rad). d_r , δ and ω_s can be calculated by the following formula:

152
$$d_r = 1 + 0.033 \times \cos\left(\frac{2\pi J}{365}\right),$$
 (3)





153
$$\delta = 0.409 \times \sin\left(\frac{2\pi J}{365} - 1.39\right),$$
 (4)

154
$$\omega_S = \arccos(-\tan\varphi\tan\delta), \qquad (5)$$

155 where J is the daily ordinal number (January 1st is 1).

156 Soil water storage estimation

157 The soil water storage was estimated based on the daily water balance (Berhuijs et al.,

158 2019; Deb et al., 2019):

159
$$\frac{\mathrm{d}S}{\mathrm{d}t} = P - ET - \max(Q, 0), \qquad (6)$$

160 Where S is the soil water storage (mm); P is precipitation (mm/d), Q is discharge 161 normalized by area (mm/d), ET is evaporation (mm/d), which can be calculated from 162 potential evapotranspiration (ET_0) :

163
$$ET = \min(0.75 \times ET_0, S),$$
 (7)

164 Topographic wetness index estimation

165 Topographic wetness index was calculated to represent the combined impacts of 166 drainage area and topographic gradient (Alfonso et al., 2011; Grabs et al., 2009):

167
$$TWI = \ln(A_d/\tan\alpha), \qquad (8)$$

168 where A_d is drainage area and α is topographic gradient estimated from DEM.

169 2.4 Quantification of the relative importance of soil moisture and precipitation170 during floods

171 The maximum discharge of each year was selected as annual flood, which was then





172	averaged across years as the mean annual maximum flood (AMF). The observed
173	rainfall on that day and the estimated soil water storage at the day before were also
174	averaged across years as daily rainfall (P) and antecedent soil moisture (S_{θ}). To examine
175	the impacts from long-lasting rainfall event, we also calculated the mean accumulated
176	rainfall from two days (rainfall on the flood day and the day before, P_2) to seven days
177	(weekly rainfall, <i>P</i> ₇).

The antecedent soil moisture (S_0) was normalized by the maximum soil moisture (*Smax*) to approximate the saturation rate (*S'*) as a surrogate of the contribution of soil moisture in flood generation. The daily rainfall (*P*) was normalized by the maximum daily rainfall (*Pmax*) to show the relative intensity (*P'*), representing the contribution of rainfall in flood generation. The accumulated rainfall was also normalized by the maximum two-day to seven-day rainfall.

To quantify the relative importance of antecedent soil moisture and rainfall in flood generation, the ratio between these two factors at the AMFs was derived: SPR = S'/P'. When SPR is larger than 1, floods at those sites are more dominated by antecedent soil moisture; when SPR is less than 1, rainfall is the primary driver of floods.

188 3 Results

189 3.1 Spatial patterns of antecedent soil moisture and precipitation during floods

Because of the latitude and the relative warm climate, floods in the middle and lower Yangtze River basin usually occur in summer, with little influence of snowmelt (Wang et al., 2015; Yang et al 2019). Therefore, in this study we mainly focus on the impacts of antecedent soil moisture and rainfall on flood generation. Figure 2 shows the spatial distribution of normalized antecedent soil moisture and daily rainfall during the annual





195 maximum floods (AMF) in the middle and lower reaches of the Yangtze River.

196 As we can see from Figure 2a, in the middle and lower reaches of YZRB, when 197 AMFs occurred, the antecedent soil saturation rate was generally higher at sites along 198 the major tributaries (i.e., >0.6): the farther away from the main stream, the more 199 saturated the soil was. On the other hand, along and near the main stream and the delta, 200 the antecedent soil saturation rate could be less than 0.4. This may be attributed to the 201 more complicated flood generation mechanism at large scale as well as the strong 202 reservoir control on main stream and water gates regulation in the delta (Gao et al., 203 2018; Long et al., 2020; Zhang et al., 2017).

