



1 **DETECTION AND ATTRIBUTION OF FLOOD TRENDS IN MEDITERRANEAN**
2 **BASSINS**

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6 Tramblay, Yves¹

7 Mimeau, Louise¹

8 Neppel, Luc¹

9 Vinet, Freddy³

10 Sauquet, Eric²

11

12 ¹ HSM (Univ. Montpellier, CNRS, IRD), 300 Av. du Professeur Emile Jeanbrau, 34090,
13 Montpellier, France

14 ² IRSTEA, UR RiverLy, Centre de Lyon-Villeurbanne, 5 rue de la Doua CS 20244, 69625
15 Villeurbanne, France

16 ³ GRED (Univ. Paul Valéry, IRD), 2 rue du Pr Henri Serres, 34000 Montpellier, France

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32 **Abstract**

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34 Floods have strong impacts in the Mediterranean region and there is a questioning about a
35 possible increase in their intensity due to climate change. In this study, a large database of 171
36 basins located in South France with daily discharge data with a median record length of 45 years
37 is considered to analyze flood trends and their drivers. In addition to discharge data, outputs of
38 precipitation, temperature, evapotranspiration from the SAFRAN reanalysis and soil moisture
39 computed with the ISBA land surface model are also analyzed. The evolution of land cover in
40 these basins is analyzed using the CORINE database. The trends in floods above the 95th and 99th
41 percentiles are detected by the Mann-Kendall test and quantile regression techniques. The results
42 show that despite the increase in extreme precipitation reported by previous studies, there is no
43 general tendency towards more severe floods. Only for a few basins, the intensity of the most
44 extreme floods is showing significant upward trends. On the contrary, most trends are towards
45 fewer annual flood occurrences above both the 95th and 99th percentiles for the majority of basins.
46 The decrease in soil moisture seems to be an important driver for these trends, since in most
47 basins increased temperature and evapotranspiration associated with a precipitation decreases are
48 leading to a reduction of soil moisture. These results implies that the observed increase in the
49 vulnerability to these flood events in the last decades is mostly caused by human factors such as
50 increased urbanization and population growth rather than climatic factors.

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59 **Keywords:**

60 **Floods, trends, France, Mediterranean, soil moisture**

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63 **1. INTRODUCTION**

64

65 A number of studies have now established that extreme precipitation could increase due to
66 climate change in particular in the Mediterranean (Westra et al., 2013, Polade et al., 2017, Ribes
67 et al., 2018, Trambly and Somot, 2018). Changes in extreme rainfall would be caused by an
68 increase in the precipitable water content in the atmosphere, related to increasing temperatures,
69 according to the principle of Clausius-Clapeyron thermodynamics (Drobinki et al., 2016, Pfahl
70 et al., 2017). Nevertheless, this relationship has a high variability in space, related to temperatures
71 and available humidity (Prein et al., 2016). Several studies observed an increase in the number of
72 dry days associated with increased rainfall intensities, suggesting that dry periods in these areas
73 would become longer, but that precipitation could be more extreme when they occur (Paxian et
74 al., 2015, Polade et al., 2017). Nevertheless, the increase in extreme rainfall would not offset the
75 decrease in precipitation totals, as the drop in cumulative rainfall associated with the decrease in
76 the frequency of low to moderate rainfall is expected to predominate over the gains resulting
77 from the intensification of extreme precipitation (Polade et al., 2017).

78

79 Beside changes in precipitation, an increase in rainfall intensity does not necessarily imply an
80 increase in flood risk (Ivancic and Shaw, 2015, Woldemeskel and Sharma, 2016). Indeed, for a
81 given rainfall accumulation, the runoff coefficient can be very variable in time and space in
82 different basins due to complex interactions between precipitation and infiltration processes on
83 hillslopes which can strongly modulate flood magnitude (Woldemeskel and Sharma, 2016,
84 Wasko and Sharma, 2017, Bennett et al., 2018). Most global studies on flood trend indicate a
85 decrease in flood intensity (Do et al. 2017, Wasko and Sharma, 2017, Sharma et al., 2018). Yet,
86 these trends are highly variable in space for different regions of the globe (Yin et al., 2018, Najibi
87 and Devineni, 2018). The attribution of these trends is rather uncertain, while Yin et al. (2018)
88 relate an increase in floods with increased temperatures; Najibi and Devineni (2018) or Hodgkins
89 et al. (2018) conclude that trends in the flood frequency and duration can be mostly attributed to
90 long-term climate variability. Nonetheless, as noted by Whitfield (2012), flood generating
91 processes do not take place at the global but rather a relatively local scale, making generalizations
92 about flooding in future climates difficult and uncertain. For Mediterranean basins, Blöschl et al.
93 (2017) indicate later winter floods and Mangini et al. (2018) noted a tendency towards increasing



94 flood magnitude and decreasing flood frequency. These findings are consistent with trends
95 detected by Mediero et al. (2014) in Spain and Giuntoli et al. (2012) for the South of France.

96

97 While much work has been done to estimate future climatic conditions, it is not clear about
98 possible changes in hydrological variables including surface conditions that can strongly
99 modulate climatic trends (Knighton et al., 2017). In particular, it is known that in many
100 catchments the initial soil moisture conditions prior to flood events play a key role in flood
101 generation (Brocca et al., 2008, Trambly et al., 2010, Raynaud et al., 2015, Woldemeskel and
102 Sharma, 2016, Wasko and Sharma, 2017, Uber et al., 2018, Wasko and Nathan, 2019) and its
103 temporal change has not been much analyzed up to now. Between two episodes of rain, the base
104 flow of the perennial rivers originates from the draining of the water contained in the soils and for
105 some basins from the aquifers. The capacity of the soil to contain water and restore it to generate
106 runoff depends on its characteristics (texture, structure, porosity ...) but also on the amount of
107 water it already contains at the beginning of a rain episode. Thus, a quasi-saturated soil will not
108 be able to store a lot of water, which, being unable to infiltrate, will contribute directly to runoff.
109 In most cases, there is a non-linear relationship between the flow rate and the initial saturation
110 state of the soil, usually with a threshold value of moisture above which a rapid flow response to
111 a rainy episode is observed (Norbiato et al., 2008, Viglione et al., 2009, Penna et al., 2011).
112 Difference in soil types could induce different relationships between floods and initial conditions
113 (Grillakis et al., 2016, Camarasa-Belmonte, 2016). For intermittent (seasonal runoff only) and
114 ephemeral streams (runoff only after a rain event), the impact of antecedent soil moisture is more
115 complex and strongly dependent on the soil type and geological context (in the presence of karst
116 in particular). In smaller basins, the impact of initial soil moisture content is usually not
117 significant and it increases with catchment size (Zhang et al., 2011).

118

119 For some Mediterranean basins, the increase in heavy rainfall associated with a reduced number
120 of rainy days could decrease the soil water content and therefore increase infiltration capacity,
121 hence reducing runoff. On the other hand, more intense rains in urbanized, impervious areas or
122 on bare soils that are subject to crusting effects could increase runoff and therefore the magnitude
123 of floods. It is therefore necessary to use hydrological or surface models capable of representing
124 these processes. Quintana-Seguí et al. (2011) using the ISBA land surface scheme with different



125 downscaling methods found a future increase in floods corresponding to a 10-year return level in
126 southern French basins, but with different magnitudes depending on the basins. Camici et al.
127 (2017), in a study on the impacts of climate change on floods in central Italy, noted a greater
128 sensitivity of basins with permeable soils to changing climatic conditions. Similarly, Piras et al.
129 (2017) in Sardinia found that impermeable and flat sub-basins are predicted to experience more
130 intense flood events in future scenarios, while more permeable and steep sub-catchments will
131 have an opposite tendency. However, there are systematic differences between projections of
132 changes in flood hazard in south Europe (Italy, Greece, Iberian Peninsula) in most European and
133 global studies using large-scale hydrological models (Kundzewicz et al., 2017). Indeed some
134 studies points towards an increase in southern Europe (Quintana-Seguí et al., 2011, Alfieri et al.,
135 2015) while others suggests a decrease (Donnelly et al., 2017, Thober et al., 2018). This is due to
136 different GCM, RCM, scenarios and downscaling approaches but also the use of large scale
137 hydrological model usually not calibrated and validated for all basins. This type of global (or
138 large scale) hydrological model (LISFLOOD, VIC, HYPE...) is usually not adapted to small
139 river basins less than 500 km², which is the typical catchment size found in the Mediterranean
140 region.

141

142 Prior to make future projections on flood hazard, there is a need to understand the main drivers of
143 changes for floods and the links between floods and climate characteristics (Merz et al., 2014).
144 Indeed, understanding the potential flood drivers and their changes may be more relevant than
145 predictions of uncertain flood changes as noted by Blöschl et al. (2016). The objective of this
146 study is to analyze trends in floods characteristics for a large sample of French Mediterranean
147 basins and to relate these trends to climate and land use dynamics. This is done using statistical
148 tests for the detection of trends and quantile regression models to relate high discharge quantiles
149 to different climatic drivers.

