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The impact of the Three Gorges Dam on summer streamflow in the Yangtze River Basin

Running title: Impact of Three Gorges Dam on summer streamflow

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26

27 **The impact of the Three Gorges Dam on summer streamflow**
28 **in the Yangtze River Basin**

29 **Abstract**

30 The Three Gorges Dam is the world's largest capacity hydropower station located in the Hubei
31 province along the Yangtze River in China, which began operations in 2003. The dam also functions
32 to store and regulate the downstream releases of water in order to provide flood control and
33 navigational support in addition to hydropower generation. Flow regulation is particularly important
34 for alleviating the impacts of low and high flow events during the summer rainy season (June, July,
35 and August). The impact of dam operations on summer flows is the focus of this work. Naturalized
36 flows are modelled using a canonical correlation analysis and covariates of subbasin-scale
37 precipitation resulting in good model skill with an average correlation of 0.92. The model is then
38 used to estimate natural flows in the period after dam operation. A comparison between modelled
39 and gauged streamflow post 2003 is made and the impact of the dam on downstream flow is
40 assessed. Streamflow variability is found to be strongly related to rainfall variability. An analysis of
41 regional streamflow variability across the Yangtze River Basin showed a mode of spatially
42 negatively correlated variability between the upper and lower basin areas. The Three Gorges Dam
43 likely mitigated the occurrence of high flow events at Yichang station located near the dam.
44 However, the high flow at the remaining stations in the lower reach is not noticeably alleviated due
45 to the diminishing influence of the dam on distant downstream flows and the impact of the lakes

46 downstream of the dam that act to attenuate flows. Three types of flow regime changes between
47 naturalized and observed flows were defined and used to assess the changes in the occurrence of
48 high and low flow events resulting from dam operations.

49 **Keywords:** Three Gorges Dam; Canonical correlation analysis; High flow; Low flow

50 **1. Introduction**

51 The Yangtze River Basin in China is home to one-third of the country's population and the largest
52 capacity hydropower project in the world to date – the Three Gorges Dam (TGD) (Figure 1). The
53 river is also known as “Changjiang” in China - its literal translation being “long river”, which
54 reflects its status as the longest river in Asia and is a significant region in terms of China's history,
55 culture, and economy (Bermudez, 2015).

56 The basin's climate is characterized by complex regional subbasin-scale patterns of precipitation
57 and temperature in terrain ranging from plateaus to plains. The combination of complexity in both
58 climate and terrain results in widely varying rainfall-runoff processes across the basin that spans an
59 area of 1.8 million km². Periodic floods and droughts have been experienced in the basin throughout
60 history and a number of extreme events have captured the public's attention due to the loss of life
61 and impacts on the regional economy (Bing et al., 2012; Wang et al., 2011; Xu et al., 2008; Yin et
62 al., 2001). For example, in 1998 floods destroyed 4 970 000 houses (Zong et al., 2000) while a
63 drought in 1994 was estimated to have resulted in economic losses of CNY 20 billion
64 (approximately \$2.3 billion USD) (Zhang et al., 2015).

65 The TGD was originally conceived almost a century ago but progress was interrupted by political
66 and social upheavals. Construction of the TGD began in 1993 with flow regulation and water
67 impoundment commencing in 2003. Chinese laws and regulations stipulate that the primary
68 function of any water conservancy project must be flood storage and control. That is, flood control
69 operations will take precedence over other functions such as hydropower, transport and water
70 security. The dam is therefore operated to reduce the impacts of floods in spring and summer, while
71 water is released over winter and autumn to maintain downstream flows for water supply and to
72 facilitate navigation (Zhao et al., 2012). The Yangtze River has experienced a number of severe low
73 flow events (e.g. in 2006, 2008, 2011) resulting in suspension of navigation in reaches downstream
74 of the TGD, loss of residential water supply for hundreds of thousands in the Hubei Province (Qiu,
75 2011), and water shortages that impacted 18 million people during the 2006 autumn drought over
76 the Sichuan and Chongqing provinces (Zhang et al., 2015). Such events have raised concerns
77 regarding the effectiveness of the dam in providing seasonal flow regulation. Past studies have
78 focused on the TGD's impacts on downstream lakes, river levels and flows (Chen et al., 2016; Jiang
79 et al., 2014; Lai et al., 2014; Li et al., 2016; Mei et al., 2015; Wang et al., 2017; Wang et al., 2013),
80 however, the changes in flow regimes between naturalized flows and observed flows impacted by
81 the dam and the underlying mechanism have been largely overlooked. Consequently, an analysis
82 that compares estimated naturalized flows against flows regulated by the TGD in the period post
83 impoundment is important for determining whether dam operations likely reduced or exacerbated

84 low flow or high flow occurrences, and how the occurrence of these flow types have changed.

85 Numerous approaches for estimating naturalized flows have been used including distributed

86 hydrological models and stochastic models. Hydrological models such as the physically based Soil

87 and Water Assessment Tool (SWAT) model, MIKE SHE, and Variable Infiltration Capacity (VIC)

88 model have been widely used in the analyses of streamflow modelling and their predictions (Kim et

89 al., 2016; Liang et al., 1994; Vansteenkiste et al., 2013; Yuan et al., 2004). However, distributed

90 hydrological models are limited by large data requirements, unclear parameter uncertainty, and

91 inaccurate modelling of physical process (Gayathri et al., 2015; Pechlivanidis et al., 2011). Data

92 constraints, such as limited data on catchment characteristics and land use limit the ability to

93 accurately model hydrological processes using physically based models.

94 Alternatively, statistically-based models, such as regression and correlation models, are

95 observation-oriented methods that seek to find functional relationships between explanatory and

96 response variables (Elsanabary et al., 2015; Kwon et al., 2009; Steinschneider et al., 2016). These

97 models are data orientated and do not explicitly consider the process driving the hydrological

98 system. In addition, stochastic models are typically comprised of a relatively small number of

99 model parameters. These models therefore allow for a parsimonious characterization of streamflow

100 variability. Yangtze River Basin streamflow has been shown to be strongly related to rainfall

101 particularly at coarser (seasonal to annual) timescales. For example, the correlation coefficient

102 between annual precipitation and streamflow at Datong station in the Lower Yangtze River Basin is

103 0.85 (Zhang et al., 2011) suggesting that statistical rainfall-runoff models can be useful in the
104 region.

