
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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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 framework to quantitatively estimate](#)
32 [potential flood risk](#) in ungauged watersheds in the middle and lower Yangtze River
33 basin.

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

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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 **and intensity** of extreme rainfall events
43 according to climate change projection (IPCC 2012; Ohmura and Wild 2002). Flood
44 control infrastructures and more accurate predictions are needed to reduce flood
45 damages, **which requires better understanding of the underlying mechanism of flood**
46 **generation as well as the drivers of change (Villarini & Wasko 2021).**

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

56 Recently, researchers started to investigate the dominant flood generation
57 mechanisms at regional scales (Berghuijs et al 2019b; Do et al 2020; Garg & Mishra
58 2019; Smith et al 2018; Trambly et al 2021; Ye et al 2017). Most of these studies are
59 conducted in North America and Europe with well-documented long-term records
60 (Berghuijs et al 2016; Bloschl et al 2019; Do et al 2020; Musselman et al 2018; Rottler

61 et al 2020). Such researches were just conducted in China recently, though still limited
62 (Yang et al 2019; Yang et al 2020).

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

72 Because of the relatively warm temperature, snowmelt has little impact on flood
73 generation in the Yangtze River basin (Yang et al 2020). Floods in the Yangtze River
74 basin usually occur during summer with relatively wet soil and high rainfall (Wang et
75 al 2021). Heavy rainfall with high antecedent soil moisture has also been identified as
76 dominant driver of floods across world (Beighuijs et al 2019b; Garg et al 2019;
77 Trambly et al 2021; Wasko et al 2020). Recently, studies started to examines the
78 relative importance of rainfall and antecedent soil moisture in flood generation
79 (Brunner et al., 2021; Wasko et al., 2021; Bennett et al., 2018; Bertola et al., 2021).
80 Quantitative evaluation of the relative contribution of rainfall and antecedent soil
81 moisture and its change across watersheds is still limited and currently unavailable in
82 China (Liu et al., 2021; Wu et al., 2015).

83 Based on the watersheds in the middle and lower Yangtze River basin, this study
84 attempts to explore the following questions: 1) is there a way to quantitatively describe

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

94 **2 Methods**

95 **2.1 Study area**

96 [The](#) Yangtze River is the largest river in China, with a total length of 6,300 kilometers
97 and annual discharge of 920km³ at the outlet (Yang et al., 2018). It drains through an
98 area of 1.8*10⁶ km², lying between 90°33'and 122°25'E and 24°30'and 35°45'N, and
99 is home to over 400 million people, most of which live in the middle and lower Yangtze
100 River basin (YZRB) (Cai et al., 2020). The elevation of the YZRB declines from west
101 to east: from over 3000m in Qinghai-Tibet Plateau, to around 1000m in the central
102 mountain region, and the 100m in Eastern China Plain (Wang et al., 2013). The
103 vegetation types in the YZRB are forests, shrubs, grassland and agricultural land,
104 accounting for 11.85%, 12.65%, 32.26% and 42.88% respectively. Grassland and
105 shrubs are the dominant vegetation in the middle and upper YZRB, while the
106 downstream YZRB is dominated by forests and agricultural land (Miao et al., 2010).
107 There are more than 51,000 reservoirs of different sizes in the whole basin, including
108 280 large ones (Peng et al., 2020).

109 Most of the YZRB is semi-humid and humid, with a typical subtropical monsoon
110 climate. The mean annual temperature is approximately 13.0 °C, varying from -4 °C
111 to 18°C downstream. The mean annual precipitation of the whole basin is about 1200
112 mm, increasing from 300mm in the western headwaters to 2400 mm downstream. (Li
113 et al., 2021). Most of the precipitation comes between June and September, the premise
114 of persistent heavy rain in the Yangtze River basin is the frequent activity of weak cold
115 air in the north (Tao et al., 1980) and the intersection of mid-latitude air mass and
116 monsoon air mass (Kato et al., 1985). Studies have found that both annual precipitation
117 and the frequency of extreme precipitation events have increased in the middle and
118 lower reaches of the Yangtze River (Qian et al., 2020; Fu et al., 2013). As a result, floods
119 have occurred frequently in the middle and lower reaches of the Yangtze River, where
120 most of the population in the YZRB live (Liu et al., 2018).

121 **2.2 Data**

122 In this work, we focus on the middle and lower reaches of the Yangtze River for the
123 high population density and increasing flood risk. The 30-meter digital elevation model
124 (DEM) was downloaded from Geospatial Data Cloud (<http://www.gscloud.cn/>), from
125 which the drainage area corresponding to the hydrological station was extracted by
126 ArcGIS. Daily precipitation data and temperature data between 1970 and 2016 from
127 247 meteorological stations within and near the YZRB were downloaded from China
128 Meteorological Data Network (<https://data.cma.cn/>) (Figure 1). The temperature data
129 was used to estimate potential evaporation. The observed precipitation and estimated
130 potential evaporation were interpolated into the whole YZRB using Thiessen polygon
131 method (Meena et al., 2013). The interpolated precipitation and potential evaporation
132 were then averaged for the drainage area corresponding to each hydrological station.

133 The daily streamflow data was collected from 267 hydrological stations from
134 Annual Hydrological Report of the People's Republic of China. Among which, 224
135 stations with at least 20 years records from both the period from 1970 to 1990 and the
136 period from 2007 to 2016 were selected (see Figure S1 for data availability).
137 Information of 361 reservoirs in the middle and lower YZRB, including capacity and
138 controlling area was downloaded and extracted from the Global Reservoir and Dam
139 database (GRanD) (Lehner et al 2011). Previous study showed that this database
140 provides reliable information of middle and large reservoirs in China (Yang et al 2021).
141 Watersheds with more than 80% of the drainage area under control reservoirs according
142 to GRanD database and/or located right downstream of reservoirs and water gates were
143 considered as watersheds under strong regulation (regulated watersheds).