150

151 **2. DATA**

152

153 171 basins located in south France are selected with a minimum of 20 years of daily discharge
154 data. The selection of basins is based on the availability of long time series of daily discharge and
155 the selected basins have no significant human influence on flow, from a previous database



156 elaborated from Sauquet and Catalogne (2011) and Snelder et al. (2013). The median record
157 length is 45 years and 56 stations have more than 50 years of data, more than 100 stations have
158 complete years, with less than 5% missing data, between 1970 and 2010. All the catchments
159 selected have a Mediterranean climate, with a precipitation deficit during summer when the low
160 flows are recorded. These basins are experiencing flash flood events caused by intense rainfall
161 events, corresponding to the only region in France when rainfall can exceed 200 mm/day
162 (<http://pluiesextremes.meteo.fr>) with the maximum occurrence between September and
163 November. Most basins have a catchment area lower than 500 km² and located below 1000 m.
164 (figure 1). The proportion of karstic areas for each basin has been obtained from the BDLISA
165 database (available here: <https://bdlisa.eaufrance.fr/>) which provides a delineation of karst
166 systems in France (Schomburgk et al., 2016). Very common geological formation in the French
167 Mediterranean region, about 50 gauged basins have more than 50% of their catchment areas with
168 carbonaceous superficial formations, indicative of Karstic areas. This means that the rainfall-
169 runoff relationship in this type of basin can be strongly modulated by the presence of karst
170 (Jourde et al., 2007).

171

172 In addition to daily discharge data, different climatic variables have been retrieved from the
173 SAFRAN reanalysis over France (Quintana-Seguí et al., 2008). This reanalysis based on
174 observed station data provides rainfall, snowfall, temperature, actual and reference
175 evapotranspiration for a 8x8km grid over France at the daily time step from 1958 until present.
176 The SAFRAN reanalysis is used to force the ISBA land surface scheme of Météo-France (Habets
177 et al., 20008), to provide among other variables the surface and root zone soil moisture at the
178 same spatial and temporal resolution than ISBA. Trambly et al. (2010) have shown that the soil
179 moisture from the root zone simulated by ISBA is an appropriate indicator of soil moisture prior
180 to flood events in French Mediterranean catchments. The catchment boundaries of the 171 basins
181 selected have been extracted from the HydroSheds database (<https://hydrosheds.org/>) providing
182 flow accumulation and flow direction maps at the 15 arc-second resolution. Then the total
183 precipitation, rainfall, air temperature, actual and reference evapotranspiration from SAFRAN
184 and the surface and root zone soil moisture from ISBA have been extracted and averaged over
185 every catchment.

186



187 The evolution of landcover between 1990 and 2018 in the 171 basins was analyzed using the
188 Corine Landcover inventory (CLC1990 and CLC 2018). Corine Landcover provides an inventory
189 of 44 classes over the European region (Büttner et al., 2002). CLC1990 and CLC2018 are
190 respectively based on Landsat-5 (50m spatial resolution) and Sentinel-2 (10m spatial resolution)
191 satellite images. A limitation of the CLC inventory lies in the difference of accuracy between the
192 CLC1990 and CLC2018 products, which may introduce an uncertainty in the estimation of the
193 evolution of the land cover in the studied basins.

194

195 3. METHODS

196

197 Two approaches are considered to evaluate trends. The first approach, presented in section 3.1
198 thereafter, relies on the Mann-Kendall test applied to the annual number of flood events above
199 two different percentiles, the 95th and the 99th computed on the whole time series and also on the
200 magnitude of these events. Using two different thresholds, which are commonly used for the
201 analysis of floods, allows considering separately the trends on moderate (above the 95th
202 percentile) and more severe (above the 99th percentile) flood events.

203

204 The second approach presented in section 3.2, is based on quantile regression to estimate the
205 temporal trend magnitude in the 95th and 99th percentiles of daily runoff in all stations. The
206 quantile regression method is also used to relate the change in runoff quantiles to changes in
207 climate characteristics, hence providing a way to attribute the observed changes to their potential
208 drivers.

209

210 Hydrological years are considered, starting September 1st and ending August, 31 of the next
211 calendar year. Years with more than 5% missing days are removed. For the first approach based
212 on event characteristics, a de-clustering is required to not include in the flood sample consecutive
213 daily threshold exceedances that belong to the same flood event. A minimum of 2 days between
214 two flood events is selected since it is the average duration of rainstorm in the region (Tramblay
215 et al. 2013). This means, if for two consecutive days the runoff is exceeding the threshold, only
216 the maximum value is retained. Moreover, different values between 1 and 5 days to separate the
217 events have been tested but preliminary tests indicated that it did not change the trend results.



218

219 **3.1 Test for trends and regional significance**

220

221 The Mann–Kendall (MK) test (Mann 1945) is used for the trend detection. Several studies have
222 noted that the presence of serial correlation may affect the results of trend analysis by increasing
223 the variance of the test statistic (Khaliq et al., 2009, Renard et al., 2008). To overcome this
224 limitation, Hamed and Rao (1998) proposed a corrected MK test statistic considering an effective
225 sample size that reflects the effect of serial correlation. This correction was applied in the present
226 study. In addition to the MK test, the method of Sen (1968) is considered to estimate the
227 magnitude of trends. In the present study, trends are considered significant at the 10% level;
228 however, sensitivity tests performed for $p \leq 0.05$, $p \leq 0.01$ revealed very similar spatial trend
229 patterns.

230

231 The significance level α for a statistical test is related to a single test and is no longer valid when
232 multiple tests are conducted (Wilks 2016). When the number of tests being conducted increases,
233 more significant values will be found. The goal of the false discovery rate (FDR) procedure
234 introduced by Benjamini and Hochberg (1995) is to identify a set of at-site significant tests by
235 controlling the expected proportion of falsely rejected null hypotheses that are actually true.
236 Renard et al. (2008), Khaliq et al. (2009) or Wilks (2016) demonstrated that the original FDR is
237 robust to cross correlations between locations and can work with any statistical test for which one
238 can generate a p-value. This method is applied to the MK test results to check if the trends are
239 regionally significant.

240

241 **3.2 Quantile regression**

242

243 As a complementary approach to detect trends in quantiles but also to investigate the relationship
244 between floods and explanatory covariates, the quantile regression (Koenker and Basset, 1978)
245 method is applied. Quantile regression could be seen as the extension of the ordinary least square
246 (OLS) regression (Koenker and Machado, 1999, Villarini and Slater 2017). In OLS, the
247 conditional mean of the response variable is modeled with respect to one or more predictors and
248 the sum of squared errors is minimized. For quantile regression, a conditional quantile of the



249 response variable is modelled as function of predictor(s), an asymmetrically weighted sum of
250 absolute errors is minimized to estimate the slope and intercept terms. In the present work, only
251 linear relationships are considered with one single covariate at a time, while more complex forms
252 of dependences could also be considered in quantile regression. The approach has been
253 previously used to detect trends in extreme precipitation or floods by Villarini and Slater (2017),
254 Yin et al. (2018) or Wasko and Nathan (2019).

255

256 Koenker and Machado (1999) introduced the R^l goodness of fit measure for quantile regression
257 models. As for the R^2 in the case of OLS, R^l lies between 0 and 1. Unlike R^2 , which measures the
258 relative success of two models for the conditional mean function in terms of residual variance, R^l
259 measures the relative success of the corresponding quantile regression models for a specific
260 quantile, by comparison with a restricted model (with slope = 0), in terms of a weighted sum of
261 absolute residuals (see Koenker and Machado, 1999). Consequently, R^l constitutes only a local
262 measure of goodness-of-fit for a particular quantile rather than a global measure over the entire
263 conditional distribution, like R^2 . This measure can help to discriminate between different models
264 using different covariates (ex: precipitation or temperature). Higher R^l values indicate that the
265 model fits better to observations. In this study, this criterion is used to identify the best covariates
266 that could explain the temporal variations in high runoff quantiles.

267

268 **4. RESULTS**

269

270 **4.1 Climatic and land cover trends**

271

272 The climate trends have been analyzed on the whole period of available SAFRAN records,
273 between 1958 and 2018. From figure 2, It can be seen a significant decrease of annual rainfall in
274 56 basins, on average of -20%, accompanied by an increase of the frequency in dry days (with
275 precipitation below 1 mm) for 46 basins. The snowfall is also decreasing in the same proportions
276 (no shown). The sole exception where an increase in rainfall is found is for the Asse River at
277 Beyne-Chabrières on the western foothills of the Alps. This station has long time series spanning
278 from 1983 to 2009, where a +15% trend in annual rainfall is detected over the whole record. Yet,
279 the detection of this trend might be an artefact since there are several consecutive wet years



280 between 1992 and 2000. This trend in rainfall can be also seen for the soil moisture trends.
281 Associated with the precipitation decrease, positive temperature trends are observed for almost all
282 basins, with an average increase of +0.5°C during the time period 1958-2015. Consequently,
283 widespread increasing trends in reference and actual evapotranspiration rates over all basins are
284 observed, similarly as in Vicente-Serrano et al. (2014) in Spain or Rivoire et al. (submitted) for
285 the whole Mediterranean region. The combined decrease in precipitation with increased
286 evapotranspiration yields to a decrease in soil moisture for the surface and the root zone layers.
287 This is in accordance with previous studies over South France such as Vidal et al. (2012) or
288 Dayon et al. (2018).