105 Here, we assess the impacts of the TGD on summer flows using a statistical approach to model
106 naturalized flows and comparing these modelled flows with observed flows in the period following
107 dam operation. The rainfall and streamflow data used in the analysis are described in Section 2
108 along with key hydrological features along the Yangtze River. The canonical correlation analysis
109 used to model streamflow across the basin is described in Section 3 and we address the issue of high
110 spatial correlation among both rainfall covariates and streamflow. In Section 4 we quantify the
111 relationship between summer (the wettest season) rainfall and streamflow along the main stem of
112 the Yangtze River downstream of the TGD. The spatial and temporal patterns of rainfall anomalies
113 over the whole basin are used to qualitatively assess whether changes in rainfall variability can
114 likely explain changes in streamflow in the post dam period compared with modelled streamflow.
115 We also assess the influence of the dam on summer flow by identifying changes in low, normal and
116 high flow regimes. We find that the TGD likely attenuates both high and low flows at downstream
117 locations close to the dam. However, the dam's impacts are largely eliminated on flows further
118 downstream.

119 **2. Study area and data description**

120 The TGD is located approximately 1880 km upstream of the estuary where the Yangtze River flows
121 into the East China Sea, and significant landmarks of the Dongting Lake (covering an area of ~

122 2625 km²) and Poyang Lake (~4125 km²) are located in this reach. To assess the likely impacts of
123 dam operations on regulating summer high and low flows along the Yangtze River, we use
124 streamflow data from four hydrological stations downstream of the dam located on the main stem at
125 Yichang, Luoshan, Hankou, and Datong (Figure 1).

126 The Yichang station is known as the “Gateway to the Three Gorges” and is located approximately
127 40 km downstream of the dam. The Luoshan station is located close to the outlet of the Dongting
128 Lake that aggregates the flows from two major subbasins (i.e. Yuanshui and Xiangjiang). As a result,
129 streamflow at Luoshan station is impacted by natural water detention in the lake. Hankou station
130 was selected due to its importance for flood monitoring given its location downstream of where the
131 Han River, which is the largest tributary, joins the main stem of the Yangtze River. Similar to the
132 Luoshan station, streamflows at Datong station are impacted by lake effects due to its location
133 downstream of Poyang Lake. The station is also critical for monitoring environmental flows and
134 outflows to the sea.

135 Daily streamflow records were obtained from the Changjiang Water Resources Commission. Data
136 have been checked and quality controlled and cover the period from 1960 to 2014 for each of the
137 four stations. The monthly averaged streamflow data was obtained by averaging the daily flow rates
138 within each month. The annual hydrograph shows the seasonality of streamflow (Figure 2), which is
139 monsoon summer (June-July-August) flow dominated. As a result, the summer season (JJA) was
140 selected to enable an analysis of changes in the occurrence of high and low flows. The summer

141 streamflow was logarithmically transformed as the streamflow at all four stations were
142 approximately log-normally distributed.

143 Daily rainfall data was accessed from the China Meteorological Data Service Center (2016).
144 Rainfall records that span from 1960 to 2014 with less than 5% missing values were selected
145 resulting in data from 136 weather stations. For each station, missing values were infilled using the
146 average values from the two nearest stations with rainfall recorded on that day. The intra-annual
147 pattern of rainfall is similar to that of streamflow with maximum monthly precipitation occurring
148 during summer. The gauged rainfall data was spatially aggregated using the 21 subbasin boundaries
149 (Figure 1) obtained from the Changjiang Water Resources Commission. The rainfall over a given
150 subbasin is the spatially averaged value across all rainfall stations within the subbasin using the
151 Thiessen Polygon Method (Zhong et al., 2018). The spatial distribution of rainfall stations is highly
152 heterogeneous. The Thiessen polygon method was applied as this approach accounts for the uneven
153 distribution of rainfall sites, allowing for a more reasonable representation of aggregated rainfall.
154 This method assigns an area called the Thiessen polygon (or the Voronoi diagram) to each rainfall
155 station thereby providing an area-weighted average, which was applied to all stations within a
156 subbasin to attain the aggregate subbasin rainfall.

157 **3. Methodology**

158 Here we present an approach for modelling unregulated summer streamflow using rainfall as a
159 covariate. Canonical correlation analysis was selected to model the relationship between rainfall

160 and streamflow to account for rainfall induced changes of the streamflow across the four
161 hydrological stations. The model performance metrics and method of analysis of the dam impact on
162 summer streamflow were then described. The conveyance period for streamflow to travel from the
163 upper catchment to the estuary is approximately 14 days (Chu et al., 2006), which is less than our 3
164 month time scale. Given the relatively coarse temporal resolution considered, we did not consider a
165 temporal lag in our analysis.

166 **3.1 Model design**

167 The date of June 2003, hereon used to define the pre- and post-dam periods, delineates the change
168 from a free-flowing river to an impounded river as this was the time at which storage in the
169 reservoir behind the TGD began to accumulate. Initial diagnostic analyses were conducted using the
170 rank correlation analysis between the rainfall $x_{t,k}$ in year t at subbasin k , and the log transformed
171 streamflow $\ln(y_{t,s})$ in year t at station s in the pre-dam period (shown in Figure 3). The summer
172 flow data for all stations have the strongest positive correlations with spatially averaged rainfall
173 over adjacent subbasins. However, rainfall over some subbasins far away from the streamflow
174 stations has a negative correlation with the flow (e.g. the correlation of flow at Luoshan and rainfall
175 over Mintuojiang in upper basin).

176 Physically, rainfall can only contribute to downstream streamflow, however, rainfall in a
177 downstream subbasin may be indicative of regional climate that influences streamflow across the
178 basin (e.g. the precipitation over a downstream catchment may be negatively correlated with

179 upstream streamflow). As a result, the model was designed using the spatially averaged rainfall over
180 each subbasin.

181 The spatial and temporal relationship between rainfall and streamflow can be expressed by a
182 formula $y = f(x)$, where y is streamflow and x is rainfall. The model was fit using data prior to
183 2003 in order to quantify the rainfall-runoff relationship prior to regulations by the TGD. The model
184 was then used to estimate the streamflows post 2003 as informed by rainfall. The estimated flows
185 are referred to as modelled naturalized flows here and are compared with the gauged flows post
186 2003 to assess the likely impacts of streamflow regulation by the TGD. A canonical correlation
187 analysis (CCA) was applied to maximize the correlation between the covariates and the target
188 variables.