144 2.3 Calculation of hydrological and topographic characteristics

145 *Potential evaporation estimation*

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

$$148 \quad ET_0 = 0.0023 \times (T_{max} - T_{min})^{0.5} \times (T_{mean} + 17.8) \times Ra \quad (1)$$

149 where ET_0 is potential evaporation (mm/d), T_{max} is the highest temperature ($^{\circ}\text{C}$), T_{min}
150 is the lowest temperature ($^{\circ}\text{C}$), T_{mean} is the mean temperature ($^{\circ}\text{C}$), and Ra is the outer
151 space radiation [$\text{MJ}/(\text{m}^2 \cdot \text{d})$], which can be calculated as follows:

$$152 \quad Ra = 37.6 \times d_r \times (\omega_s \sin \varphi \sin \delta + \cos \varphi \cos \delta \sin \omega_s), \quad (2)$$

153 where d_r is the reciprocal of the relative distance between the sun and the earth, ω_s is
154 the angle of sunshine hours, δ is the inclination of the sun (rad), φ is geographic

155 latitude (rad). d_r , δ and ω_s can be calculated by the following formula:

156
$$d_r = 1 + 0.033 \times \cos\left(\frac{2\pi J}{365}\right), \quad (3)$$

157
$$\delta = 0.409 \times \sin\left(\frac{2\pi J}{365} - 1.39\right), \quad (4)$$

158
$$\omega_s = \arcsin(-\tan \varphi \tan \delta), \quad (5)$$

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

160 *Soil water storage estimation*

161 The soil water storage was estimated based on the daily water balance (Berhuijs et al.,
162 2016, 2019):

163
$$\frac{dS}{dt} = P - ET - \max(Q, 0), \quad (6)$$

164 Where S is the soil water storage (mm), which is initially set to 0. Due to the long term
165 of simulation, the change of initial value wouldn't significantly affect the results. P is
166 precipitation (mm/d), Q is discharge normalized by area (mm/d), ET is evaporation
167 (mm/d), which can be calculated from potential evapotranspiration (ET_0), where the
168 soil water storage (S) is used as the upper limit of daily ET:

169
$$ET = \min(0.75 \times ET_0, S), \quad (7)$$

170 The estimation of soil water storage and ET are highly simplified and is not used for
171 prediction but to capture the first order of the seasonal variation and to help develop a
172 framework that differentiates the relative contribution of precipitation and soil moisture
173 in flood generation.

174 *Topographic wetness index estimation*

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

$$177 \quad \quad \quad TWI = \ln(A_d/\tan\alpha), \quad (8)$$

178 where A_d is drainage area and α is topographic gradient estimated from DEM. TWI
179 represents the propensity of subsurface flow accumulation and frequency of saturated
180 conditions, thus can be used to predict relative surface wetness and hydrological
181 responses (Meles et al 2020). It is widely used to quantify topographic impact on
182 hydrological processes (i.e., spatial scale effects, hydrological flow path, etc.), as well
183 as in land surface models for hydrological, biogeochemical and ecological processes
184 (Sorensen et al 2006).

185 **2.4 Quantification of the relative importance of soil moisture and precipitation** 186 **during floods**

187 The maximum daily discharge of each year was selected as annual flood, which was
188 then averaged across years as the mean annual maximum flood (AMF). The observed
189 rainfall on that day and the estimated soil water storage at the day before were also
190 averaged across years as daily rainfall (P) and antecedent soil moisture (S_0). Since
191 almost all the AMFs in our study region come during rainy season when rainfall comes
192 in most of the days. To avoid the bias that may be caused in event separation, the soil
193 moisture at the day before AMF was used as antecedent soil moisture, instead of the
194 day before the event of AMF. To examine the impacts from long-lasting rainfall event,
195 especially for the large watersheds with longer concentration time, we also calculated
196 the mean accumulated rainfall from two days (rainfall on the flood day and the day

197 before, P_2) to seven days (weekly rainfall, P_7).

198 The percentile of antecedent soil moisture (S_0) was calculated to represent the
199 relative saturation of soil moisture in the time series; while the percentile of daily
200 rainfall (P) was estimated to show the relative intensity (P'), representing the relative
201 magnitude of rainfall in flood generation. The percentile of accumulated rainfall was
202 also calculated for the two-day to seven-day rainfall.

203 To quantify the relative importance of antecedent soil moisture and rainfall in flood
204 generation, the ratio between these two factors at the AMFs was derived: $SPR = S'/P'$.
205 When SPR is large, the antecedent soil moisture is much closer to the maximum, while
206 the daily rainfall is less extreme, floods are more affected by the antecedent soil
207 moisture. On the other hand, a smaller SPR indicates relatively larger magnitude of
208 rainfall comparing with antecedent soil moisture, that is, rainfall is more extreme and
209 influential in flood generation.

210 **3 Results**

211 **3.1 Spatial patterns of antecedent soil moisture and precipitation during floods**

212 Figure 2 shows the spatial distribution of the percentile of antecedent soil moisture and
213 daily rainfall during the annual maximum floods (AMFs) in the middle and lower
214 reaches of the Yangtze River. As we can see from Figure 2a, in the middle and lower
215 reaches of YZRB, when AMFs occurred, the percentile of antecedent soil saturation
216 was generally high, most of them are larger than 0.6: the farther away from the main
217 stream, the more saturated the soil was. On the other hand, along and near the main
218 stream and the delta, the antecedent soil saturation rate could be much smaller, even
219 less than 0.4.

220 Figure 2b shows the daily rainfall during the AMFs. As we can see, the percentile
221 of daily rainfall is relatively high (>0.8) at more than half of the study sites, while it is
222 small (<0.5) for the sites along the main stream and in the delta (Figure 2b). Comparison
223 between Figure 2a and b suggests that, except the sites on the main stream and in the
224 delta, sites with relatively high antecedent soil saturation rate (i.e., >0.8 , the blue dots)
225 during AMFs are also the ones with relatively small daily rainfall contribution (i.e.,
226 <0.8 , the light blue and cyan dots). That is, for these sites, the AMFs are usually
227 occurring at a near saturated soil condition while extreme rainfall at flood day is not
228 necessary, suggesting the relative importance of soil saturation rate. For the sites with
229 both the percentile of soil moisture and rainfall between 0.6 and 1, both the antecedent
230 soil moisture and rainfall play important roles in flood generation. As for the sites on
231 the main stream and in the delta, both antecedent soil moisture and rainfall are low
232 during AMFs, this is likely due to the regulations from large reservoirs and water gates.