289

290 About land cover (figure 3), most basins have low urban areas (below 10%) and the basins with
291 the highest coverage are found mostly in the South East. An increase of urban areas up to +20%
292 of total catchment surface can be seen between 1990 and 2018 for basins mostly located close to
293 the Mediterranean coast and in particular in the Provence-Alpes-Côte-d'Azur region. The class
294 representing discontinuous urban fabric represents 73% of artificialized areas and increased by
295 +36% between 1990 and 2018. The increase of urbanized areas could have a strong impact on
296 runoff generation, in particular for small basins, with the increase of impervious surfaces favoring
297 surface runoff. In contrast, the agricultural and forest land cover can reach 100% of the basin
298 surface, in particular in the western Tarn regions for agriculture. We can notice a reduction of
299 forest cover in the Northern Cévennes areas associated with an increase in agricultural surfaces.
300 When looking in details from the original classification, for some catchments of size 500 km² or
301 less, the percentage of vineyards could exceed 70% of the total catchment areas in particular for
302 basins located in the Occitanie region. For almost all basins, the percentage of vineyards has
303 decreased between 1990 and 2018. The other dominant land use classes related to agriculture are
304 pastures (27.8% of all catchments), complex cultivation patterns (21.9%) and land principally
305 occupied by agriculture with significant areas of natural vegetation (27.7%). Forested areas are
306 mostly represented by broad-leaved forest (35%), coniferous forest (19%) and mixed forest
307 (14.4%) classes. It must be noted that the land cover change analysis is hampered by the short
308 duration of the land use maps available, 28 years between 1990 and 2018, and possibly different
309 sensors during this period leading the different attribution to some land use classes.

310



311 **4.2 Flood trends**

312

313 To analyze flood trends, all flood events above the 95th or 99th percentiles of daily runoff
314 computed on the whole time series are extracted. The trend MK test is applied to the number of
315 annual exceedances above these two thresholds and also on the magnitude of the threshold
316 exceedances. From figure 4 it can be seen a general tendency towards a decrease in the annual
317 number of flood events above the 95th percentile, that is significant in 67 catchments, and to a
318 lesser extend also in the number of events above the 99th percentile in 45 catchments. These
319 trends are regionally significant according to the FDR procedure and particularly over the
320 northern ridge of the Cévennes mountainous areas. According to the Sen Slope method to
321 estimate the decrease in the annual number of events above the 95th percentile; for most basins
322 the trends are ranging between -0.5 and -1 event per decade. For the most extreme cases the
323 trends can reach up to -2.5 events per decade. Since for all catchments the number of events
324 above the 95th percentile per year is 4.5 on average (min =2, max =6, after de-clustering), the
325 magnitude of these trends can be considered low. For the 99th percentile the magnitude of trends
326 are similar, with a maximum decrease of -1.4 events per decade, and for most stations on average
327 -0.4 events per decade (with an average annual number of 1.6 events above the 99th percentile,
328 after de-clustering). In addition to the trends in the annual number of events, there is also a weak
329 signal of an increase of the magnitude of floods, in particular above the 99th percentile for 16
330 stations, yet these trends are not regionally significant.

331

332 Beside this event-based analysis, the temporal trends in the 95th and 99th percentiles of the daily
333 runoff time series have been investigated using quantile regression. The approach is
334 complementary but different to the testing of trends on the annual occurrence and the magnitude
335 of the events, since quantile regression allows evaluating the possible changes on the quantiles of
336 daily runoff time series. This analysis reveals that for a majority of catchments, a decreasing
337 trend in these two percentiles is detected. The procedure is to apply a quantile regression of the
338 percentile of interest with time as a covariate, and to validate if the slope of the quantile
339 regression model is significantly different than zero at the 10% level a bootstrap resampling
340 approach (Efron, 1979) has been considered. For the 95th percentile, a decreasing trend in 147
341 stations is found and an increase in only 12 stations. For the 99th percentile, 89 negative trends



342 are found and 15 stations with increasing trends. The relative changes in the 95th and 99th
343 percentiles are ranging for most stations between 0 and -0.5 as shown on figure 5. The number of
344 detected trends with quantile regression for the 95th and 99th percentiles is larger than the number
345 of trends detected with the MK test. However, for many basins the trends in the 95th and 99th
346 percentiles are of small magnitude and only for the largest trends the MK test also detect
347 significant changes in the annual number of events above these thresholds.

348

349 In an attempt to relate the detected trends to catchment characteristics, the Student t-test has been
350 used to compare the catchment descriptors between the group of basins with or without trends.
351 The catchments where decreasing trends in flood occurrence are detected tend to be are larger
352 catchments (mean size of 369 km² vs. 253 km² for the catchments with no significant trends),
353 with a lower proportion of karstic areas (33% vs. 41%) and urban areas (1.7% vs 3.79%). Also
354 more decreasing trends are detected in agricultural catchments than in forested areas. Yet, no
355 clear link can be found between land cover changes and flood trends, probably due to the short
356 duration of the land cover dataset available. The only exception is about trends in urbanization,
357 with a lower increase in urbanization (+0.77% average increase in urban areas) in catchments
358 where floods are decreasing by comparison with catchments with no flood trends (+1.41%
359 average increase in urban areas). It must be noted that there is a strong spatial variability of the
360 observed trends highlighting the complex interplays between the different catchment
361 characteristics, as similarly noted by Snelder et al. (2013) over France. For example, the
362 magnitude of the detected trends is not correlated with the different catchment properties. This
363 implies that it would be very challenging to propose a typology of basins with similar changes in
364 floods according to catchment properties.

365

366 **4.3 Changes in event precipitation and antecedent soil moisture conditions**

367

368 For each event, the cumulative catchment precipitation average is computed as the sum of non-
369 zero consecutive rainy days, on a time window up to 10 days prior to the flood event. The
370 antecedent soil moisture is taken as the root zone soil moisture corresponding to the day prior the
371 start of the rainfall event. Figure 6 show the Mann-Kendall test results for these two indicators for
372 floods above the 95th and the 99th percentiles. An increase of precipitation associated with floods



373 using both thresholds is observed (for 34 catchments for the 95th percentile and 36 catchments for
374 the 99th percentile), associated with a decrease in antecedent soil moisture conditions prior to
375 floods in up to 40 catchments for floods above the 95th percentile. There is a correlation between
376 the reduction of antecedent soil moisture prior to flood events and the decrease of the annual
377 number of flood events above the 95th percentile ($r=0.44$), also to a lesser extent for the number
378 of floods above the 99th percentile ($r=0.34$). Consequently, as observed in Australia by Wasko
379 and Nathan (2019) it can be hypothesized that the decrease of antecedent soil moisture is an
380 important driver leading to the reduction of the annual number of floods, despite the increase in
381 event precipitation already pointed out by several studies in this region (Tramblay et al., 2013,
382 Ribes et al., 2018, Blanchet et al., 2018). Indeed, for 12 catchments an increase of event rainfall
383 is detected when for the same catchments a decrease in the annual number of events above the
384 95th percentile is also observed. It is also the case of 11 catchments for the events above the 99th
385 percentile with an increase of event rainfall accompanied by a decrease in the annual number of
386 events. However as shown before, the increased event precipitation for several basins is probably
387 the cause of higher flood magnitudes for the most severe events (above the 99th percentile).

388

389 **4.3 Explanatory covariates for high runoff quantiles**

390

391 To test the influence of different covariates on the variation of the 95th and 99th percentile values,
392 quantile regression models using time, temperature, soil moisture from the root zone, actual
393 evapotranspiration (AE), reference evapotranspiration (ET₀) and precipitation have been
394 compared. The goal here is not to select the best covariates for each station but to identify
395 relevant covariates at the regional scale. Since climatic covariates could influence the
396 hydrological response at different time scales (Mediero et al., 2014, Villarini and Slater 2018,
397 Wasko and Nathan, 2019), three different aggregation periods to compute moving averages have
398 been compared. For the event scale, the different covariates have been averaged with a 3-day
399 time lag preceding each event. At the monthly time scale representing the seasonal variability, the
400 covariates have been averaged in the same manner but on 30 days. Finally, for the annual time
401 scale the covariates have been averaged over 365 days. At the event scale, the precipitation rather
402 represents the intensity of rainfall during the event than the preceding soil moisture. On the
403 opposite, for the monthly and annual aggregation periods the precipitation is here a proxy for soil



404 moisture and its long term variability. To test which covariate provides the best reproduction of
405 the observed 95th and 99th percentiles of the daily discharge time series, the R^I metric is
406 computed, for each covariate, between the quantile regression model built with the covariate and
407 a constrained model with a constant slope (0).

408

409 The results are plotted on figure 9. A similar pattern can be seen for both percentiles, with
410 decreasing R^I values for longer time aggregation periods for the covariates. For the event-scale,
411 both precipitations and soil moisture are outperforming other covariates, including time. The
412 same results are found for the annual time scale, yet with a different interpretation because annual
413 precipitation is representing the average level of soil moisture storage rather than event rainfall.
414 The link observed between the 95th and 99th percentiles with annual precipitation or soil moisture
415 is an indication that the long-term decrease observed for these two variables (figure 2) could be
416 the cause of the observed decrease in the frequency of floods above these two percentiles. At the
417 monthly time scale, the cumulative precipitation plays the most important role when the effects of
418 soil moisture, actual evapotranspiration and temperature are similar. For almost all covariates,
419 there is an improvement by comparison to the quantile regression model using time only.

420

421 Overall, the R^I coefficients are decreasing with increasing slopes and basin mean attitude.
422 However, these two variables are correlated ($r=0.61$). This is an indication that antecedent soil
423 moisture condition may have a lower influence on flood generation in mountainous areas,
424 probably due to shallower soils and steeper slopes. For event based soil moisture and
425 precipitation, there is an inverse relationship with basin size: for small basins (less than 500km²)
426 event soil moisture and precipitation are very good predictors for the time variations of the 95th
427 and the 99th percentiles, with R^I values up to 0.6, when for larger basins the R^I values are much
428 lower (about 0.1 to 0.2). When averaged at the monthly or annual time step, the relation is
429 opposite with a larger influence of soil moisture and antecedent precipitation for larger basins
430 with higher R^I coefficients. This finding is fully consistent with results obtained for different
431 regions of the globe (Zhang et al., 2011, Ivancic and Shaw, 2015, Woldemeskel and Sharma,
432 2016, Wasko and Sharma, 2017), highlighting the buffering effects of large basins with the
433 capacity to store more water than smaller basins.