204 Figure 2b shows the normalized daily rainfall during the AMFs. As we can see, 205 the daily rainfall is relatively high (>0.4) at more than half of the study sites, while it is 206 small (<0.2) for the sites along the main stream, main tributaries, and in the delta 207 (Figure 2b). Comparison between Figure 2a and b suggests that, except the sites on the 208 main stream and in the delta, sites with relatively high antecedent soil saturation rate 209 (i.e., >0.8, the green dots) during AMFs are also the ones with relatively small daily 210 rainfall contribution (i.e., <0.2, the red dots). That is, for these sites, the AMFs are 211 usually occurring at a near saturated soil condition while heavy rainfall at flood day is 212 not necessary, suggesting the relative importance of soil saturation rate. For the sites 213 with both saturation rate and normalized rainfall between 0.4 and 0.6, both the 214 antecedent soil saturation and rainfall play important roles in flood generation. As for 215 the sites on the main stream and in the delta, both antecedent soil moisture and rainfall 216 are low during AMFs, this is likely due to the regulations from large reservoirs and 217 water gates.

218 **3.2** The scaling effect in the contribution of antecedent soil moisture and rainfall





To further investigate the contribution of antecedent soil moisture and rainfall in flood generation and the potential influential factors, we examined their correlation with catchment area (Figure 3). Given the complicated environmental and social impacts, the regulated watersheds and sites on the main stream are presented separately (the green dots and red dots in Figure 3 respectively). Our study will focus on the sites that are not dominated by regulation (the blue dots in Figure 3), for simplicity, we will refer them as natural watersheds.

As we can see from Figure 3, during the occurrence of AMFs, the antecedent soil saturation rate increase with watershed area (*p*-value<0.001), while the normalized daily rainfall decreases with watershed area (*p*-value<0.001). That is, with the increase of watershed size, antecedent soil moisture becomes more and more important in flood generation while the contribution of daily rainfall declines. On the other hand, heavy rainfall in a single day could only be the dominant driver of floods when watershed area is smaller than 1000km².

233 As for the regulated watersheds (green dots in Figure 3), there is no clear 234 correlation between drainage area and antecedent soil saturation rate or normalized 235 rainfall, which is understandable. Meanwhile, both antecedent soil saturation rate and 236 normalized rainfall decreases with watershed area for main stream sites. One 237 explanation is flood regulation: as the major responsibilities of reservoirs on the main 238 stream are to reduce peak flow and postpone the time to flood peak (Volpi et al., 2018). 239 The other explanation is that when watershed size is larger than 100,000km², the impact 240 of antecedent soil moisture declines. To examine this hypothesis, more data from 241 watersheds larger than 100,000km² and with limited human intervention is needed. 242 However, this is above the scope of this work and requires future studies.





243 **3.3** The scaling impacts on accumulated rainfall

244	The saturation of soil before floods could be due to previous rainfall events, and could
245	also be caused by accumulated rainfall in long-lasting rainfall events that eventually
246	generate floods (Xie et al., 2018). Figure 4 presents the correlation between normalized
247	accumulated rainfall and drainage area. When single day rainfall is considered, it is
248	negatively correlated with drainage area (Figure 3a); when accumulated rainfall is
249	considered, the correlation gradually shifts from negative to positive correlation (Figure
250	4). For example, when two-day rainfall was examined, the correlation between
251	accumulated rainfall and drainage area shifts from negative to positive at 1000 km ² ; the
252	negative correlation in Figure 3a is only valid for watersheds larger than 1000 $\rm km^2$
253	(Figure 4a). This transition area increases from 1000 km^2 for two-day rainfall to $10,000$
254	km ² for three-day rainfall (Figure 4b), and 100,000 km ² for five-day rainfall (Figure
255	4e). Eventually, the weekly rainfall has similar positive correlation with drainage area
256	like antecedent soil moisture (Figure 4f). The increase of transition area may be
257	explained by the increasing response time and confluence time in large watersheds: it
258	takes days for the flow events generated by heavy rainfall to reach outlets where it can
259	be observed in large watersheds. This is also consistent with the conclusion in the
260	Yellow River Basin (Ran et al., 2020) and our previous findings of the dominant flood
261	generation mechanism in the middle and lower YZRB: weekly rainfall is the dominant
262	flood driver for sites on the main streams and the major tributaries (Wang et al 2021).
263	The regulated watersheds don't show significant correlation which is understandable
264	for the strong human intervention. For the negative correlation between accumulated
265	rainfall and drainage area at main stream sites, it is difficult to decide whether it is due
266	to scaling effect or human intervention.