189 **3.2 Canonical correlation analysis**

190 The proximity of the 21 subbasins results in covariates that are significantly correlated. Preliminary
191 analysis (not presented here) showed that the correlations between many subbasins were greater
192 than 0.9. Regression models require the explanatory variables to be significantly independent in
193 order to avoid multicollinearity. Thus, canonical correlation analysis (CCA) (Hotelling, 1936) was
194 implemented to allow for dimension reduction among the highly correlated rainfall data. CCA
195 differs from other methods of dimension reduction such as principle component analysis (PCA)
196 (Jolliffe, 2002) and archetype analysis (AA) (Cutler et al., 1994; Steinschneider et al., 2015; Stone
197 et al., 1996) as CCA is constructed to maximize the correlation between the explanatory and target

198 variable through a linear transform. It has been used in applications of reconstructing streamflow
199 using paleoclimate records (Ho et al., 2016), refining climate projections (Chen et al., 2017) and
200 assessing regional flood frequency (Ouarda et al., 2001).

201 A brief description of CCA is presented here and we refer readers to Hotelling (1936) for more
202 details. CCA consists of the linear transform of two variables to canonical form in order to
203 maximize the correlation between them. Given two random variables X and Y , where X and Y
204 consist of p and q variables respectively, there exist two vectors a (of length p) and b (of length q)
205 such that $U = a' X$ and $V = b' Y$ where the correlation between U and V is maximized. U and V
206 yield the first pair of canonical variables. Subsequent canonical pairs may be derived but with the
207 additional constraint that they are orthogonal to previously identified canonical variables.
208 Dimension reduction is achieved by selecting fewer than $\min(p, q)$ pairs of canonical variables. In
209 this study, the explanatory variable X is comprised of rainfall $x_{t,k}$ for $k \in 1:21$ subbasins and the
210 target variable Y is the log-transformed streamflow $\ln(y_{t,s})$ for $s \in 1:4$ and $t \in 1960:2014$ for
211 both X and Y . In our study, the algorithm of CCA is implemented using the CCA package in R
212 language by Gonzalez et al. (2008).

213 The first 4 pairs of canonical variables were retained to develop models relating rainfall and
214 streamflow. For each pair, a linear regression equation was built using Equation (1).

$$215 \quad V_i = \beta_i U_i + \alpha_i + \varepsilon_i \quad (1)$$

216 where, i is a value ranging from 1 to 4, denoting the order of the 4 pairs of canonical variables. β_i

217 is the regression coefficient and α_i is the intercept in i -th linear equation. ε_i is the error term and
 218 involves various uncertainties stemming from variability in streamflow unaccounted for by rainfall.
 219 Rainfall in the post dam period is used as covariates to estimate the naturalized streamflow using
 220 the canonical covariates a and Equation (1). Log-transformed streamflow is obtained by calculating
 221 V and back transforming V using canonical weights b . Streamflows were then back transformed
 222 from the log space.

223 **3.3 Model performance metrics**

224 A leave- m -out cross validation method was applied to verify the skill of the model. In this process,
 225 m randomly selected observations of streamflow data are retained for validation, while the
 226 remaining $n-m$ data are used for model calibration, where n is the total number of observations. In
 227 this study, we first used $m = 5$ (approximately 10% of the total number of observations) and
 228 randomly selected m data points without replacement from the time series. The remaining 38 years
 229 of data were used for model calibration. The procedure was repeated 100 times to produce a set of
 230 validation metrics computed from all the cross validated models. Two metrics, the reduction of error
 231 (RE) and the coefficient of efficiency (CE) (Cook et al., 1994; Wilson et al., 2010) for those original
 232 variables rather than the canonical variables, were chosen to verify the performance and
 233 predictability of the model. These are defined as follows in Eq. (2) and Eq. (3):

$$234 \quad \text{RE} = 1 - \frac{\sum(x_i - \hat{x}_i)^2}{\sum(x_i - \bar{x}_c)^2} \quad (2)$$

$$235 \quad \text{CE} = 1 - \frac{\sum(x_i - \hat{x}_i)^2}{\sum(x_i - \bar{x}_v)^2} \quad (3)$$

236 Where, x_i and \hat{x}_i are the observed and predicted streamflow in year i of the validation period,
237 respectively. \bar{x}_c is the mean of observations used for calibration and \bar{x}_v is the mean of
238 observations used for validation. Theoretically, the values of both RE and CE range from the
239 negative infinity to a maximum of 1, with higher values indicating better prediction skill. A positive
240 value indicates that the model has better predictive skill than using the climatology of the
241 verification period. The RE and CE statistics are similar to the Nash-Sutcliffe efficiency score
242 (McCuen et al., 2006; Nash et al., 1970) and have previously been used to assess the performance
243 of streamflow models (Chen et al., 2014; Devineni et al., 2013; Ho et al., 2016).

244 **3.4 Analysis of the impact of the TGD on summer streamflow**

245 Based on the modelled streamflow across the four hydrological stations, 5-year moving averages of
246 the modelled naturalized streamflow at each gauge were calculated. The observed and modelled
247 streamflows are first qualitatively assessed. To analyze the variability of the non-regulated flow in
248 both space and time principal component analysis (PCA) was applied.

249 Changes in streamflow during the post-dam period are induced by both rainfall variations and dam
250 regulation. We therefore analyze and interpret variations in streamflow in the post dam period by
251 considering the probable impacts of both variable rainfall and dam regulation. A qualitatively
252 assessment of the influence of rainfall variability on summer flow was first investigated to show
253 whether the changes in flow were possibly due to changes in rainfall. Then changes in flows at the
254 four streamflow locations due to dam regulations were quantitatively assessed using estimates of

255 naturalized flows informed by rainfalls.

256 To quantitatively assess the differences between the modelled and observed streamflow during the
257 post-dam period the frequency of high flow and low flow events and quantiles of summer flow were
258 calculated for modelled and observed streamflow pre and post 2003. The comparison between
259 streamflow before and after dam operation provides an approach to assessing the impact of the dam
260 on either exacerbating or mitigating extreme events. We define a high flow occurrence as a
261 streamflow event greater than the 75th percentile summer flow of the pre-dam period. Similarly, a
262 low flow occurrence is streamflow less than the 25th percentile summer flow of the pre-dam period.