434



435 **5. CONCLUSIONS**

436

437 The results obtained in the present study show that despite the increase in extreme precipitation
438 events reported by previous studies over the same domain (Ribes et al., 2018) there is not a
439 general increase in flood occurrence. Only for a few basins, the intensity of the most extreme
440 floods is showing significant upward trends. On the contrary, a global tendency towards fewer
441 annual flood occurrences is observed for events of moderate intensity, above the 95th percentile.
442 The same signal, with a lower magnitude, is also seen for higher floods above the 99th percentile.
443 Overall, there are much more trends detected for the annual occurrence of floods than for their
444 intensity. It should be also emphasized that the magnitude of these trends remains moderated,
445 with only a few events less by decade and consequently these trends are only noticeable over
446 long time periods. The decrease in soil moisture seems to be an important driver for these
447 detected changes, indeed in all basins an increase of temperature and evapotranspiration
448 associated with a decrease in precipitation is leading to a reduction of soil moisture over time. For
449 several basins, the soil moisture decrease can offset the increase in extreme precipitation and
450 generate less frequent floods. These changes are mostly observed for larger agricultural basins,
451 with low urbanization and karstic areas. Wasko and Sharma et al. (2017) previously noted the
452 importance of catchment size for the influence of soil moisture on flood runoff due to higher
453 potential of soil moisture storage. The trends detected in the present work are consistent with
454 those found in other Mediterranean regions such as Spain (Mediero et al., 2014) and Australia
455 (Wasko and Nathan, 2019). An important finding of the present work is that with the same large
456 scale climatic drivers (in terms of temperature, evapotranspiration and precipitation) the flood
457 trends in the basins can be different. This shows the importance of basins characteristics to buffer
458 climatic variability. Indeed, even if similar patterns of changes in the 95th and 99th percentiles are
459 found, the analysis of individual catchments is revealing spatial differences even for neighboring
460 basins caused by different topography, soil and land cover combinations. This is a factual
461 demonstration of the commentary of Whitfield (2012) stating that it would be very difficult, if
462 not scientifically irrelevant, to make general statements about the plausible future evolution of
463 flood risk.

464



465 These results showing a lack of a generalized upward trend in floods should be put into
466 perspective with the observed increase in the vulnerability to these episodes. Indeed many reports
467 such as Llasat et al (2013) indicate an increase in the number of floods inducing damages
468 between 1981 and 2010 in South France and North Spain, which they attribute to an increased
469 vulnerability and land use changes. The French Mediterranean regions are concentrating 66% of
470 the total cost of flood damage to private properties in France (Vinet, 2011) and the total assets
471 lost due to floods are rising as in many other regions (CCR, 2018, Paprotny et al., 2018). The
472 areas close to the Mediterranean have seen a population increase and an extension of urbanized
473 areas, driven in part but not solely by the increase of touristic activities (Vinet, 2011, Vinet and
474 De Richemond, 2017). Bouwer (2011) concluded after a review of 22 disaster loss studies that
475 there is no trends in flood losses, corrected for changes (increases) in population and capital at
476 risk, which could be attributed to anthropogenic climate change”. Therefore, it can be concluded
477 that, at least for Southern France, as noted previously by Neppel et al. (2003) the increasing cost
478 of damages caused by floods is rather due to the increase in socio-economic vulnerability rather
479 than a climate change signal towards an increase in the severity of floods. Nonetheless, the
480 evolution of flood frequency and intensity is a key question for risk prevention. Flood related
481 mortality in the Mediterranean basin is conditioned both by hazards drivers (rainfall intensity,
482 discharge...) but also by social drivers (behaviors, characteristics of buildings...) as shown in
483 different studies (Ruin et al., 2008, Vinet, 2011, Boudou et al., 2016). Deeper knowledge in
484 rainfall and flood trends must be crossed with exposure (e.g. population in flood prone zones) and
485 vulnerability data (e. g. elderying of population in the future) to anticipate evolution in human
486 mortality in relation with flash floods in the Mediterranean basin (Petrucci et al. 2017). As
487 pointed out in previous research projects (Merz et al., 2014, Meyer et al., 2014) there is a need to
488 integrate climate change scenarios with socio-economic change scenarios to better quantify
489 changes in flood risk. To achieve this task, it is necessary to develop databases on vulnerability
490 and exposure to be analyzed in conjunction with hydrometeorological data (Saint-Martin et al.,
491 2018).

492

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494



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497 are made available to the research community upon request.

498

499 **References**

500

501 Alfieri, L., Burek, P., Feyen, L., and Forzieri, G.: Global warming increases the frequency of river floods
502 in Europe, *Hydrol. Earth Syst. Sci.*, 19, 2247–2260, <https://doi.org/10.5194/hess-19-2247-2015>, 2015.

503

504 Benjamini, Y. and Hochberg, Y.: Controlling the false discovery rate: A practical and powerful
505 approach to multiple testing, *J. Roy. Stat. Soc. B*, 57, 289–300, 1995.

506

507 Bennett B., Leonard, M., Deng Y., and Westra, S.: An empirical investigation into the effect of antecedent
508 precipitation on flood volume, *J. Hydrol.*, 567, 435–445, 2018.

509

510 Blanchet, J., Molinié, G., and Touati, J.: Spatial analysis of trend in extreme daily rainfall in southern
511 France, *Clim Dyn.*, 51, 799–812, 2018.

512

513 Blöschl, G., Gaál, L., Hall, J., Kiss, A., Komma, J., Nester, T., et al.: Increasing river floods: fiction or
514 reality? *WIREs Water*, 2, 329–344, 2015.

515

516 Blöschl, G., Hall, J., Parajka, J., Perdigão, R.A., Merz, B., Arheimer, B., et al.: Changing climate shifts
517 timing of European floods, *Science*, 357, 588–590, 2017.

518

519 Boudou, M., Lang, M., Vinet, F., and Cœur, D.: Comparative hazard analysis of processes leading to
520 remarkable flash floods (France, 1930–1999), *J. Hydrol.*, 541, 533–552, 2016.

521

522 Bouwer, L. M.: Have disaster losses increased due to anthropogenic climate change? *Bull. Am. Met. Soc.*
523 92 (1), 39–46, 2011.

524

525 Brocca, L., Melone, F., and Moramarco, T.: On the estimation of antecedent wetness conditions in
526 rainfall–runoff modelling, *Hydrol. Processes*, 22, 629–642, 2008.

527



- 528 Camarasa-Belmonte, A.M.: Flash floods in Mediterranean ephemeral streams in Valencia Region, J.
529 Hydrol., 541, 99–115, 2016.
- 530
- 531 Camici, S., Brocca, L., and Moramarco, T.: Accuracy versus variability of climate projections for flood
532 assessment in central Italy, *Climatic Change* 141(2), 273–286, 2017.
- 533
- 534 CCR 2018. Conséquences du changement climatique sur les coûts des catastrophes naturelles en France à
535 Horizon 2050.
536 [https://www.ccr.fr/documents/23509/29230/Etude+Climatique+2018+version+complete.pdf/6a7b6120-](https://www.ccr.fr/documents/23509/29230/Etude+Climatique+2018+version+complete.pdf/6a7b6120-7050-ff2e-4aa9-89e80c1e30f2)
537 [7050-ff2e-4aa9-89e80c1e30f2](https://www.ccr.fr/documents/23509/29230/Etude+Climatique+2018+version+complete.pdf/6a7b6120-7050-ff2e-4aa9-89e80c1e30f2)
- 538
- 539 Dayon, G., Boé, J., Martin, E., and Gailhard, J.: Impacts of climate change on the hydrological cycle over
540 France and associated uncertainties, *Comptes Rendus Geosciences*, 350(4), 141–153, 2018.
- 541
- 542 Do, H.X., Westra, S., and Leonard, M.: A global-scale investigation of trends in annual maximum
543 streamflow, *J. Hydrol.*, 552, 28–43, 2017.
- 544
- 545 Donnelly, C., Greuell, W., Andersson, J., Gerten, D., Pisacane, G., Roudier, P., and Ludwig, F.: Impacts
546 of climate change on European hydrology at 1.5, 2 and 3 degrees mean global warming above
547 preindustrial level, *Climatic Change*, 19, 1–14, 2017.
- 548
- 549 Efron, B.: Bootstrap Methods: Another Look at the Jackknife, *Ann. Stat.*, 7, 1–26, 1979.
- 550
- 551 Giuntoli, I., Renard, B., and Lang, M.: Floods in France. In: Z.W. Kundzewicz, ed. *Changes in flood risk*
552 *in Europe*. Wallingford, UK: IAHS and CRC/Balkema, IAHS Special Publ. 10, 212–224, 2012.
- 553
- 554 Grillakis, M.G., Koutroulis, A.G., Komma, J., Tsanis, I.K., Wagner, W., and Blöschl, G.: Initial
555 soil moisture effects on flash flood generation - A comparison between basins of contrasting
556 hydro-climatic conditions, *Journal of Hydrology* 541, 206–217, 2016.
- 557
- 558 Habets, F., Boone, A., Champeaux, J.-L., Etchevers, P., Franchis-teguy, L., Leblois, E., Ledoux, E., Le
559 Moigne, P., Martin, E., Morel, S., Noilhan, J., Quintana-Segui, P., Rousset-Regimbeau, F., and Viennot,
560 P.: The SAFRAN-ISBA-MODCOU hydrometeorological model applied over France, *J. Geophys. Res.*,
561 113, D06113, doi:10.1029/2007JD008548, 2008.