267 3.4 The interlink of watershed characteristics, flood, antecedent soil moisture and

268 rainfall

Figure 5 presents the contribution of antecedent soil moisture and rainfall to the AMFs at the study watersheds, the circles are scaled by watershed size and colored with topographic gradient. Except the watersheds with strong human intervention (regulated ones and the ones on main stream), there is a negative correlation between the contribution of rainfall and antecedent soil moisture, indicating the shift of dominance from rainfall to antecedent soil moisture across watersheds.

275 Figure 6 shows the influential factors of the relative importance of antecedent soil 276 moisture and rainfall. For the natural watersheds (the circles), SPR increases with 277 drainage area and declines with topographic gradient. That is, the larger the drainage 278 area is, the more essential the contribution of antecedent soil moisture to floods is, and 279 the less influential rainfall is in flood generation. For watersheds with similar drainage 280 area (i.e., the blue and light blue dots in Figure 6b), topographic gradient also cast 281 impacts on SPR: SPR decreases with slope. That is, the relative importance of rainfall 282 increases at steeper watersheds. This may be attributed to the shortened hydrological 283 response time due to the steep topography which facilitates rainfall induced floods 284 generation. As a combination of both drainage area and topographic gradient, TWI is 285 positively correlated with SPR at natural watersheds, with less scatter than the 286 correlation between SPR and drainage area or topographic gradient alone. There is also 287 positive correlation between SPR and TWI for the regulated watersheds along 288 tributaries (black triangles), though much scatter. However, the sites on main stream 289 show opposite pattern: the SPR at these sites decreases with TWI and drainage area. It 290 is difficult to determine whether this is because of reservoir regulation or not. More data





- 291 about watersheds larger than 10,000km² but with limited human intervention are
- 292 needed to examine this hypothesis.
- Besides TWI, SPR is also correlated with the magnitude of AMF (Figure 7). As Figure 7 shows, the area normalized flood peak declines with flood-generation SPR. Watersheds with large flood peak are mostly the ones with steep topographic gradient and small SPR (i.e., SPR<1) and *vice versa*. Similar correlation was also found at event scale in our experimental mountainous watershed, which locates at a headwater of Yangtze River (Liu et al 2021).

299 4 Discussion

300 4.1 The relative importance of antecedent soil moisture and rainfall in flood301 generation

302 While soil moisture and rainfall are the two main drivers of floods in the middle and 303 lower Yangtze River basin, the dominance of each factor varies across watersheds. 304 Floods in large watershed are usually generated when soil is almost saturated despite 305 of the relatively small rainfall amount, while heavy rainfall is the dominant driver in 306 small to medium watersheds (Figure 3). This shift of dominance may be attributed to 307 the longer confluence time in the large watersheds and the fact that small watershed is 308 easy to reach saturation (Sharma et al., 2018). The rising contribution of antecedent soil 309 moisture in large watersheds was consistent with the findings in Australian watersheds 310 (Wasko & Nathan, 2019); and the declining influence of rainfall at larger watersheds 311 was also found in Indian watersheds (Garg et al 2019).

312 As a result, the natural watersheds in Figure 5 could be grouped into three classes 313 based on their drainage area and topographic gradient. When a watershed is large and





flat, flood occurrence is mainly determined by soil saturation; when a watershed is small and steep, heavy rainfall takes over the dominance; when a watershed is small and the topographic gradient is also gentle, the occurrence of AMF requires both highly saturated soil and relatively heavy rainfall.

318 4.2 Linkage between topographic characteristics, SPR and floods

319 The correlation between TWI and SPR (Figure 6c) suggests that the relative importance 320 of soil moisture and rainfall could be inferred from topographic characteristics. Despite 321 of the strong human intervention, this correlation sustains even in those regulated 322 watersheds (black squares in Figure 6c). That is, we could derive the relative dominance 323 of soil moisture and rainfall in flood generation in specific watershed from its TWI. 324 This helps quantify the importance of soil moisture and rainfall in flood generation in 325 the existing work. Rainfall and soil moisture level have been identified as dominant 326 drivers of floods, individually or together, in watersheds worldwide (Berghuijs et al 327 2016, 2019b; Garg & Mishra 2019; Tramblay et al 2021; Ye et al 2017). Our findings 328 further identified the influential factors of their importance and provide a way to 329 quantitatively estimate it from topographic characteristics that are easy to measure.