263 **4. Results**

264 **4.1 Model results**

265 The summer streamflows for all gauges along the main stem were modelled using the spatially
266 averaged rainfall data and resulted in mean adjusted R^2 value of 0.80. The mean modelled
267 streamflows along with the 5th-95th prediction interval are shown in Figure 4. The modelled
268 streamflows closely match the observations, and the average correlation across all stations was 0.92.
269 Some rainfall-runoff models have shown improved results when spatially averaged rainfall is used,
270 compared with using gauged data (Tramblay et al., 2011).

271 To verify the performance and predictive skill of the model, cross validation was performed. The
272 RE and CE values were calculated, and the results are presented in boxplots (Figure 5). There

273 resulted in a model with median RE and CE values of around 0.8 for all streamflow sites, indicating
274 that the associated model predictions are more informative than climatology of the verification
275 period.

276 **4.2 Main modes of natural streamflow variability**

277 Summer averaged streamflows at the four selected gauges within the Yangtze River Basin modelled
278 using subbasin-scale rainfall were used to further analyze changes in streamflow response pre and
279 post dam regulation. Both modelled and observed summer streamflow for the period 1960-2014 are
280 shown in log space in Figure 6.

281 Prior to dam regulation the temporal variability of modelled streamflows closely matches that of the
282 observations for all stations. The low flows in the early 1970s and high flows in the late 1990s
283 (Gemmer et al., 2008) are visible in both the observed and modelled flows at all stations. The 5-year
284 moving averages show the flows for the four stations fluctuate periodically. In the early 1970s all
285 stations experienced negative summer flow anomalies. Negative streamflow anomalies are also
286 observed and modelled in three of the stations (i.e. Luoshan, Hankou and Datong) in the late 1980s.
287 In contrast, the early 1980s showed a persistent flood regime with positive anomalies seen in the
288 5-year moving averages for all stations.

289 PCA was used to extract the main spatiotemporal modes of logarithmically transformed modelled
290 streamflows across all four stations to investigate their relationships both in space and time. A scree
291 test was used to inform the number of PCs to retain (scree plot was shown in Figure S1 of the

292 supporting information). The first two PCs were retained and together explain over 97% of the
293 streamflow variance (PCs were shown in Figure 7).

294 For PC1, the loading values are of the same sign across all stations representing a regionally
295 cohesive signal. Both the flood in 1998 and drought in 1972 (Heng et al., 1999; Jiang et al., 2017;
296 Yang et al., 1982; Zhang et al., 2015) are evident in the PC scores. During the 1998 floods, water
297 transport from three regions, the Indian Ocean, the Bay of Bengal-South China Sea, and the Pacific
298 Ocean, resulted in persistent heavy rainfall over the southwestern, southern and southeastern
299 subbasins, respectively, culminating in a widespread severe flood (Jiang et al., 2017). Conversely, a
300 large negative score for PC1 can be seen in 1978, when the middle and downstream Yangtze River
301 Basin were struck by a severe drought event resulting from reduced summer precipitation (Zhang et
302 al., 2015).

303 Unlike PC1, PC2 represents a non-homogeneous pattern of streamflow variability in the Yangtze
304 River Basin (see loading values for PC2 in Figure 7). The loading values are of opposite sign for
305 Yichang station close to the dam versus those located further downstream. This mode of variability
306 distinguishes between streamflow in the mid to upper basin and the lower basin and is
307 representative of the differences in regional climate and streamflow response. For example, Figure 6
308 shows anomalously low flows at Yichang station in 1969, however the remaining three downstream
309 stations experience average to above average flows. A similar signal is seen in 1996 where flow at
310 Yichang is average or slightly above average, while streamflows at the remaining three stations are

311 anomalously high. This signal is reversed in 1981 when flow at Yichang is above average and flows
312 at the downstream stations show negative anomalies.

313 **4.3 Influence of rainfall variability on summer streamflow**

314 High correlations between streamflow and precipitation suggest that streamflow variability is
315 strongly influenced by precipitation variability. The spatial and temporal patterns of rainfall
316 anomalies over the whole basin (Figure S2 of supporting information) are therefore first used to
317 qualitatively assess whether changes in rainfall variability can likely explain changes in streamflow
318 in the post dam period.

319 In the years of 2003 and 2010, anomalously high rainfalls were experienced in the subbasins located
320 the northern section of upstream of Yichang station (e.g., Jialingjiang basin), which should have
321 resulted in a high flow event. However, the dam operation likely reduced the magnitude of the flow
322 as evident by the reduced streamflow observations compared to the modelled flows at Yichang
323 station, which was likely due to decreased water releases thereby retaining the floodwaters and
324 reducing the downstream flows. Streamflow at the remaining three gauges further downstream were
325 not notably impacted by dam operations.

326 In 2008, streamflow observations at all four gauges recorded below average flows despite rainfall in
327 the northern section of the lower basin being slightly above average. It is likely that decreased
328 outflows from the dam in combination with decreased rainfall in the lower basin region resulted in
329 low flow conditions at Hankou and Datong stations downstream of the Dongting Lake.

330 In 2012, average rainfall was observed over the central region of the Yangtze River Basin with the
331 exception of the Qingjiang subbasin, a relatively small subbasin located in the center of the basin
332 (Figure 1). No notable change was observed at Yichang station when comparing modelled and
333 observed streamflows. However, water supply from the lakes downstream (e.g. Dongting Lake and
334 Poyang Lake), where above average rainfalls were observed, may have contributed to increased
335 flows at Luoshan, Hankou and Datong stations compared with naturalized flows.

336 **4.4 Dam impacts on summer streamflow**

337 The modelled streamflows represent naturalized flows unimpacted by dam regulation and were
338 compared with the observed flows for all stations in the post dam period to assess potential changes
339 in the occurrence of high and low flows. The difference between the modelled and the gauged
340 values therefore represents the difference in streamflow as a result of dam operations in addition to
341 model errors. Figure 8 shows the distribution of high flow and low flow events for the
342 non-regulated flows from 1960 to 2014 with results for the period of 2003-2014 obtained using
343 modelled flows. Figure 8 shows that low flows and high flows may be wide spread across the basin
344 (e.g. basin-wide high flows occurred in 1962, and basin-wide low flows occurred in 1971). These
345 high and low flow also can co-occur in different regions of the basin (e.g. 1981 with low flow
346 conditions in the lower basin and high flow conditions in the middle basin). Streamflows across the
347 basin are neutral (i.e. in middle two quartiles) in approximately 16% of years.