233 3.2 The scaling effect in the contribution of antecedent soil moisture and rainfall

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

241 As we can see from Figure 3, during the occurrence of AMFs, the percentile of
242 antecedent soil saturation increases with watershed area (p -value <0.001), while the
243 percentile of daily rainfall decreases with watershed area (p -value <0.001). That is, with

244 the increase of watershed size, antecedent soil moisture becomes more and more
245 saturated while the precipitation is less and less extreme during AMFs; suggesting the
246 rising contribution of antecedent soil moisture and declining importance of daily
247 precipitation in flood generation. As for the regulated watersheds (green dots in Figure
248 3), there is no clear correlation between drainage area and the percentile of antecedent
249 soil moisture or rainfall, which is understandable. Meanwhile, both the percentile of
250 antecedent soil moisture and rainfall decreases with watershed area for main stream
251 sites.

252 3.3 The scaling impacts on accumulated rainfall

253 The saturation of soil before floods could be due to previous rainfall events, and could
254 also be caused by accumulated rainfall in long-lasting rainfall events that eventually
255 generate floods (Xie et al., 2018). Figure 4 presents the correlation between the
256 percentile of accumulated rainfall and drainage area. When single day rainfall is
257 considered, it is negatively correlated with drainage area (Figure 3a); when
258 accumulated rainfall is considered, the correlation gradually shifts from negative to
259 positive correlation (Figure 4). For example, when two-day rainfall was examined, the
260 correlation between accumulated rainfall and drainage area shifts from negative to
261 positive at 10,000 km²; the negative correlation in Figure 3a is only valid for watersheds
262 larger than 10,000 km² (Figure 4a). This transition area increases from 10,000 km² for
263 two-day rainfall to 100,000 km² for four-day rainfall (Figure 4c). The number of
264 watersheds with negative correlation also decreases. Eventually, the weekly rainfall has
265 similar positive correlation with drainage area like antecedent soil moisture (Figure 4f).
266 The increase of transition area may be explained by the increasing response time and
267 confluence time in large watersheds: it takes days to generate flow events by heavy

268 rainfall and for them to reach outlets where it can be observed in large watersheds. This
269 is also consistent with the conclusion in the Yellow River Basin (Ran et al., 2020) and
270 our previous findings of the dominant flood generation mechanism in the middle and
271 lower YZRB: weekly rainfall is the dominant flood driver for sites on the main streams
272 and the major tributaries (Wang et al 2021). The regulated watersheds don't show
273 significant correlation which is understandable for the strong human intervention. For
274 the negative correlation between accumulated rainfall and drainage area at main stream
275 sites, it is difficult to decide whether it is due to scaling effect or human intervention.

276 **3.4 The interlink of watershed characteristics, flood, antecedent soil moisture and** 277 **rainfall**

278 Figure 5 presents the percentile of antecedent soil moisture and rainfall during the
279 AMFs at the study watersheds, the circles are scaled by watershed size and colored with
280 topographic gradient. Except the watersheds with strong human intervention (regulated
281 ones and the ones on main stream), there is a negative correlation between the
282 contribution of rainfall and antecedent soil moisture. The lower right of the scatter are
283 mostly big blue dots, which are large watersheds with gentle topographic gradient. That
284 is, AMFs usually occur when soil moisture is close to saturation while extreme rainfall
285 is not necessary for AMFs in these watersheds. On top of the scatter are relatively small
286 yellow and green dots, those are medium to small watersheds with steep topographic
287 gradient. That is, AMFs are usually generated with extreme rainfall, while the saturation
288 of soil moisture is not necessary. This negative correlation indicates the shift of
289 dominance in AMFs generation from extreme rainfall to antecedent soil saturation from
290 small steep watersheds to large flat ones.

291 Figure 6 shows the relative importance of antecedent soil moisture and rainfall. For

292 the natural watersheds (the circles), SPR increases with drainage area and declines with
293 topographic gradient. That is, the larger the drainage area is, the more essential the
294 contribution of antecedent soil moisture to floods is, and the less influential rainfall is
295 in flood generation. For watersheds with similar drainage area (i.e., the [green](#) or light
296 blue dots in Figure 6b), topographic gradient also cast impacts on SPR: SPR decreases
297 with slope. That is, the relative importance of rainfall increases at steeper watersheds.
298 This may be attributed to the shortened hydrological response time due to the steep
299 topography which facilitates rainfall induced floods generation. As a combination of
300 both drainage area and topographic gradient, TWI is positively correlated with SPR at
301 natural watersheds, with less scatter than the correlation between SPR and drainage
302 area or topographic gradient alone. [That is, watersheds with larger area and gentler
303 topographic gradient that are easier to get wet tend to have larger SPR: soil saturation
304 is more important in flood generation. There is no significant correlation between SPR
305 and TWI for the regulated watersheds along tributaries \(black triangles\).](#) However, the
306 sites on main stream show opposite pattern: the SPR at these sites decreases with TWI
307 and drainage area. It is difficult to determine whether this is because of reservoir
308 regulation or not. More data about watersheds larger than 10,000km² but with limited
309 human intervention are needed to examine this hypothesis.

310 Besides TWI, SPR is also correlated with the magnitude of AMF (Figure 7). As
311 Figure 7 shows, the area normalized flood peak declines with flood-generation SPR.
312 Watersheds with large flood peak are mostly the ones with steep topographic gradient
313 and small SPR (i.e., SPR<1) and *vice versa*. Similar correlation was also found at event
314 scale in our experimental mountainous watershed, which locates at a headwater of
315 Yangtze River (Liu et al 2021).

316 4 Discussion

317 4.1 The relative importance of antecedent soil moisture and rainfall in flood 318 generation

319 While soil moisture and rainfall are the two main drivers of floods in the middle and
320 lower Yangtze River basin, the dominance of each factor varies across [the relatively](#)
321 [natural](#) watersheds. Floods in large watersheds are usually generated when soil is almost
322 saturated despite of the relatively small rainfall amount, while extreme rainfall is
323 usually observed during floods in small to medium watersheds ([blue dots in Figure 3](#)).
324 The rising contribution of antecedent soil moisture in large watersheds was consistent
325 with the findings in Australian watersheds (Wasko & Nathan, 2019); and the declining
326 influence of rainfall at larger watersheds was also found in Indian watersheds (Garg et
327 al 2019). [This contrast correlation with watershed size indicates a shift of dominance](#)
328 [in AMFs generation, which may be attributed to the longer confluence time in the large](#)
329 [watersheds and less heterogeneity in small watersheds.](#)

330 [This shift of dominance can be observed more straightforward from the negative](#)
331 [correlation between the percentile of rainfall and antecedent soil moisture in Figure 5.](#)
332 The natural watersheds in Figure 5 could be grouped into three classes based on their
333 drainage area and topographic gradient. When a watershed is large and flat, flood
334 occurrence is mainly determined by soil saturation ([i.e., the big blue dots at the lower](#)
335 [right of the scatter](#)); on the other hand, when a watershed is small and steep, heavy
336 rainfall takes over the dominance ([i.e., the small yellow and green dots at the upper left](#)
337 [of the scatter](#)). Between these two groups are relatively small watersheds with gentle
338 topographic gradient, where the occurrence of AMF requires both highly saturated soil
339 and relatively heavy rainfall. That is, the dominant influential factor(s) in AMFs

340 generation across watersheds is correlated with the topographic characteristics (i.e.,
341 watershed size and topographic gradient). This helps quantify the relative importance
342 of soil moisture and rainfall in flood generation in the existing work.