Meanwhile, the SPR also present a negative correlation with the magnitude of AMFs (Figure 7). Similar correlation was also found in the observations from our experimental watershed, a headwater of Yangtze River (Liu et al 2021). The ratio of observed antecedent soil moisture and event precipitation also presents similar decline trend with discharge at event scale. That suggests that the negative correlation between flow and SPR is not only valid across watersheds, and may also be applied at event scale within watershed.





337	That is, based on the topographic characteristics, we could derive the relative
338	importance of soil moisture and rainfall in flood generation (SPR); and from this
339	relative importance ratio, we could further infer the average flood magnitude at these
340	watersheds. As a result, we could link the topographic characteristics and annual floods
341	through the characteristic SPR during the AMFs.

342 4.3 Implications

These findings could be helpful for potential flood risk evaluation and early warning in ungauged basins, e.g., headwaters in the mountainous region. With the construction of large reservoirs, the capability of flood risk control has improved substantially along main stream (Zou et al., 2011; Zhang et al., 2015). However, it is still difficult for early warnings in upstream mountainous watersheds, which are vulnerable to floods but difficult for hydrological modeling and prediction due to little hydrologic records.

349 Our findings suggest that we could derive the flood-generation SPR of each 350 watershed from drainage area and topographic gradient that are easy to measure. Using 351 the soil moisture from remote sensing data and precipitation forecast, we could have 352 real-time prediction of SPR values. Once the predicted SPR value gets closer to the 353 flood-generation SPR in Figure 6c, early warnings of floods can be generated. 354 Mountainous watersheds are usually small and steep, where heavy rainfall is the 355 dominant driver of AMFs (Figure 5). Thus, the requirement of the accuracy of 356 antecedent soil moisture would be lower, soil moisture from recent remote sensing 357 images may be used for approximation. Combining with real-time precipitation forecast, 358 early warnings of AMFs could be generated once the estimated SPR is close to the 359 flood-generation SPR. Besides, the correlation between SPR and flood peak provides 360 information of the likely flood magnitude in ungauged watersheds. Flood control





361 infrastructures could then be designed based on the potential flood peak derived from

362 the flood-generation SPR that is estimated from topographic characteristics.

363 4.4 Limitations

Previous works usually identify the dominant flood generation mechanism based on the comparison of the timing of events (Berghuijs et al 2016; 2019b; Bloschl et al 2017; Ye et al 2017). Similar work has been implemented in our study watersheds, suggesting the importance of soil moisture and rainfall (Wang et al 2021). Based on that, we further looked into the records to quantitatively evaluate the relative importance of soil moisture and rainfall in flood generation. However, there are limitations in our methods.

370 The precipitation data we used were averaged for the study watersheds from 247 371 meteorological stations. Given the large area and considerable spatial heterogeneity, the 372 precipitation data we used may not always be representative of the actual precipitation 373 events. The daily data could also average the rainfall intensity at hourly scale, which 374 could be influential in small mountainous watersheds. The estimation of soil moisture 375 is for sure highly simplified, which cannot be considered as precise estimation. Yet, 376 after normalizing by the maximum, this can still provide the generation variation trend 377 of soil moisture in the water cycle, and used for the relative comparison with rainfall. 378 For further implementation of this method, more sophisticated models or remote 379 sensing data are needed to improve our estimation of soil moisture, and refine the 380 estimation of the flood-generation SPR.

381 Moreover, this work is focused on the importance soil moisture and rainfall, 382 without consideration of snowmelt due to the warm and humid climate in the study 383 watersheds. To apply our findings to cold watersheds with significant impact of snow,





384	the snowmelt component needs to be incorporated. In addition, our method is based on
385	the average values from many years. While previous work indicated that the occurrence
386	of floods in our study watersheds are highly concentrated (Wang et al 2021), there could
387	be strong inter-annual variability in other watersheds. In future studies, annual scale
388	and event scale analysis are needed to examine and improve our findings before it can
389	be applied to watersheds with more diverse climate and landscape conditions.