348 The frequency of high flow and low flow events in the pre and post dam periods is shown in Table 1

349 for both modelled naturalized and observed flows. The frequency of high flow events for Yichang
350 increased by 0.12 when comparing naturalized flows. The remaining stations show a decreased
351 frequency in high flow events with a decrease in high flow frequency of 0.09 for Luoshan and
352 Hankou and 0.03 for Datong. The frequency of low flow events for all stations increased by
353 0.04~0.12. Compared with the modelled naturalized flows, the frequency of high flow events at
354 Yichang and Datong are reduced in the post-dam period, which indicates that the operation of the
355 dam has likely contributed to the greatly reduced occurrence of high flow events. While the TGD
356 appears to have attenuated the magnitude of high flow events, the occurrence of low flow events
357 appear to have been magnified as seen when comparing modelled and observed low flow
358 occurrences in the period post 2003.

359 The observed and modelled flows in the post dam period (2003-2014) are plotted as quantiles of the
360 flows in the pre-dam period in Figure 9. The threshold lines of high flow and low flow events were
361 plotted in Figure 9 shown by dashed and solid line respectively. These threshold values are used to
362 identify three types of changes in occurrence of high flows or low flows at the four stations
363 resulting from dam operation. The first type of flow regime change is defined as a year in which a
364 high flow or low flow is identified in the naturalized flows, but has been alleviated by dam
365 operations as evident in the observed flows. The second case is when no high flow or low flow is
366 evident in the naturalized flows, but occurs in the observed record. The third case is when a high
367 flow or low flow is observed in the naturalized flows, but the observed flows show the opposite

368 flow event of a low or high flow event respectively. The years in which these three different types
369 of flow regime changes occur are tabulated in Table 2.

370 From Table 2 and Figure 9, it can be seen that in the 12 years between 2003 and 2014 that the dam
371 likely mitigated high flow events at Yichang in three of these years (i.e. 2003, 2010, 2014) and at
372 Luoshan in 2014 exemplifying the first type of flow regime change. In contrast, these high flow
373 events were not alleviated by dam operations for the lower reach. These middle and lower basin
374 regions are influenced by lake effects upstream of the stations (Wang et al., 2017) and consequently
375 dam operations have limited noticeable impacts further downstream. An example of the second case
376 of flow regime change occurred in 2012, where either a high or low flow event is observed, but
377 does not appear in the modelled naturalized flows. Here, the naturalized flows are within the
378 25th-75th percentile range at Luoshan, Hankou and Datong. However, the observed high flows
379 indicate that the combination of dam operations and lake effects resulted in increased flows at these
380 locations. Similarly, in 2013, a low flow event for Luoshan and Hankou was observed, while was
381 not evident in the modelled naturalized flows. A similar occurrence is seen in 2008 at Yichang and
382 Luoshan. In the same year of 2008, the third type of flow regime change occurred in the two
383 stations further downstream of the dam. That is, the naturalized flow for Hankou and Datong show
384 a high flow event while the observation indicates the occurrence of a low flow event. The flow for
385 Datong in 2013 shows a similar change in events.

386 There are a number of potential reasons for differences in the dam's impact on high flow and low

387 flow occurrence in the middle and lower portions of the basin. These reasons include: 1) The
388 topography of the lower basin is dominated by relatively flat plains containing large areas of lakes
389 (a total lake area of 15 000 km²). These lakes naturally attenuate the tributary flows, which impact
390 flows to the main stem. These lake effects on streamflow occur in addition to the impacts imposed
391 by the TGD (Wang et al., 2017). 2) The Yangtze River Basin consists of 21 subbasins (see Figure 1),
392 whose runoff form the main components of inflows at the four downstream hydrological stations.
393 However, spatial heterogeneity in rainfall and catchment responses mean that high flow and low
394 flow events do not necessarily co-occur across the different sub-basins. Perhaps more importantly,
395 the geographical distribution of the precipitation in the Yangtze basin generally increases from
396 northwest to southeast. The dam's locations in the drier part of the upper catchment limits the ability
397 to attenuate floods resulting from storm events in the wetter eastern catchments downstream of the
398 dam.

399 **5. Discussion and conclusions**

400 A novel analysis of the impact of the Three Gorges Dam on the summer streamflow in the Yangtze
401 River Basin was presented using canonical correlation analysis to quantify the relationship between
402 the summer streamflow and rainfall across the Yangtze River Basin. The development of the CCA
403 model was motivated by the need to investigate the potential impact of the TGD on summer
404 streamflows by modelling naturalized streamflow and comparing these modelled flows with
405 observed streamflows that are potentially influenced by dam operations.

406 Another model using rainfalls over each station as covariates was trialed, but the results were not
407 presented in this paper and was hindered by the collinearity and the high dimension of the data.
408 While using a weighted average of the rainfall results in a reduction in variance, this reduction is
409 akin to the natural spatial aggregation represented in the streamflow. The smaller number of
410 sub-basins results in the improved ability to resolve relationships between rainfall and streamflow
411 using CCA compared with using a large number of rainfall stations.

412 The complexity of the model was constrained by the dearth of readily available data that would
413 have improved the streamflow model such as data on catchment characteristics. The approach of
414 CCA addressed the issues of high dimensionality and large spatial correlations of rainfall records
415 over the whole basin.

416 Our study was focused on changes in occurrence of high flow and low flow events as informed by
417 streamflow before and after the commencement of operations of the TGD in 2003 and evaluating
418 the impact of dam operations on mitigating the magnitude of these events. The modelled naturalized
419 streamflows generally closely match the observed streamflows for the four stations. Our analysis
420 shows that the streamflows at Yichang station, located at a relative close proximity to the dam, are
421 more noticeably impacted by the TGD as evident by the reduced occurrence of high flow events
422 after 2003. Flows at the other three stations appear to be less impacted by the dam operations due to
423 the increased distance from the dam, the impacts of the lakes located downstream of the dam (Gao
424 et al., 2013; Guo et al., 2012; Lai et al., 2014) and the inability to attenuate flows generated by

425 runoff in the higher precipitation regions of the lower basin.

426 Three types of flow regime change were identified and summarized for the post dam period. The
427 dam did not consistently mitigate the occurrence of either high or low flow events at the four
428 locations considered. These changes in flow regimes, sometimes resulting in high or low flows,
429 suggest that the dam operations are unable, in isolation, to alleviate high flow or low flow events in
430 the lower basin as other factors likely influence these events at distant locations. Although many
431 previous studies demonstrate that the flow regulation in impounded rivers are solely a result of
432 dams (Nilsson et al., 2005), Dai et al. (2012)'s analysis support our result, where they concluded
433 that the flows from the upstream basin of the TGD can only account for about 50% of changes in
434 multiyear mean flow over the period 1950-2005 at Datong station, and the discharge from the
435 downstream basin accounted for the remaining difference, where flows were influenced by other
436 factors including the effects of Dongting Lake and Poyang Lake and potentially other changes in
437 water use and management. Further, impacts of the TGD on flow so far downstream would largely
438 be masked by significant inflow to the river from the wetter catchments in the eastern Yangtze River
439 basin.