343 This shift of dominance is not observed in the main stream sites (i.e., cyan dots in
344 Figure 3), where the percentile of both antecedent soil moisture and precipitation
345 declines with drainage area. This may be attributed to the more complicated flood
346 generation mechanism at large scale as well as the strong human intervention on main
347 stream (e.g., reservoirs, water gates regulation, etc.) (Gao et al., 2018; Long et al., 2020;
348 Zhang et al., 2017). The major responsibilities of reservoirs on the main stream are to
349 reduce peak flow and postpone the time to flood peak (Volpi et al., 2018). As a result,
350 the original flood peak would be delayed by regulation and the actual flood peak would
351 occur when rainfall declines/stops and soil water drains. Another possibility is that
352 when watershed size is larger than 100,000km², the impact of antecedent soil moisture
353 declines as well. To examine this hypothesis, more data from watersheds larger than
354 100,000km² and with limited human intervention is needed. However, this is above the
355 scope of this work and requires future studies.

356 **4.2 Linkage between topographic characteristics, SPR and floods**

357 The correlation between TWI and SPR (Figure 6c) demonstrates that the relative
358 importance of soil moisture and rainfall could be inferred from topographic
359 characteristics quantitatively. We could derive the relative dominance of soil moisture
360 and rainfall in flood generation in specific watershed from its TWI for the natural
361 watersheds without significant human intervention. Rainfall and soil moisture level
362 have been identified as dominant drivers of floods, individually or together, in
363 watersheds worldwide (Berghuijs et al 2016, 2019b; Garg & Mishra 2019; Trambly et

364 al 2021; Ye et al 2017). Our findings provide a framework to quantify the relative
365 importance of rainfall and soil moisture and to further identify the influential factors of
366 their importance based on topographic characteristics that are easy to measure.

367 Meanwhile, the SPR also present a negative correlation with the magnitude of
368 AMFs (Figure 7). That is, we could infer the mean annual AMF based on SPR for each
369 watershed. Since the characteristic SPR could be estimated from TWI, we could derive
370 quantitative estimation of the mean AMFs from topographic characteristics that are
371 easy to measure, even in watersheds with little hydrologic records. This would be
372 helpful for flood control management in ungauged watersheds, especially in the
373 mountainous watersheds with risks of flash floods. Similar correlation was also found
374 in the observations from our experimental watershed, a headwater of Yangtze River
375 (Liu et al 2021). The ratio of observed antecedent soil moisture and event precipitation
376 also presents similar decline trend with discharge at event scale. However, the
377 correlation between SPR and discharge at event scale is preliminary, more data with
378 higher resolution and detailed analysis are needed for validation at event scale. For this
379 study, our goal is to present the framework to derive flood generation SPR that could
380 be estimated from topographic characteristics and to provide information of mean
381 AMFs.

382 In conclusion, based on the topographic characteristics, we could derive the relative
383 importance of soil moisture and rainfall in flood generation (SPR); and from this
384 relative importance ratio, we could further infer the average flood magnitude at these
385 watersheds. As a result, we could link the topographic characteristics and annual floods
386 through the characteristic SPR during the AMFs.

387 4.3 Implications

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

394 Our findings suggest that we could derive the flood-generation SPR of each
395 watershed from drainage area and topographic gradient that are easy to measure. The
396 correlation between SPR and flood peak provides information of the mean annual
397 floods in ungauged watersheds. Therefore, in regions without observation data, to build
398 flood control infrastructure such as dams and gates, the mean annual flood peak
399 obtained by SPR based on the topographic characteristics can be used to provide
400 quantitative information for flood control and disaster management. Flood control
401 infrastructures could be designed based on the estimated mean annual flood peak as
402 well as the demographic information. With further validation of this framework at event
403 scale, by using the observed soil moisture from remote sensing data and precipitation
404 forecast to generate real-time prediction of SPR values, we could further provide early
405 warning of floods in these ungauged watersheds. But this needs more data and
406 examination in future studies.

407 **4.4 Limitations**

408 Previous works usually identify the dominant flood generation mechanism based on the
409 comparison of the timing of events (Berghuijs et al 2016; 2019b; Bloschl et al 2017; Ye
410 et al 2017). Similar work has been implemented in our study watersheds, suggesting
411 the importance of soil moisture and rainfall (Wang et al 2021). Based on that, we further

412 looked into the records to quantitatively evaluate the relative importance of soil
413 moisture and rainfall in flood generation. However, there are limitations in our methods.

414 The precipitation data we used were averaged for the study watersheds from 247
415 meteorological stations. Given the large area and considerable spatial heterogeneity, the
416 precipitation data we used may not always be representative of the actual precipitation
417 events. The daily data could also average the rainfall intensity at hourly scale, which
418 could be influential in small mountainous watersheds. ET was scaled as $0.75 \cdot ET_0$ to
419 make sure it is smaller than the potential evaporation. This is a simplified estimation of
420 ET; more sophisticated method is needed in further analysis on specific catchments at
421 event scale.

422 The estimation of soil moisture is also highly simplified, which cannot be
423 considered as precise estimation at event scale. To reduce the influence from this
424 simplification, we used the percentile of soil moisture to represent the relative saturation
425 of soil moisture as well as the seasonal trend of soil moisture, which was then compared
426 with the percentile of rainfall. While more sophisticated models can be used for soil
427 moisture estimation, there could still be substantial uncertainties (Zaherpour et al., 2018;
428 Ran et al 2020). Yet the seasonal trend and the relative magnitude, after averaging
429 through long-term records would be less impacted by the simplification in estimation
430 (Berghuijs et al 2019; Zhang et al 2019).