390 5 Conclusions

391 Heavy rainfall on highly saturated soil was identified as the dominant flood generation 392 mechanism across world (Berghuijs et al 2019; Wang et al 2021; Wasko et al 2020). 393 This study aims to further evaluate the relative importance of antecedent soil moisture 394 and rainfall on floods generation and the controlling factors. Climate and hydrological 395 data from 224 hydrological stations and 247 meteorological stations in the middle and 396 lower reaches of the Yangtze River basin was analyzed, along with the modeled soil 397 moisture. Except the regulated watersheds, the relative importance of antecedent soil 398 moisture and daily rainfall present significant correlation with drainage area: the larger 399 the watershed is, the more essential antecedent soil saturation rate is in flood generation, 400 the less important daily rainfall is.

401 Using the antecedent soil saturation rate and normalized rainfall as coordinates, the 402 flood generation mechanism(s) of study watersheds could be grouped into three classes: 403 antecedent soil moisture dominated large watersheds, heavy rainfall dominated steep 404 and small to middle size watersheds, and small to middle size watersheds with gentle 405 topographic gradient where floods occurrence requires both highly saturated soil and 406 heavy rainfall. Our analysis further shows that the ratio of relative importance between 407 antecedent soil moisture and rainfall (SPR) can be predicted by topographic wetness





index. When the topographic wetness index is large, the dominance of antecedent soil
moisture for extreme floods is stronger, and *vice versa*. The SPR also presents negative
correlation with area normalized flood peak.

411 With the potential increase of extreme rainfall events (Gao et al., 2016; Chen et 412 al., 2016), upstream mountainous watersheds in the middle and lower Yangtze River 413 basin are facing higher risk of extreme floods. The lack of hydrological records further 414 increases the vulnerability of people in these watersheds. Our findings provide a 415 quantitative way to estimate the possible flood risk for these ungauged watersheds. 416 Based on measurable watershed characteristics (i.e., drainage area and topographic 417 gradient), the flood-generation SPR of each watershed can be derived. The real-time 418 SPR can be approximated by recent soil moisture data from remote sensing images and 419 real-time rainfall forecast, especially in the small and steep watersheds where heavy 420 rainfall is the most dominant factor. Early warnings could then be generated once the 421 real-time SPR approaches the flood-generation SPR. Flood control infrastructures 422 could also be designed based on the flood magnitude estimated from the flood-423 generation SPR.

Future analysis at event scale could help generate the flood-generation curve between SPR and discharge at event scale to further improve flood risk predictions in these small ungauged watersheds. With more data from other regions and improved estimation or observation of soil moisture, we could expand our analysis to watersheds with more diverse climate and topographic characteristics to examine and refine our findings and to enhance our understandings of flood generation.

430

431 Data availability





432	DEM data was downloaded from Geospatial Data Cloud at http://www.gscloud.cn/.
433	Climatological data used in this study was obtained from China Meteorological Data
434	Network, which can be accessed at http://data.cma.cn/. Discharge data comes from
435	Annual Hydrological Report of the People's Republic of China issued by Yangtze River
436	Water Resources Commission.
437	
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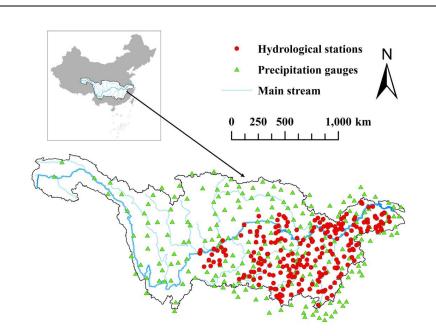




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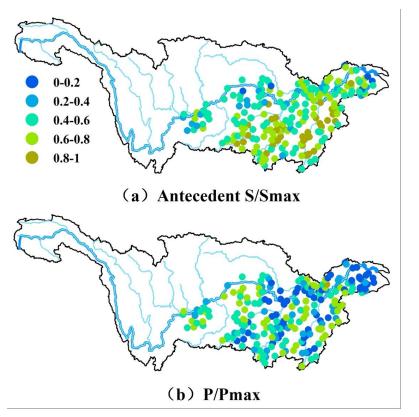




- 714
- 715 Figure 1: Map of the Yangtze River basin, and the climate stations and hydrological
- 716 stations. The blue line is the main stream of Yangtze River.
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719 Figure 2: the spatial distribution of (a) antecedent soil moisture during annual

720 maximum flood, normalized by maximum storage; (b) daily precipitation during annual

721 maximum flood, normalized by maximum daily precipitation.