562
563 Hamed, K.H. and Rao, A.R.: A modified Mann-Kendall trend test for autocorrelated data, *J. Hydrol.*, 204,
564 182–196, 1998.
565
566 Hodgkins, G. A., Whitfield, P. H., Burn, D. H., Hannaford, J., Re-nard, B., Stahl, K., Fleig, A. K.,
567 Madsen, H., Mediero, L., Ko-rhonen, J., Murphy, C., and Wilson, D.: Climate-driven variability in the
568 occurrence of major floods across North America and Europe, *J. Hydrol.*, 552, 704–717, 2017.
569
570 Ivancic, T.J., Shaw S.B.: Examining why trends in very heavy precipitation should not be mistaken for
571 trends in very high river discharge, *Climatic Change*, 133, 681–693, 2015.
572
573 Jourde, H., Roesch, A., Guinot, V. and Bailly-Comte, V.: Dynamics and contribution of karst
574 groundwater to surface flow during Mediterranean flood, *Environmental Geology*, 51, 725–730, 2007.
575
576 Khaliq, M.N., Ouarda, T.B.M.J, Gachon, P., Sushama, L., and St-Hilaire, A.: Identification of
577 hydrological trends in the presence of serial and cross correlations: A review of selected methods and
578 their application to annual flow regimes of Canadian rivers, *J. Hydrol.*, 368, 117–130, 2009.
579
580 Knighton, J.O., DeGaetano, A., and Walter, M.T.: Hydrologic state influence on riverine flood discharge
581 for a small temperate watershed (Fall Creek, United States): negative feedbacks on the effects of climate
582 change, *J. Hydrometeorol.*, 18(2), 431–449, 2017.
583
584 Koenker, R. and Basset, B.G.: Regression quantiles, *Econometrica*, 46(1), 33–50, 1978.
585
586 Koenker, R., Machado J.A.F.: Goodness-of-fit and related inference processes for quantile regression,
587 *Journal of the American Statistical Association*, 94(448), 1296–1310, 1999.
588
589 Kundzewicz, Z.W., Krysanova, V., Dankers, R., Hirabayashi, Y., Kanae, S., Hattermann, F.F., Huang, S.,
590 Milly, P.C.D., Stoffel, M., Driessen, P.P.J., Matczak, P., Quevauviller, P., and Schellnhuber, H.-J.:
591 Differences in flood hazard projections in Europe – their causes and consequences for decision making,
592 *Hydrological Sciences Journal*, 62(1), 1–14, 2017.
593



- 594 Llasat, M.C., Llasat-Botija, M., Petrucci, O., Pasqua, A. A., Rosselló, J., Vinet, F., and Boissier, L.:
595 Towards a database on societal impact of Mediterranean floods within the framework of the HYMEX
596 project, *Nat. Hazards Earth Syst. Sci.*, 13, 1337-1350, 2013.
597
- 598 Mann, H. B.: Nonparametric tests against trend, *Econometrica*, 13,245– 259, 1945
599
- 600 Mangini, W., Viglione, A., Hall, J., Hundecha, Y., Ceola, S., Montanari, A., Rogger, M., Salinas, J.L.,
601 Borzì, I., and Parajka, J.: Detection of trends in magnitude and frequency of flood peaks across Europe,
602 *Hydrological Sciences Journal*, 63(4), 493-512, 2018.
603
- 604 Mediero, L., Santillán, D., Garrote, L., and Granados, A.: Detection and attribution of trends in magnitude,
605 frequency and timing of floods in Spain, *J. Hydrol.*, 517, 1072–1088, 2014.
606
- 607 Merz, B., Aerts, J., Arnbjerg-Nielsen, K., Baldi, M., Becker, A., Bichet, A., Blöschl, G., Bouwer, L. M.,
608 Brauer, A., Cioffi, F., Delgado, J. M., Gocht, M., Guzzetti, F., Harrigan, S., Hirschboeck, K., Kilsby, C.,
609 Kron, W., Kwon, H.-H., Lall, U., Merz, R., Nissen, K., Salvatti, P., Swierczynski, T., Ulbrich, U.,
610 Viglione, A., Ward, P. J., Weiler, M., Wilhelm, B., and Nied, M.: Floods and climate: emerging
611 perspectives for flood risk assessment and management, *Nat. Hazards Earth Syst. Sci.*, 14, 1921-1942,
612 2014.
613
- 614 Meyer, V., Becker, N., Markantonis, V., Schwarze, R., van den Bergh, J. C. J. M., Bouwer, L. M.,
615 Bubeck, P., Ciavola, P., Genovese, E., Green, C., Hallegatte, S., Kreibich, H., Lequeux, Q., Logar, I.,
616 Papyrakis, E., Pfuertscheller, C., Poussin, J., Przyluski, V., Thielen, A. H., and Viavattene, C.: Review
617 article: Assessing the costs of natural hazards – state of the art and knowledge gaps, *Nat. Hazards Earth*
618 *Syst. Sci.*, 13, 1351-1373, <https://doi.org/10.5194/nhess-13-1351-2013>, 2013.
619
- 620 Najibi, N. and Devineni, N.: Recent trends in the frequency and duration of global floods, *Earth Syst.*
621 *Dynam.*, 9, 757-783, 2018.
622
- 623 Neppel L., Bouvier C., Desbordes M., and Vinet F.: A possible origin for the increase in floods in the
624 Mediterranean region. *Revue des sciences de l'eau* 16(4), 389-494, 2003.
625



- 626 Norbiato, D., Borga, M., Esposti, S.D., Gaume, E., and Anquetin, S.: Flash flood warning based on
627 rainfall thresholds and soil moisture conditions: An assessment for gauged and ungauged basins, *Journal*
628 *of Hydrology*, 362, 274-290, 2008.
- 629
- 630 Penna, D., Tromp-van Meerveld, H.J., Gobbi, A., Borga, M., and Dalla Fontana, G.: The influence of soil
631 moisture on threshold runoff generation processes in an alpine headwater catchment, *Hydrology and Earth*
632 *System Sciences*, 15, 689-702, 2011.
- 633
- 634 Paprotny, D., Sebastian, A., Morales-Nápoles, O., and Jonkman, S. N.: Trends in flood losses in Europe
635 over the past 150 years, *Nature Communications*, 9(1),1985, doi:10.1038/s41467-018-04253-1, 2018.
- 636
- 637 Petrucci, O., Papagiannaki, K., Aceto, L., Boissier, L., Kotroni, V., Grimalt, M., Llasat, M. C.,
638 Llasat-Botija, M., Rosselló, J., Pasqua, A. A., and Vinet, F.: MEFF: The database of MEDiter-ranean
639 Flood Fatalities (1980 to 2015), *J. Flood Risk Manage.*, e12461, <https://doi.org/10.1111/jfr3.12461>, 2018.
- 640
- 641 Piras, M., Mascaro, G., Deidda, R., and Vivoni, E.R.: Impacts of climate change on precipitation and
642 discharge extremes through the use of statistical downscaling approaches in a Mediterranean basin,
643 *Sci. Total Environ.*, 543, 952–964, 2016.
- 644
- 645 Polade, S.D., Gershunov, A., Cayan, D.R., Dettinger, M.D., and Pierce, D.W.: Precipitation in a warming
646 world: Assessing projected hydro-climate changes in California and other Mediterranean climate regions,
647 *Scientific Reports* 7, 10783, doi:10.1038/s41598-017-11285-y, 2017.
- 648
- 649 Quintana-Seguí, P., Habets, F., and Martin, E.: Comparison of past and future Mediterranean high and low
650 extremes of precipitation and river flow projected using different statistical downscaling methods, *Nat.*
651 *Hazards Earth Syst. Sci.*, 11, 1411-1432, 2011.
- 652
- 653 Quintana-Seguí, P., Le Moigne, P., Durand, Y., Martin, E., Habets, F., Baillon, M., Canellas, C.,
654 Franchisteguy, L., and Morel, S.: Analysis of Near-Surface Atmospheric Variables : Validation of the
655 SAFRAN Analysis over France, *J. Appl. Meteor. Climatol.*, 47, 92-107, 2008.
- 656
- 657 Raynaud, D., Thielen, J., Salamon, P., Burek, P., Anquetin, S., and Alfieri, L.: A dynamic runoff
658 coefficient to improve flash flood early warning in Europe: validation on the 2013 Central Euro-pean
659 floods in Germany. *Met. Apps*, 22: 410-418, 2015.