431 It is possible that soil moisture at the day before the AMFs may not be the soil
432 moisture before event in large catchments due to the long concentration time. We
433 estimated the concentration time for 10 sites with largest drainage area (larger than
434 100,000 km²): the ones on the main stream and at the outlets of major tributaries
435 following the USBR method (USBR 1973; Gericke & Smithers 2014). The

436 concentration time is mostly within two days for main stream sites and is less than 24hr
437 for sites at the outlets of major tributaries (Table S1). Since the rest of the sites are all
438 smaller than these ones, so would be the concentration time. That is, for the natural
439 watersheds we focused on, the concentration time is likely to be within one day. Thus,
440 the soil moisture at the day before AMFs would contribute to the generation of AMFs,
441 and should be applicable for this study.

442 Besides, the exchange with groundwater was not considered in the soil moisture
443 estimation. The exchange with groundwater is more complicated and heterogenous (i.e.,
444 rivers could receive groundwater recharge in hilly area and recharge groundwater in
445 lower land (Che et al 2021)). According to Huang et al. (2021), the variation of
446 groundwater level in the Yangtze River basin is relatively small. Since the goal of this
447 study is to capture the first order seasonal variation of soil moisture and develop a
448 framework that differentiates the relative importance of precipitation and soil moisture
449 in flood generation, in this study, we estimated the soil moisture following Berhuijs (et
450 al 2016, 2019) with a simple water balance equation.

451 Moreover, this work is focused on the relative importance soil moisture and rainfall,
452 the impact of snowmelt is not considered due to the warm and humid climate in the
453 study watersheds. To apply our findings to cold watersheds with significant impact of
454 snow, the snowmelt component needs to be incorporated. In addition, our method is
455 based on the average values from many years. While previous work indicated that the
456 occurrence of floods in our study watersheds are highly concentrated (Wang et al 2021),
457 there could be strong inter-annual variability in other watersheds. In future studies,
458 annual scale and event scale analysis are needed to examine and improve our findings
459 before it can be applied to watersheds with more diverse climate and landscape

460 conditions. more comprehensive models with sophisticated groundwater component,
461 remote sensing data, and reanalysis product with higher spatial-temporal resolution are
462 needed to provide more accurate estimation of soil moisture, ET, and refine the
463 estimation of the flood-generation SPR.

464 5 Conclusions

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

475 Using the percentile of antecedent soil moisture and rainfall as coordinates, the
476 flood generation mechanism(s) of study watersheds could be grouped into three classes:
477 antecedent soil moisture dominated large flat watersheds, heavy rainfall dominated
478 steep and small to middle size watersheds, and small to middle size watersheds with
479 gentle topographic gradient where floods occurrence requires both highly saturated soil
480 and heavy rainfall. Our analysis further shows that the ratio of relative importance
481 between antecedent soil moisture and rainfall (SPR) can be predicted by topographic
482 wetness index. When the topographic wetness index is large, the dominance of
483 antecedent soil moisture for extreme floods is stronger, and *vice versa*. The SPR also

484 presents negative correlation with area normalized flood peak.

485 With the potential increase of extreme rainfall events (Gao et al., 2016; Chen et
486 al., 2016), upstream mountainous watersheds in the middle and lower Yangtze River
487 basin are facing higher risk of extreme floods. The lack of hydrological records further
488 increases the vulnerability of people in these watersheds. *The flood risks could be
489 reduced by construction of flood control facilities, but it is difficult to set flood control
490 standards in these ungauged watersheds. Our findings provide a framework to
491 quantitatively estimate the possible flood risk for these ungauged watersheds. Based on
492 measurable watershed characteristics (i.e., drainage area and topographic gradient), the
493 flood generation SPR could be derived, which could then be used to estimate the mean
494 annual flood. This information can provide scientific support for flood control
495 management as well as infrastructures construction.*

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

502

503 **Data availability**

504 DEM data was downloaded from Geospatial Data Cloud at <http://www.gscloud.cn/>.
505 Climatological data used in this study was obtained from China Meteorological Data
506 Network, which can be accessed at <http://data.cma.cn/>. Discharge data comes from
507 Annual Hydrological Report of the People's Republic of China issued by Yangtze River
508 Water Resources Commission.