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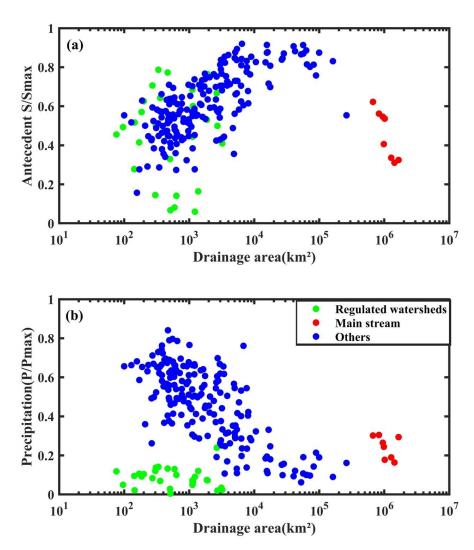


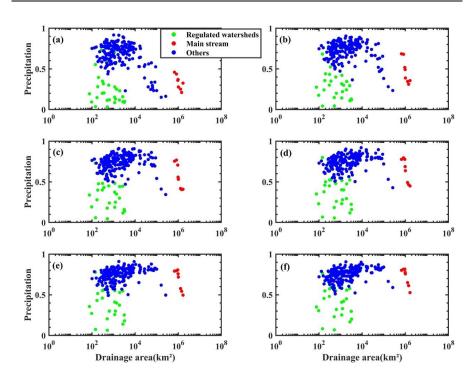


Figure 3: Scatterplot between the drainage area and (a) the antecedent soil moisture of AMF events normalized by maximum storage (the linear regression for blue dots: $R^2 =$ 0.46, *p*-value<0.001); (b) the precipitation at the day of AMF events normalized by maximum daily precipitation (the linear regression for blue dots: $R^2 = 0.61$, *p*value<0.001). The green ones represent the regulated watershed, the red ones represent the sites on the main stream, and the rest sites are shown in blue.

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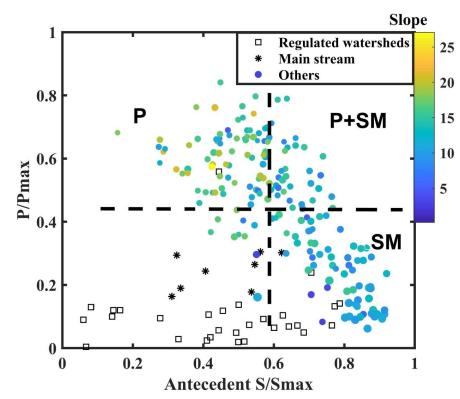


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Figure 4: Scatterplot between the drainage area and the accumulated rainfall of (a) two
days; (b) three days; (c) four days; (d) five days; (e) six days; and (f) seven days on
AMF events, normalized by maximum of accumulated precipitation.







739 Figure 5: Scatterplot of the normalized rainfall and antecedent soil moisture, the color

740 represents topographic gradient and the size of circles is scaled by drainage area.

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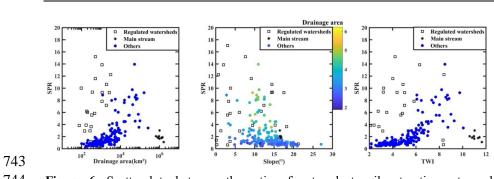
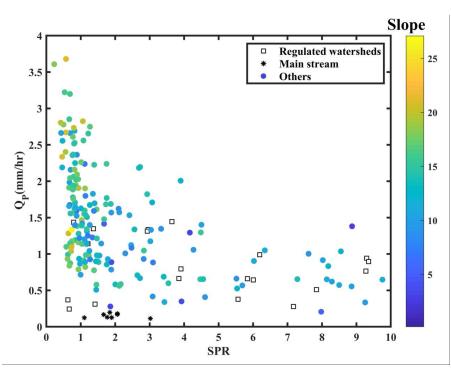


Figure 6: Scatterplots between the ratio of antecedent soil saturation rate and normalized precipitation (SPR) and (a) drainage area; (b) slope; and (c) topographic

746 wetness index (TWI).







749 Figure 7: Scatterplot between the ratio of antecedent soil saturation rate and normalized

750 precipitation (SPR) and area weighted annual maximum discharge (QP), the color

751 represents topographic gradient.