660
661 Renard, B., Lang, M., Bois, P., Dupeyrat, A., Mestre, O., Niel, H., Sauquet, E., Prudhomme, C., Parey, S.,
662 Paquet, E., Neppel, L., and Gailhard, J.: Regional methods for trend detection: assessing field significance
663 and regional consistency, *Water Resour. Res.*, 44, W08419, doi:10.1029/2007WR006268, 2008.
664
665 Ribes, A., Soulihanh, T., Vautard, R., Dubuisson, B., Somot, S., Colin, J., Planton, S., and Soubeyroux, J-
666 M. : Observed increase in extreme daily rainfall in the French Mediterranean. *Clim Dyn.* 52, 1095-1114,
667 2019.
668
669 Rivoire, P., Tramblay, Y., Neppel, L., Hertig, E., and Vicente-Serrano, S. M.: Impact of the dry day
670 definition on Mediterranean extreme dry spells analysis, *Nat. Hazards Earth Syst. Sci. Discuss.*,
671 <https://doi.org/10.5194/nhess-2019-31>, in review, 2019.
672
673 Ruin, I., Creutin, J.-D., Anquetin, S. and Lutoff, C.: Human exposure to flash-floods - Relation between
674 flood parameters and human vulnerability during a storm of September 2002 in southern France,
675 *J. Hydrol.*, 361, 199-213, 2008.
676
677 Saint-Martin, C., Javelle, P., and Vinet, F.: DamaGIS: a multisource geodatabase for collection of flood-
678 related damage data, *Earth Syst. Sci. Data*, 10, 1019-1029, <https://doi.org/10.5194/essd-10-1019-2018>,
679 2018.
680
681 Sauquet, E. and Catalogne, C.: Comparison of catchment grouping methods for flow duration curve
682 estimation at ungauged sites in France, *Hydrol. Earth Syst. Sci.*, 15, 2421-2435,
683 <https://doi.org/10.5194/hess-15-2421-2011>, 2011.
684
685 Sen, P.K.: Estimates of the regression coefficient based on Kendall's tau, *J. Am. Stat. Assoc.*,
686 63, 1379-1389, 1968
687
688 Schomburgk S., Allier D., and Seguin J.J.: The new aquifer Reference system BDLISA in France and the
689 representation of karst units : challenges of small-scale mapping, in *Grundwasser - Mensch - Ökosysteme*.
690 25. Tagung des Fachsektion Hydrogeologie in der DGGV 2016, Karlsruher Institut für Technologie
691 (KIT), 13. -17. April 2016, Germany. KIT Scientific Publishing. n° ISBN : 978-3-7315-0475-7, 2016.
692



- 693 Sharma, A., Wasko, C., & Lettenmaier, D.P.: If precipitation extremes are increasing, why aren't floods?
694 Water Resources Research, 54, 8545–8551, 2018.
- 695
- 696 Snelder, T. H., Datry, T., Lamouroux, N., Larned, S. T., Sauquet, E., Pella, H., and Catalogne, C.:
697 Regionalization of patterns of flow intermittence from gauging station records, Hydrol. Earth Syst. Sci.,
698 17, 2685-2699, <https://doi.org/10.5194/hess-17-2685-2013>, 2013.
- 699
- 700 Thober, S., Kumar, R., Wanders, N., Marx, A., Pan, M., Rakovec, O., Samaniego, L., Sheffield, J.,
701 Wood, E. F., and Zink, M.: Multi-model ensemble projections of European river floods and high flows
702 at 1.5, 2, and 3 degree global warming, Environ. Res. Lett., 13, 1–22, 2018.
- 703
- 704 Trambly Y., Somot S.: Future evolution of extreme precipitation in the Mediterranean, Climatic Change
705 151(2), 289–302, 2018.
- 706
- 707 Trambly, Y., Neppel, L., Carreau, J., and Najib, K.: Non-stationary frequency analysis of heavy rainfall
708 events in southern France, Hydrol. Sci. J., 58, 1–15, 2013.
- 709
- 710 Trambly, Y., Bouvier, C., Martin, C., Didon-Lescot, J.F., Todorovik, D., and Domergue, J. M.:
711 Assessment of initial soil moisture conditions for event-based rainfall-runoff modelling, J. Hydrol., 387,
712 176-187, 2010.
- 713
- 714 Uber, M., Vandervaere, J.-P., Zin, I., Braud, I., Heistermann, M., Legouët, C., Molinié, G., and Nord, G.:
715 How does initial soil moisture influence the hydrological response? A case study from southern France,
716 Hydrol. Earth Syst. Sci., 22, 6127-6146, <https://doi.org/10.5194/hess-22-6127-2018>, 2018.
- 717
- 718 Vicente-Serrano, S.M., Azorin-Molina, C., Sanchez-Lorenzo, A., Revuelto, J., López-Moreno, J.I.,
719 González-Hidalgo, J.C., and Espejo, F.: Reference evapotranspiration variability and trends in Spain,
720 1961–2011. Global and Planetary Change, 121, 26–40, 2014.
- 721
- 722 Vidal, J.-P., Martin, E., Kitova, N., Najac, J., and Soubeyroux, J.-M.: Evolution of spatio-temporal
723 drought characteristics: validation, projections and effect of adaptation scenarios, Hydrol. Earth Syst. Sci.,
724 16, 2935-2955, <https://doi.org/10.5194/hess-16-2935-2012>, 2012.
- 725



- 726 Viglione, A., Merz, R., and Blöschl, G.: On the role of the runoff coefficient in the mapping of rainfall to
727 flood return periods, *Hydrol. Earth Syst. Sci.*, 13, 577-593, <https://doi.org/10.5194/hess-13-577-2009>,
728 2009.
729
- 730 Villarini G. and Slater L.: Examination of Changes in Annual Maximum Gauge Height in the Continental
731 United States Using Quantile Regression. *J. Hydrol. Eng.*, DOI:10.1061/(ASCE)HE.1943-5584.0001620,
732 2017.
733
- 734 Vinet, F.: Flood Risk Assessment and Management in France. The Case of Mediterranean Basins, Flood
735 Prevention and Remediation. WIT Press, Southampton, UK, p. 105–132, 2011.
736
- 737 Vinet F. and Meschinet de Richemond N.: Changes in Flood Risk: Retrospective and Prospective
738 Approach, Chap. 14 in F. VINET (ed.) *Floods 1: risk knowledge* ISTE edition London, p. 311-323, 2017.
739
- 740 Wasko, C. and Sharma, A.: Global assessment of flood and storm extremes with increased temperatures,
741 *Sci Rep*, 7(1), 7945, doi:10.1038/s41598-017-08481-1, 2017.
742
- 743 Wasko C. and Nathan R.: Influence of changes in rainfall and soil moisture on trends in flooding, *J.*
744 *Hydrol.*, 575, 432-441, 2019.
745
- 746 Westra, S., Alexander, L. V., and Zwiers, F. W.: Global increasing trends in annual maximum daily
747 precipitation, *J. Climate*, 26, 3904–3918, 2013
748
- 749 Whitfield, P.: Changing floods in future climates. *J. Flood Risk Manage.*, 5: 336-365, 2012.
750
- 751 Wilks, D.S.: The stippling shows statistically significant grid points: how research results are routinely
752 overstated and over interpreted, and what to do about it, *Bull. Am. Meteorol. Soc.*, 97, 2263–2273, 2016.
753
- 754 Woldemeskel, F. and Sharma, A.: Should flood regimes change in a warming climate? The role of
755 antecedent moisture conditions, *Geophys. Res. Lett.*, 43, 7556–7563, doi:10.1002/2016GL069448, 2016.
756
- 757 Yin, J., Gentine, P., Zhou, S., Sullivan, S.C., Wang, R., Zhang, Y., and Guo, S.: Large increase in global
758 storm runoff extremes driven by climate and anthropogenic changes. *Nat. Commun.* 9, 4389,
759 DOI:10.1038/s41467-018-06765-2, 2018.



760

761 Zhang, Y., Wei, H., and Nearing, M. A.: Effects of antecedent soil moisture on runoff modeling in small
762 semiarid watersheds of southeastern Arizona, *Hydrol. Earth Syst. Sci.*, 15, 3171-3179,
763 <https://doi.org/10.5194/hess-15-3171-2011>, 2011.

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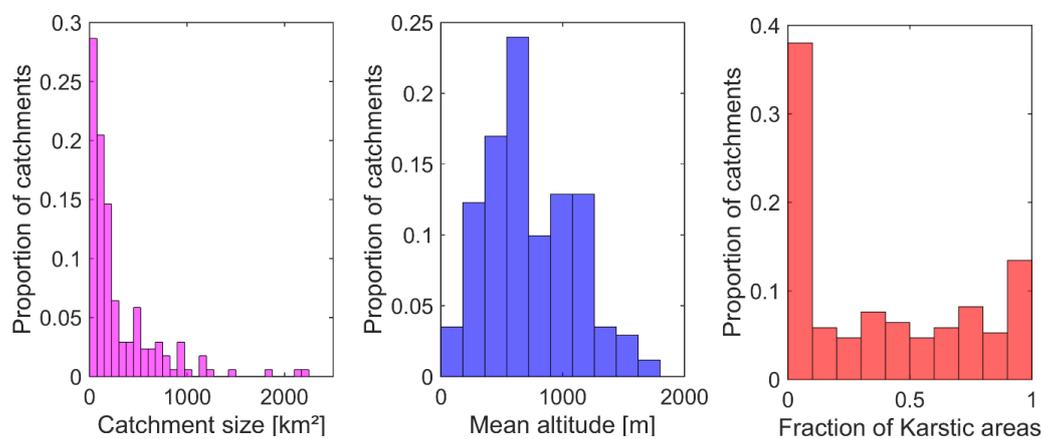
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795 **FIGURES**