509

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514

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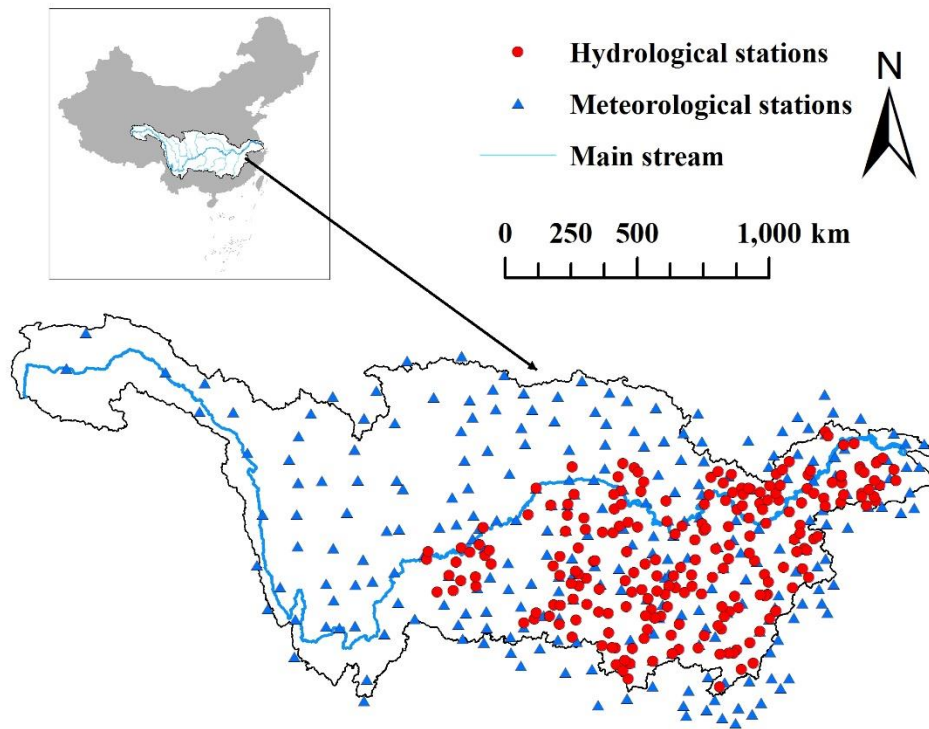
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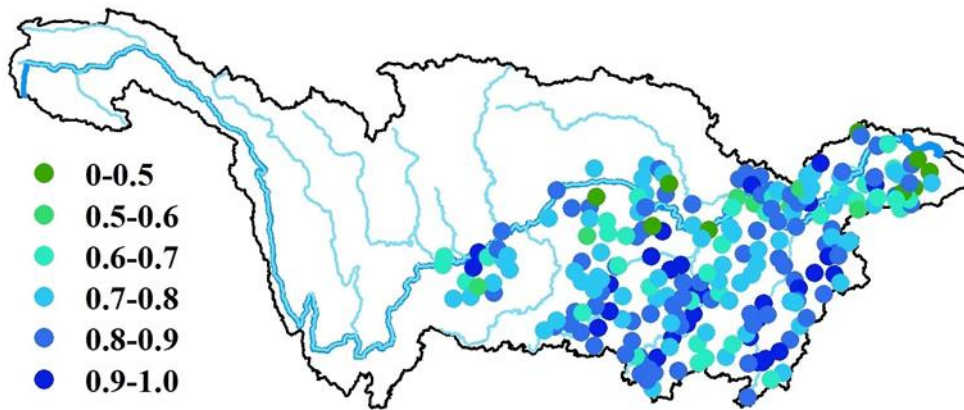
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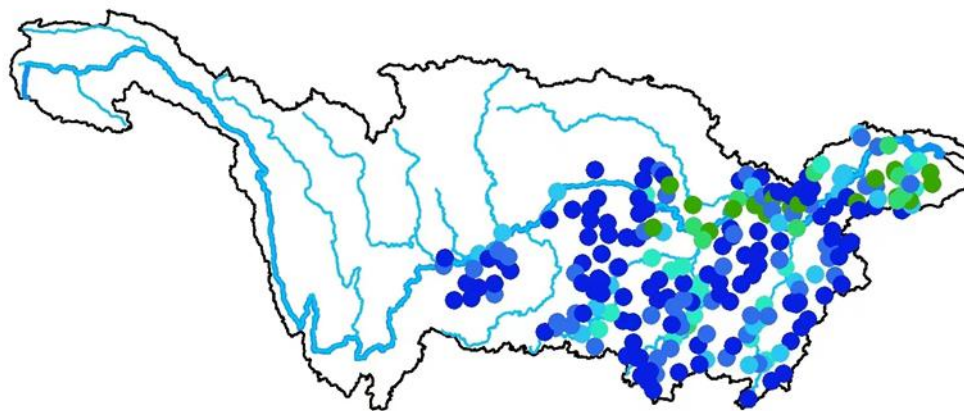
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827 **Figure 1:** Map of the Yangtze River basin, and the meteorological stations and
828 hydrological stations. The blue line is the main stream of Yangtze River.

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(a) Percentile of antecedent soil moisture

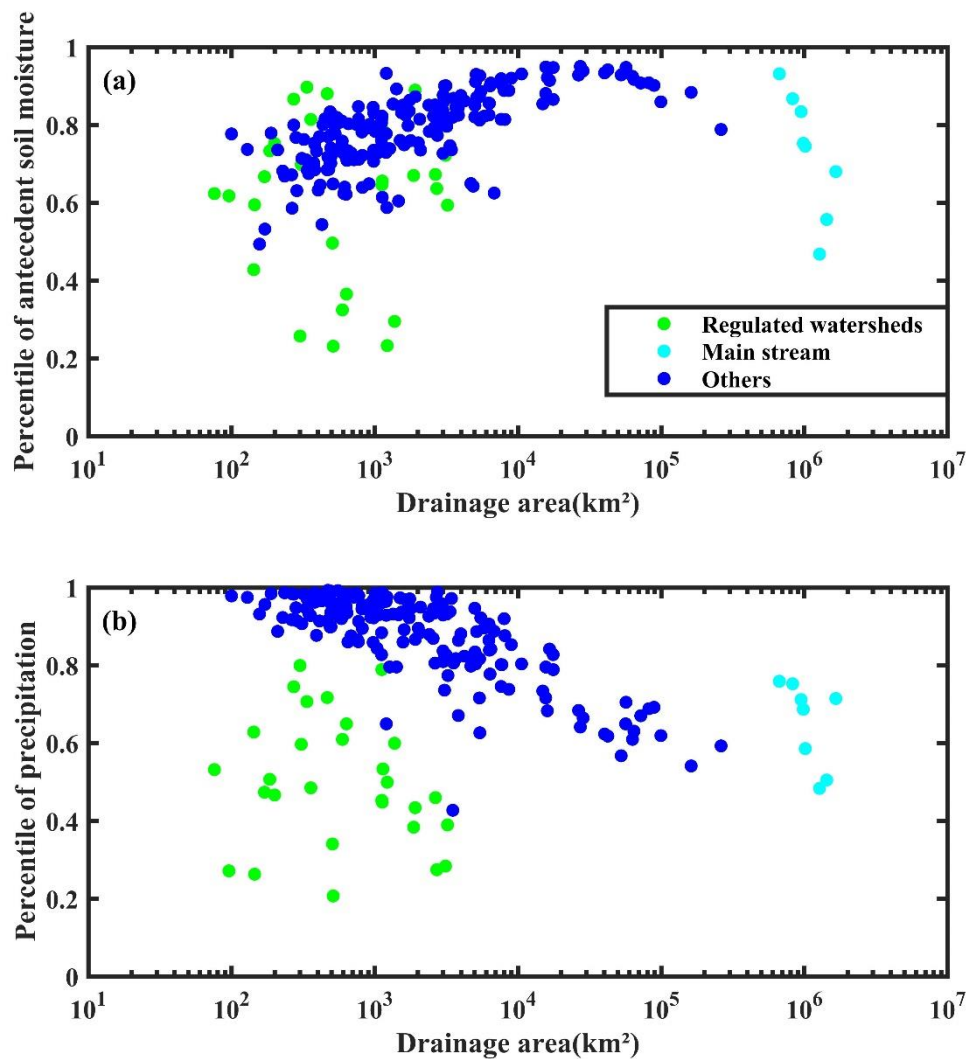


(b) Percentile of precipitation

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831 **Figure 2:** The spatial distribution of (a) the percentile of antecedent soil moisture during
832 annual maximum flood; (b) the percentile of daily precipitation during annual
833 maximum flood.