440 Several limitations could be addressed in future analyses. First, the model was developed using only
441 summer rainfall excluding other factors, such as the land use change and changes in large scale
442 atmospheric circulation (Xu et al., 2007; Xu et al., 2017). Theoretically, our analysis may result in a
443 conservative estimate of the streamflow but it provides an original approach to assess the likely

444 influence of the TGD on streamflows. Second, the approach of CCA demonstrated here considered
445 both the correlations of the streamflow stations and spatial correlations of rainfall stations over the
446 entire Yangtze Basin. An improved understanding of the complex physical processes that influence
447 precipitation occurrence and runoff characteristics would aid in the understanding of changes in
448 runoff and streamflow in the Yangtze River catchment. Such future endeavors are needed to build
449 on past studies such as Chen et al. (2016) and Lai et al. (2014) that have contributed to this
450 understanding and to further our understanding of the influence of the TGD on streamflow. Third,
451 only seasonal high flow regimes were considered here rather than distinct flood events that likely
452 pose the greatest risk to downstream regions.

453 **Data Availability Statement**

454 Daily streamflow records were obtained from the Changjiang Water Resources Commission. While
455 this data are not publicly available, they may be requested from the Changjiang Water Resources
456 Commission. Daily rainfall data was accessed from the China Meteorological Data Service Center
457 (<http://data.cma.cn/en>).

458 **Supporting information**

459 Figure S1. Scree plot showing the variance of each PC for the modelled naturalized streamflows.
460 Figure S2. The spatial and temporal distributions of rainfall anomalies over subbasins of the
461 Yangtze River Basin.

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610

611 **Table 1** The frequency of high flow and low flow events pre and post 2003. The modelled
 612 frequency for all stations from 2003 to 2014 is only influenced by rainfall, and the observed
 613 frequency is influenced by rainfall and TGD.

Stations	High flow events			Low flow events		
	Observed (1960-2002)	Modelled (2003-2014)	Observed (2003-2014)	Observed (1960-2002)	Modelled (2003-2014)	Observed (2003-2014)
Yichang	0.30	0.42	0.17	0.21	0.25	0.33
Luoshan	0.26	0.17	0.17	0.21	0.33	0.42
Hankou	0.26	0.17	0.17	0.21	0.25	0.42
Datong	0.28	0.25	0.17	0.30	0.42	0.50
Influence factors	Observations	Rainfall	Rainfall + TGD	Observations	Rainfall	Rainfall + TGD

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 615

616 **Table 2** The years when the flow regime changed during the post-dam period. Three cases are
 617 considered, which include Case 1: “Risk to none” denotes hydrological extremes from occurrence
 618 to none; Case 2: “No risk to risk” denotes hydrological extremes from none to occurrence; Case 3:
 619 “High/low flow event to low/high flow event” denotes hydrological extremes changed from high
 620 flow event to low flow event or from low flow event to high flow event. The signs “▲” and “▼” with
 621 different colors represent the high flow event and low flow event respectively, red color is Case 1,
 622 blue is Case 2, green is Case 3, black means no risk changed.

Year	Yichang		Luoshan		Hankou		Datong	
	modelled	observed	modelled	observed	modelled	observed	modelled	observed
2003	▲							
2004	▼	▼	▼				▼	▼
2005	▲	▲						
2006	▼	▼	▼	▼	▼	▼	▼	▼
2007							▼	
2008		▼		▼	▲	▼	▲	▼
2009			▼	▼	▼	▼	▼	▼
2010	▲		▲	▲	▲	▲	▲	▲
2011	▼	▼	▼	▼	▼	▼	▼	▼
2012	▲	▲		▲		▲		▲
2013				▼		▼	▲	▼
2014	▲		▲					

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624

625 **Figure 1** The Yangtze River Basin with the main river stem shown in bold and subbasins delineated
626 and annotated with names shown above the figure. The location of the rainfall stations used in the
627 analysis are shown in green. The four hydrological stations that recorded streamflow are shown in
628 red.

629 **Figure 2** Monthly average streamflow (bar plot) for each streamflow station and rainfall amount
630 averaged over the entire Yangtze River Basin.

631 **Figure 3** Rank correlation map between the spatial average rainfall over subbasins in summer (JJA)
632 and the log-transformed average streamflow for each hydrological station in summer (JJA) in
633 pre-dam period.

634 **Figure 4** Model results with 5th-95th prediction interval for the streamflows over the Yangtze River
635 Basin calibrated using available data (1960-2002) and the canonical variate of subbasin-scale
636 rainfall. The subplots in first row are the result of developed models using four pairs of canonical
637 variables, and second row is the result of observed and modelled streamflow for each hydrological
638 station.

639 **Figure 5** Boxplots of RE (left plot) and CE (right plot) from cross validation with results from the
640 model using sub-basin scale rainfall data.