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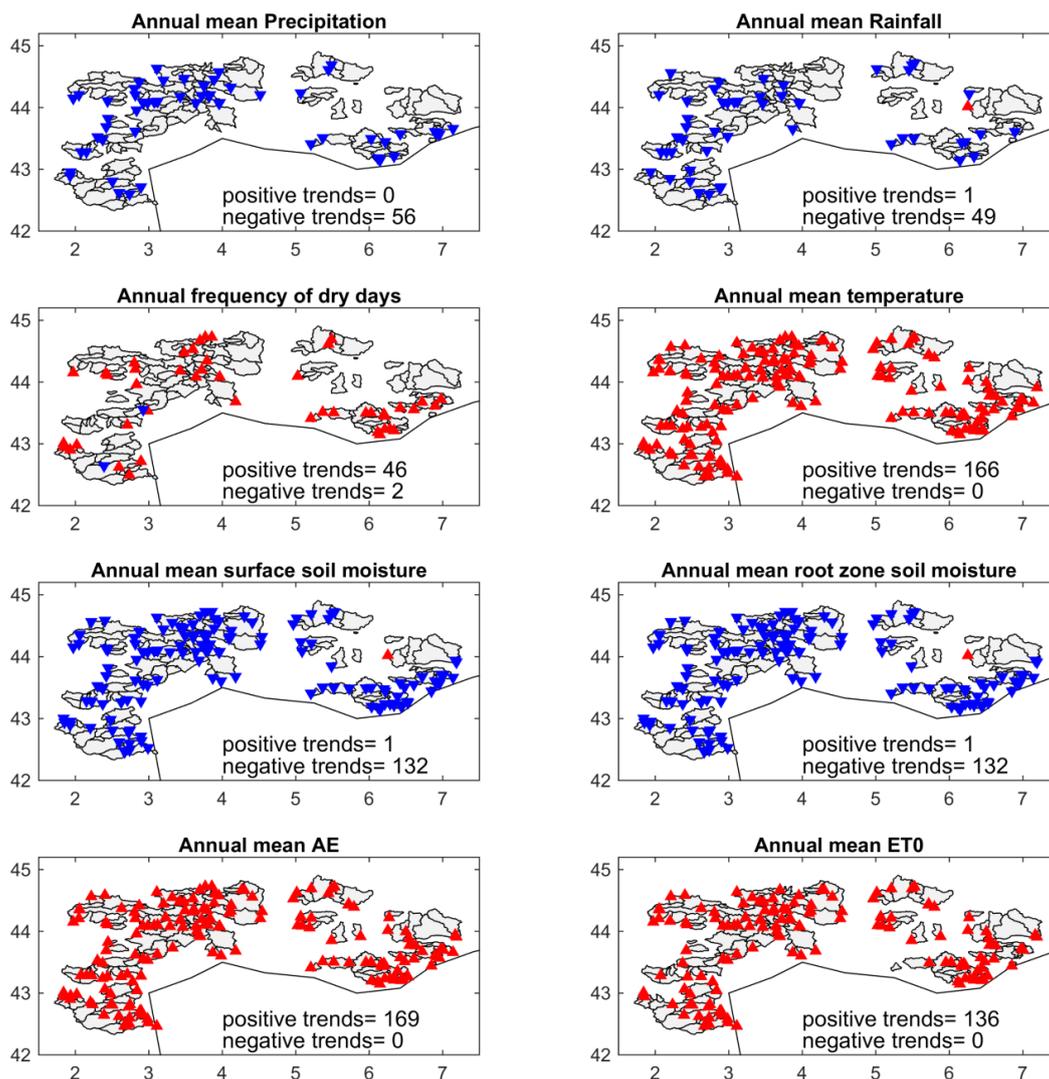
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Figure 1: Catchment size, mean altitude and fraction of karstic areas



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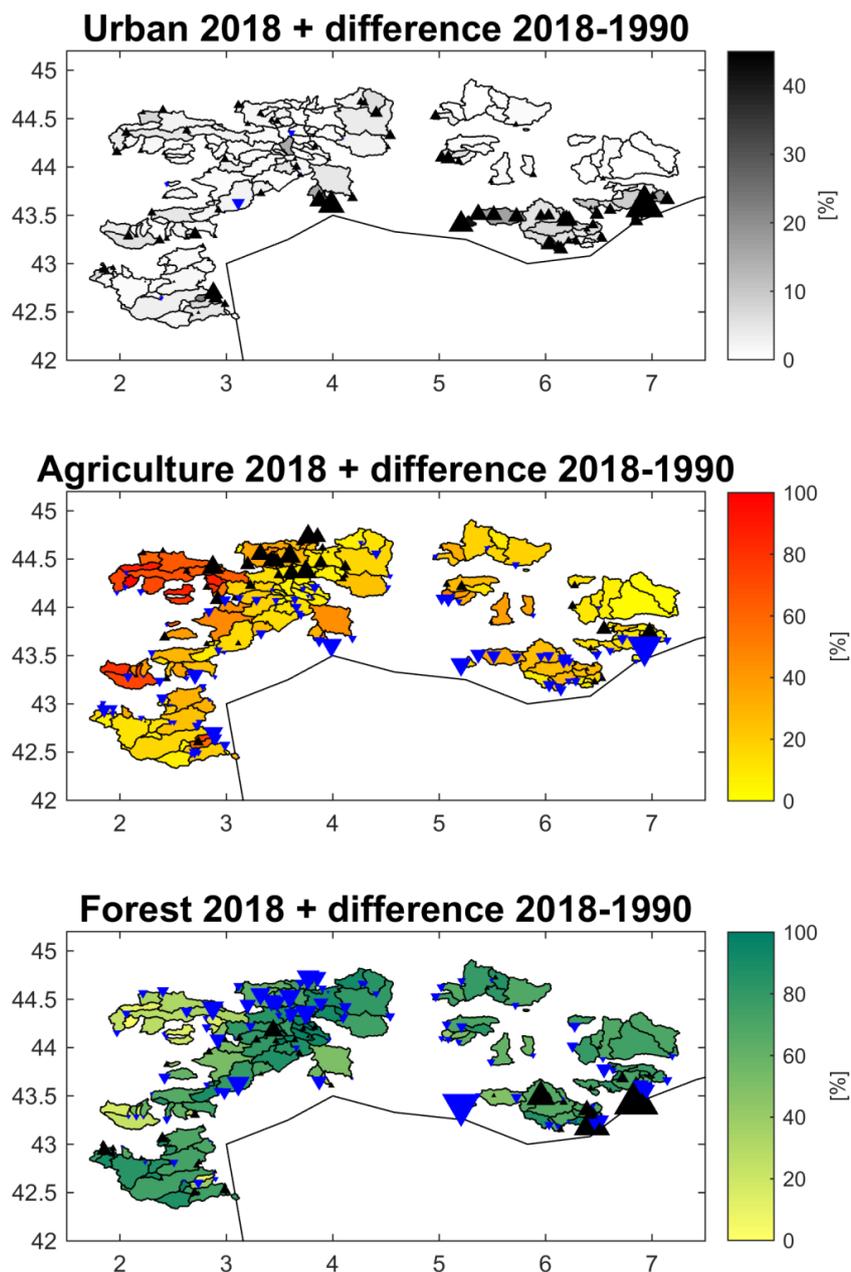
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Figure 2: Annual trends between 1958 and 2018 in precipitation, temperature, soil moisture,

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actual evapotranspiration (AE) and reference evapotranspiration (ET0).

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807 Figure 3: Urban, Agricultural and Forest cover by catchment from the CORINE database for the

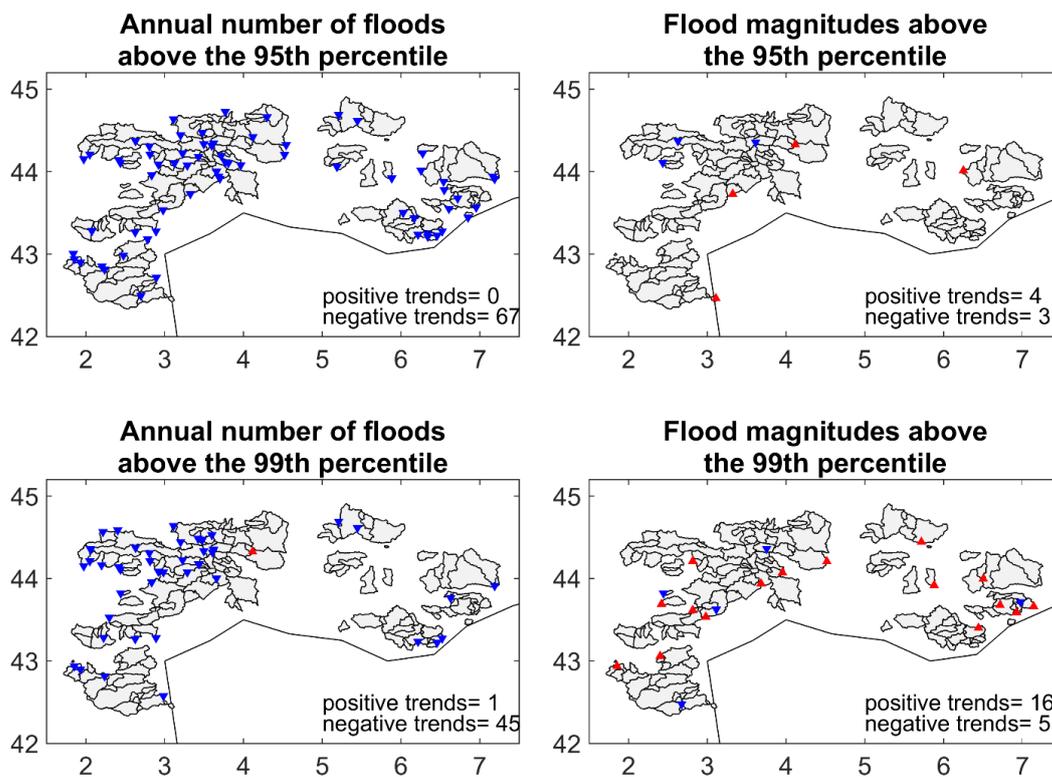
808 year 2018 and difference between 1990 and 2018 (upward black triangles indicate an increase,

809 downward blue triangles a decrease).



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813 Figure 4: Trends in the annual number of flood events above the 95th and 99th percentiles (left)

814 and in the magnitude of these threshold exceedances (right). Blue triangles indicate a decrease

815 and red triangles an increase.