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836 **Figure 3:** Scatterplot between the drainage area and (a) the percentile of antecedent soil
 837 moisture of AMF events (the linear regression for blue dots: $R^2 = 0.46$, p -value <0.001);

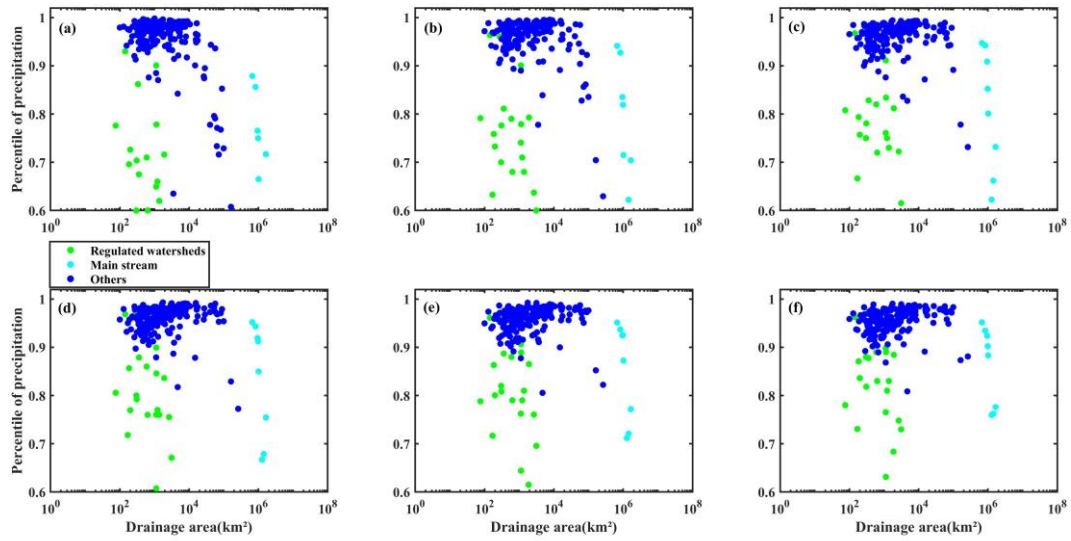
838 (b) the percentile of precipitation at the day of AMF events (the linear regression for
 839 blue dots: $R^2 = 0.61$, p -value <0.001). The green dots represent the regulated watershed,

840 the cyan dots represent the sites on the main stream, and the rest sites are shown in blue.

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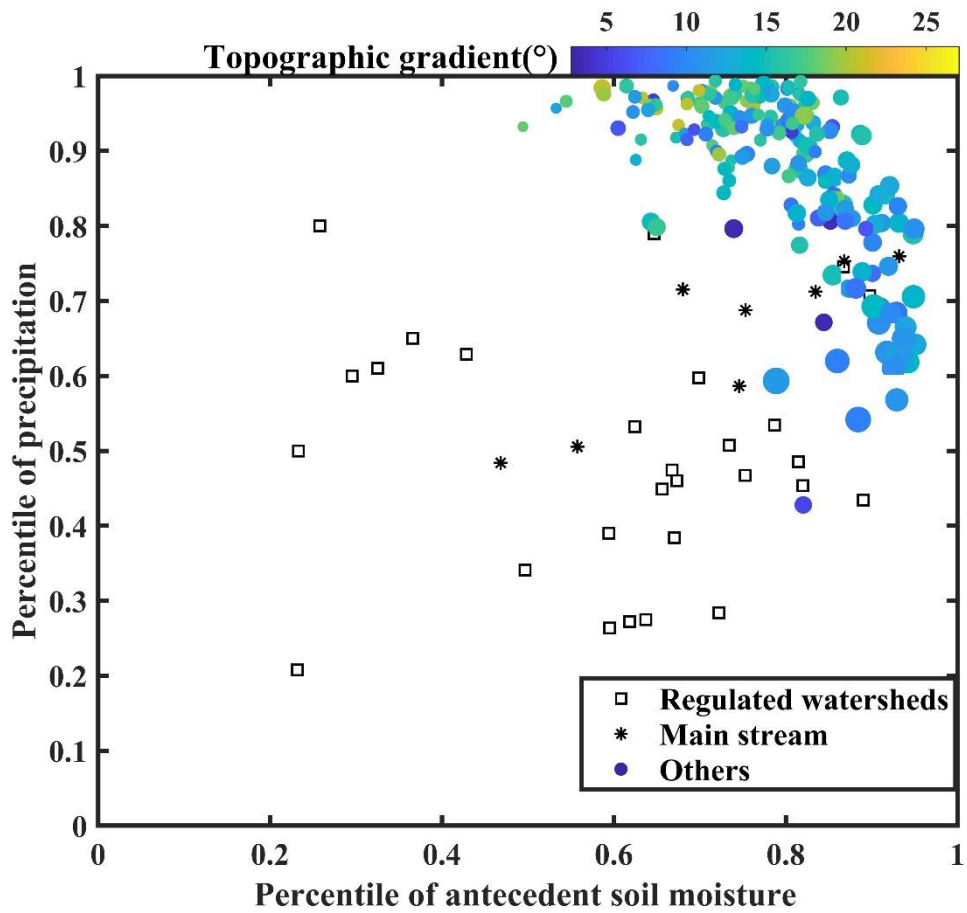
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845 **Figure 4:** Scatterplot between the drainage area and the percentile of accumulated

846 rainfall of (a) two days; (b) three days; (c) four days; (d) five days; (e) six days; and (f)

847 seven days on AMF events.