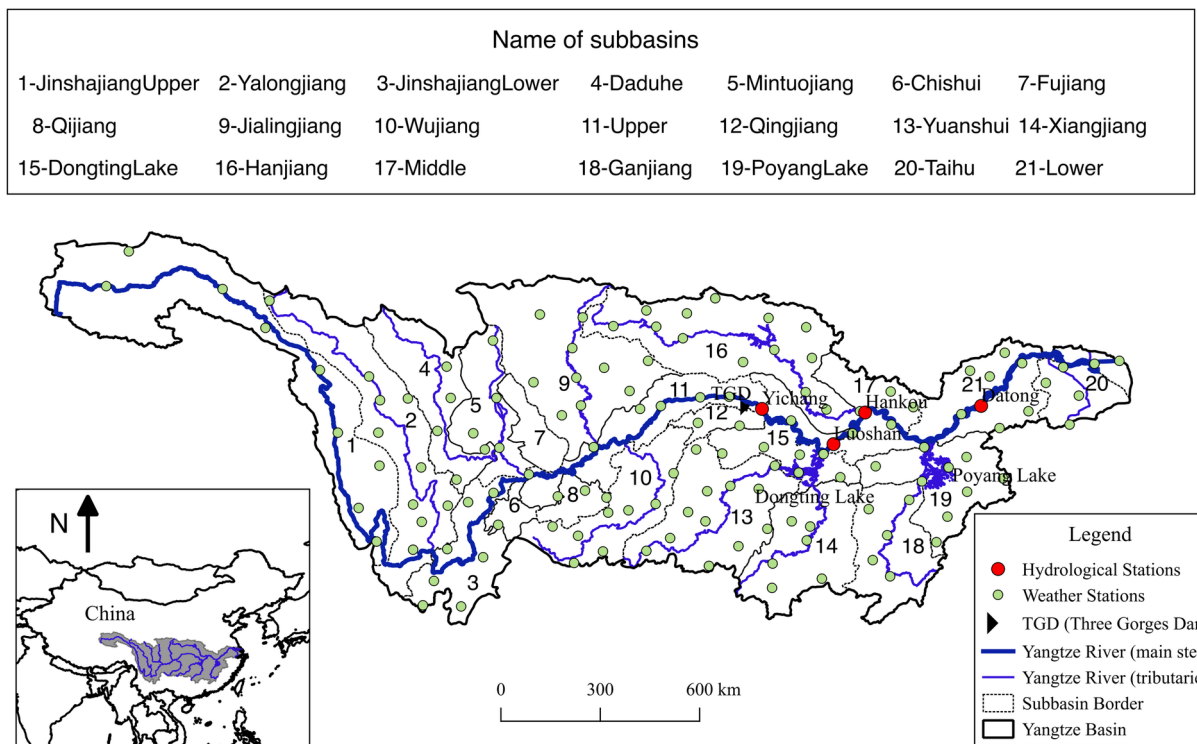
641 **Figure 6** Model results and observations with 5-yr moving average of the streamflow over Yangtze
642 River Basin. The horizontal dashed line is the average pre-dam period streamflow for each station
643 and the vertical dashed line represents the time before and after the dam starting to retain water (i.e.

644 Jun. 2003). The blue line with a marker denotes the modelled naturalized streamflow, and the black
645 line with a marker denotes the observed streamflow. The five-year moving averages were also
646 shown with green line and red line representing the modelled naturalized streamflow and observed
647 streamflow, respectively.

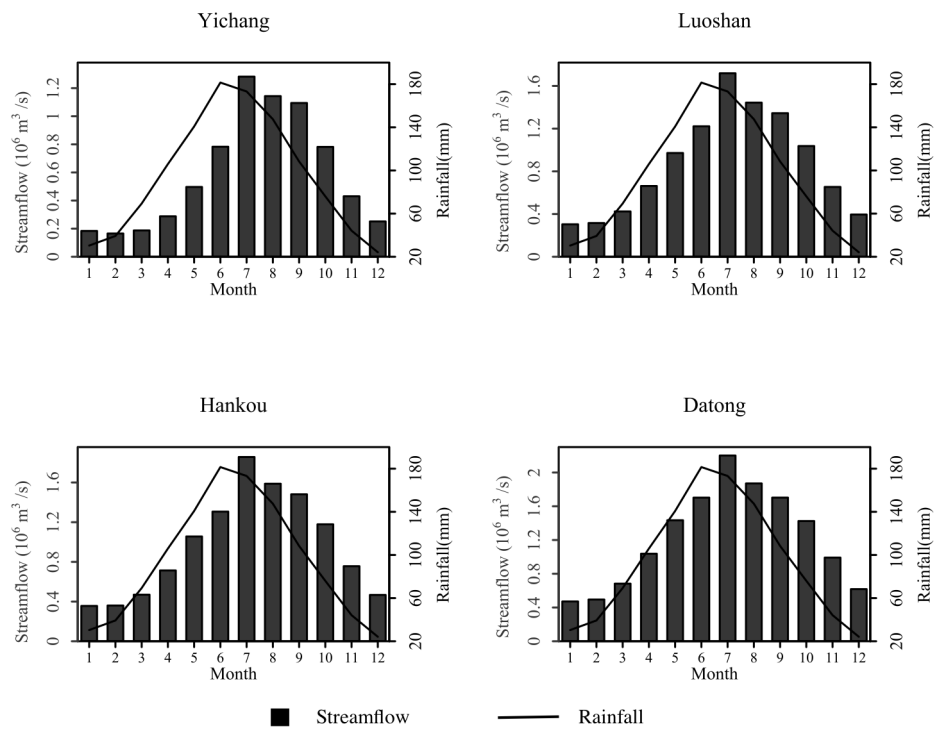
648 **Figure 7** The first two PCs of summer average streamflows modelled using CCA across the four
649 hydrological stations and their 5-yr moving average over Yangtze River Basin (left), and the
650 corresponding loadings for each station (right). The percentage variance explained is shown in the
651 parenthesis.

652 **Figure 8** Distribution of high flow and low flow events for the non-regulated flow from 1960 to
653 2014. Observed streamflows are used in the 1960-2002 period, while naturalized flows modelling
654 using canonical correlation analysis are used in the 2003-2014 period.