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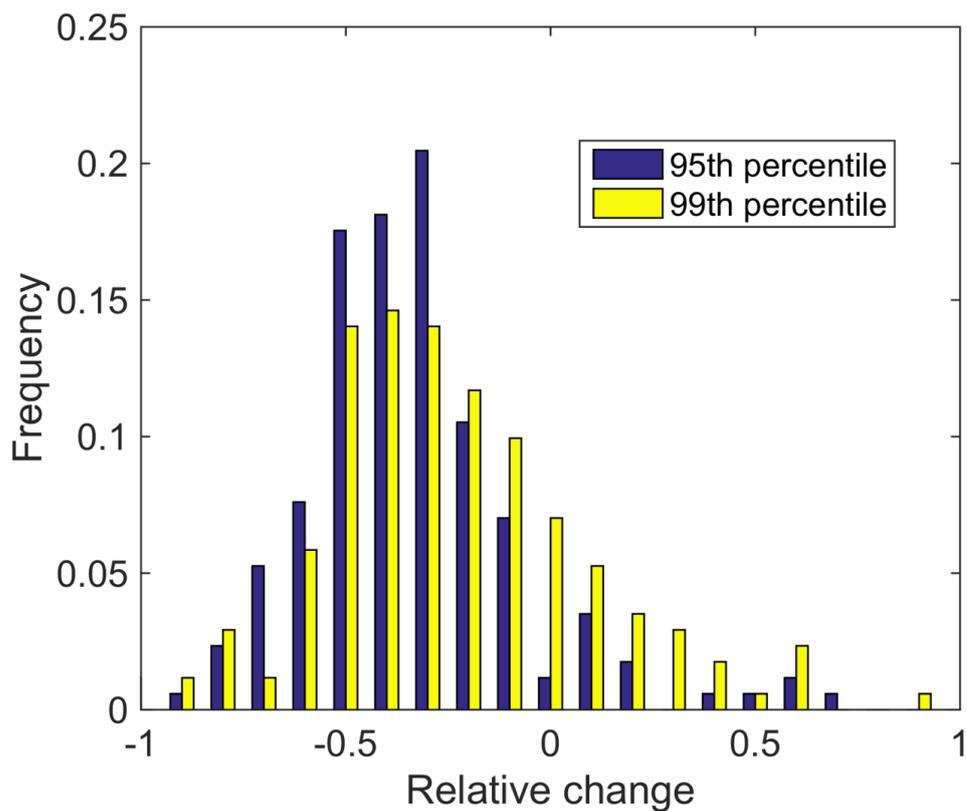
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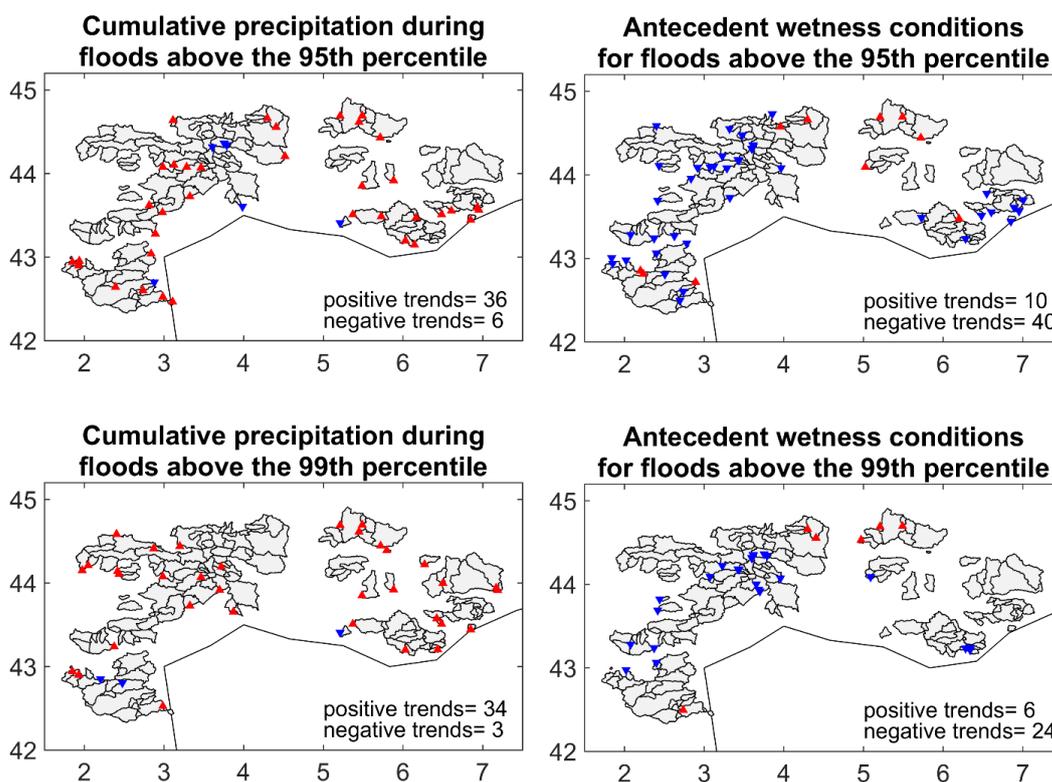
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Figure 5: Histogram of the relative changes in the 95th and 99th percentile estimated from the quantile regression models with time



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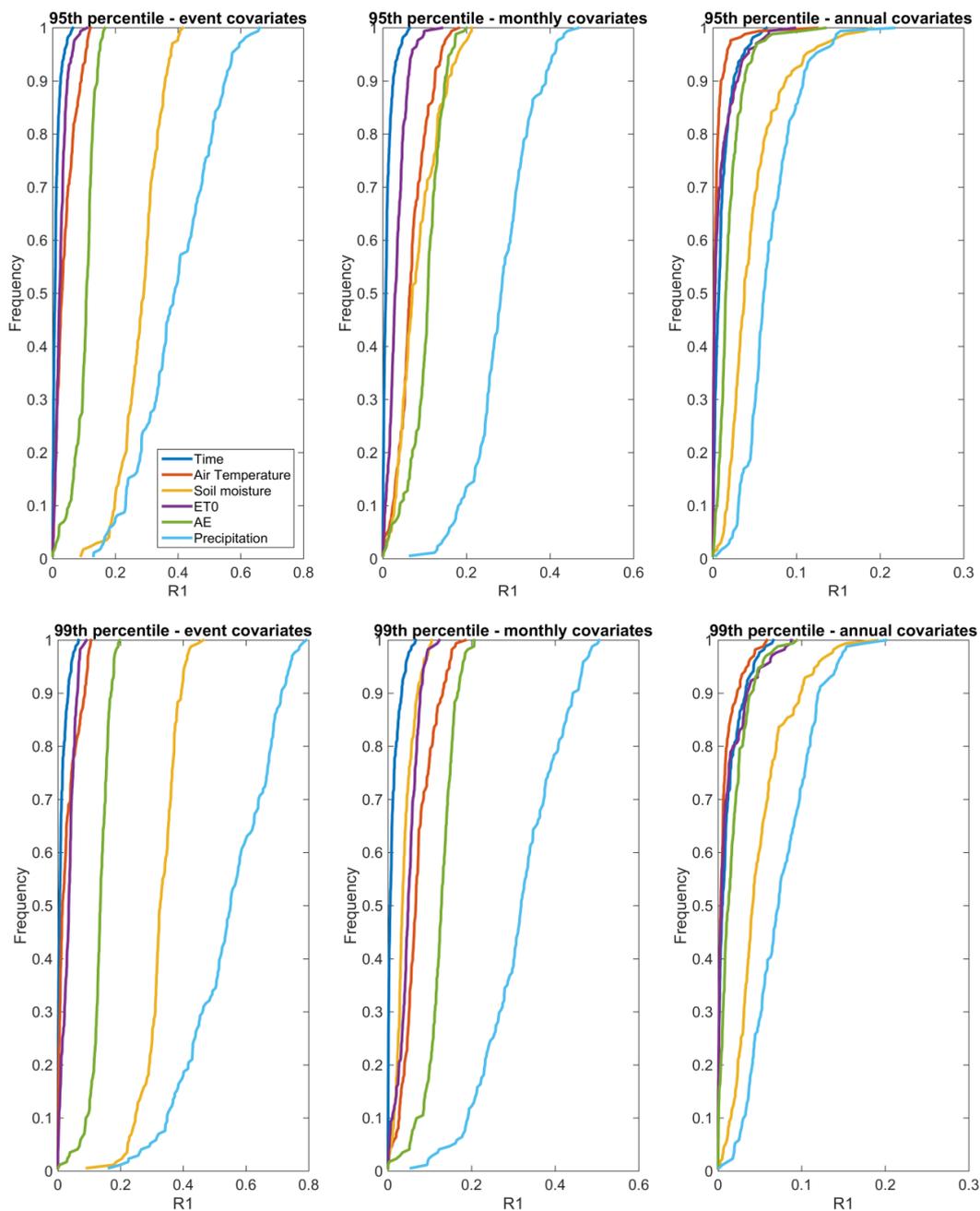
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Figure 6: Trends in cumulative precipitation during flood events above the 95th and 99th percentile (left) and in the soil moisture initial conditions (right). Blue triangles indicate a decrease and red triangles an increase.



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842 Figure 7: Distribution of the R^j coefficients for different covariates for the 95th or 99th percentiles
843 of daily runoff, averaged at: (i) the event scale (3 days), left panels, (ii) the monthly scale, central
844 panels, and (ii) annual timescale, right panels.

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