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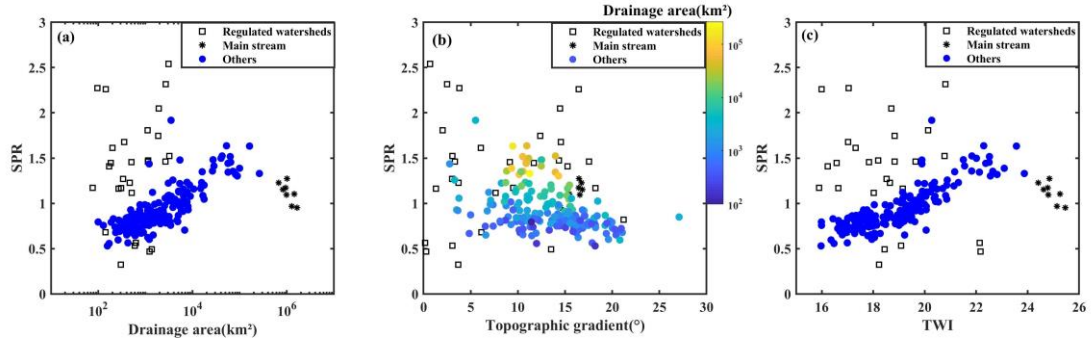
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850 **Figure 5:** Scatterplot of the percentile of precipitation and antecedent soil moisture, the

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

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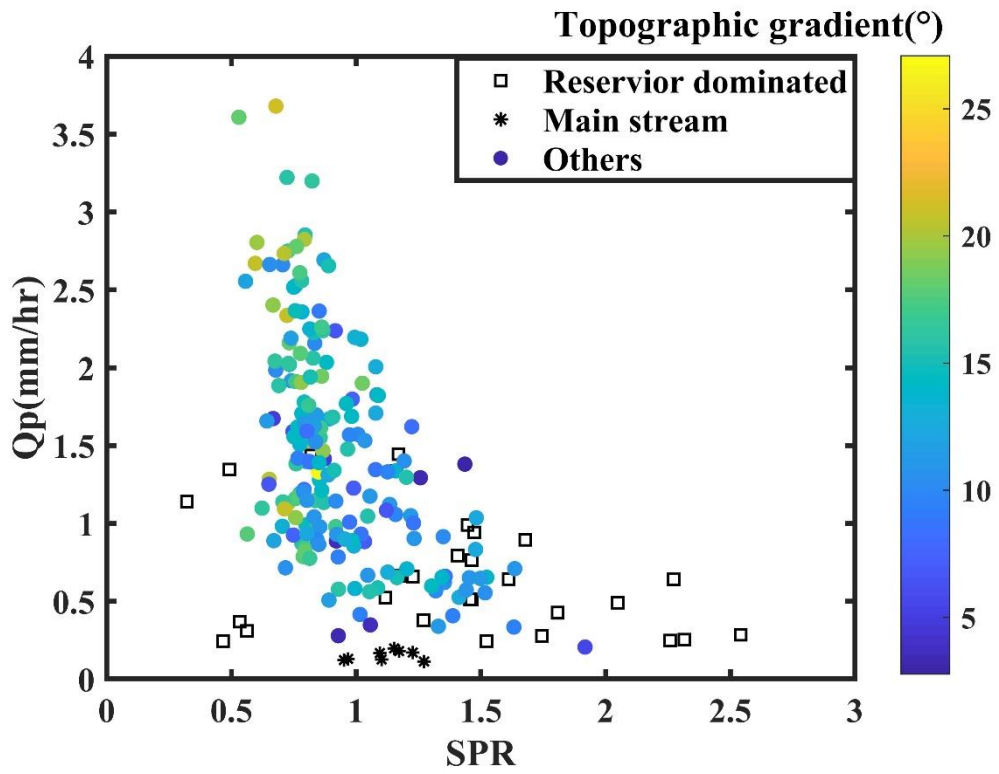


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855 **Figure 6:** Scatterplots between the ratio of antecedent soil moisture and precipitation

856 (SPR) and (a) drainage area; (b) slope; and (c) topographic wetness index (TWI).

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859 **Figure 7:** Scatterplot between the ratio of antecedent soil moisture and precipitation

860 (SPR) and area weighted annual maximum discharge (Q_p), the color represents

861 topographic gradient.

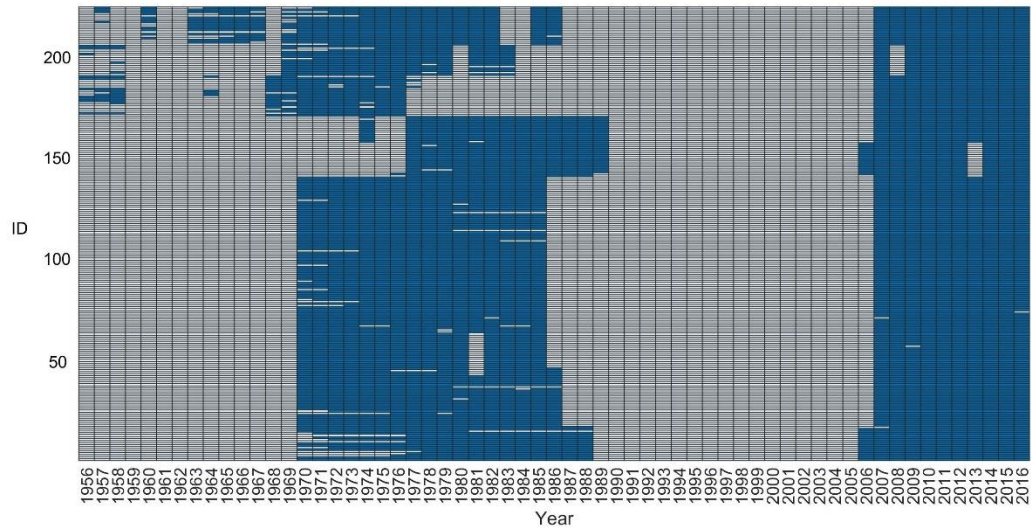
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Supplementary

Table S1: Estimated concentration time for 10 sites with largest drainage area: the ones on main stream (MS) and the ones at the outlets of major tributaries (TR).

Site Name	Concentration Time (hr)	Drainage Area (km ²)
TR-Hukou	17.9	161,979
TR-Chenglingji	18.8	261,986
MS-Zhutuo	32.7	668,661
MS-Cuntan	32.8	827,799
MS-Wanxian	37.6	948,524
MS-Yichang	41.5	982,948
MS-Jianli	45.2	1,014,690
MS-Luoshan	46.3	1,276,676
MS-Hankou	51.0	1,432,008
MS-Datong	54.3	1,657,604

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Figure S1: The data availability of each station, each column indicates each year while each row is corresponding to each station, blue grid indicates there is record of this year.