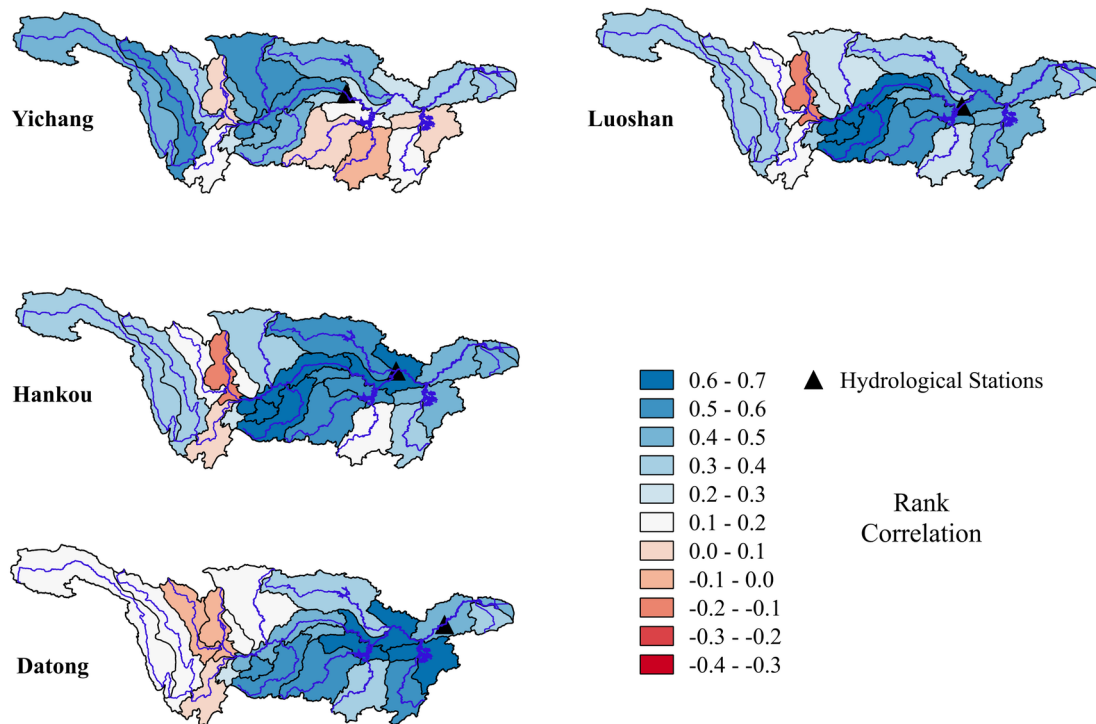
655 **Figure 9** Comparison of observed and modelled naturalized flows for the four streamflow stations
656 plotted as quantiles of pre-dam flows. The dashed line denotes the 75th percentile (high flows are
657 defined as flows in the top quartile) and the solid line denotes the 25th percentile (low flows are
658 defined as flows in the bottom quartile).



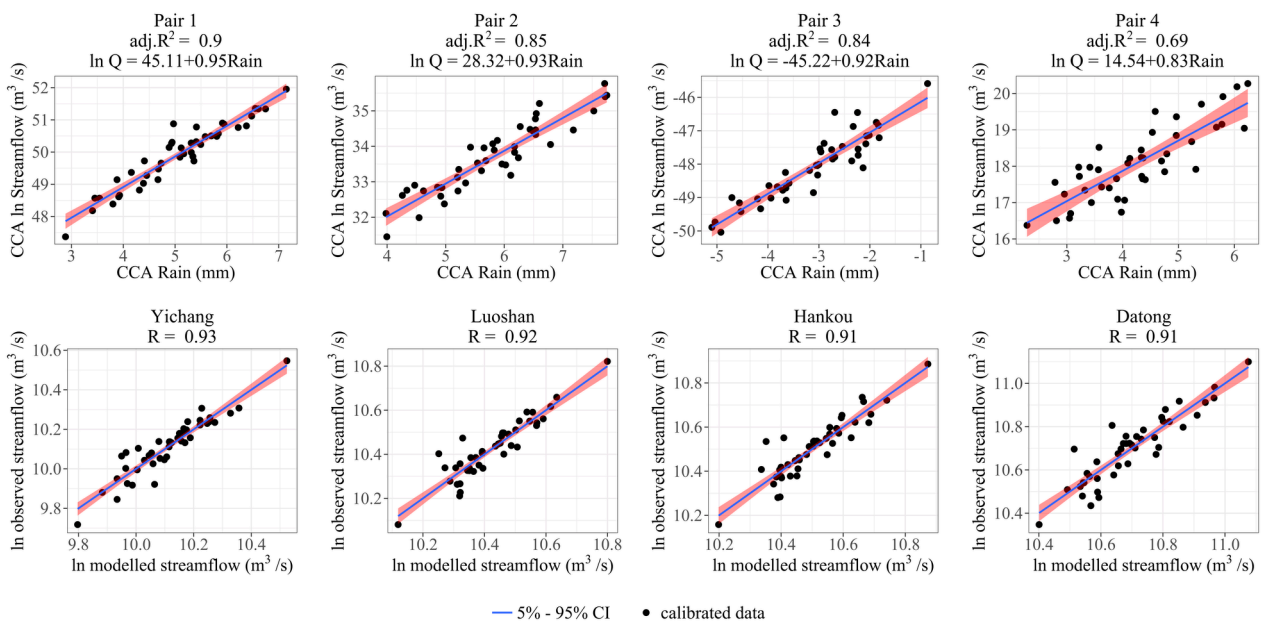
HYP_13619_Figure 1_final.tif



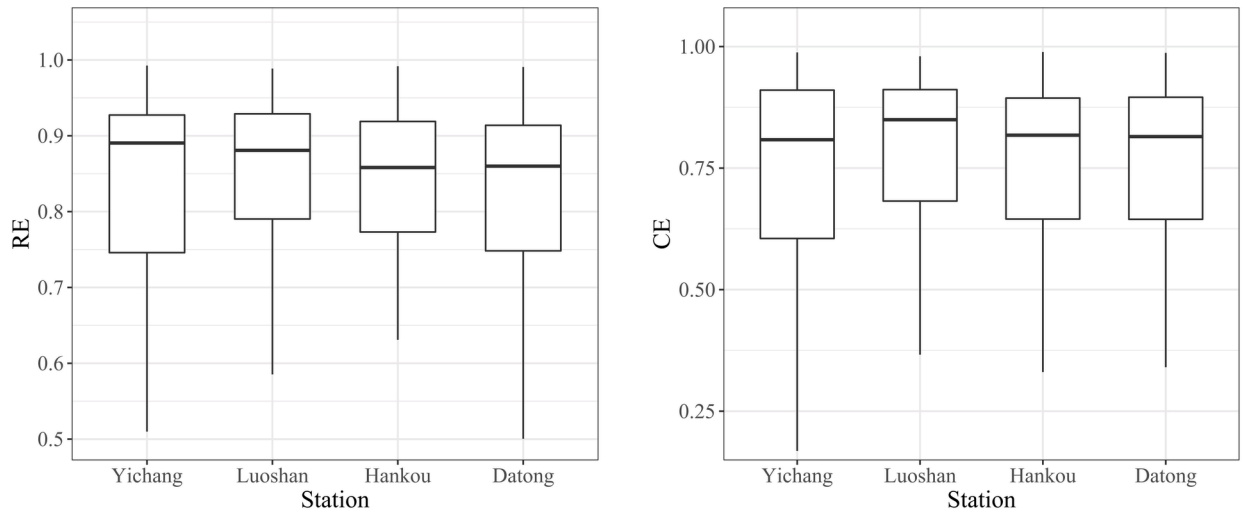
HYP_13619_Figure 2_final.tif



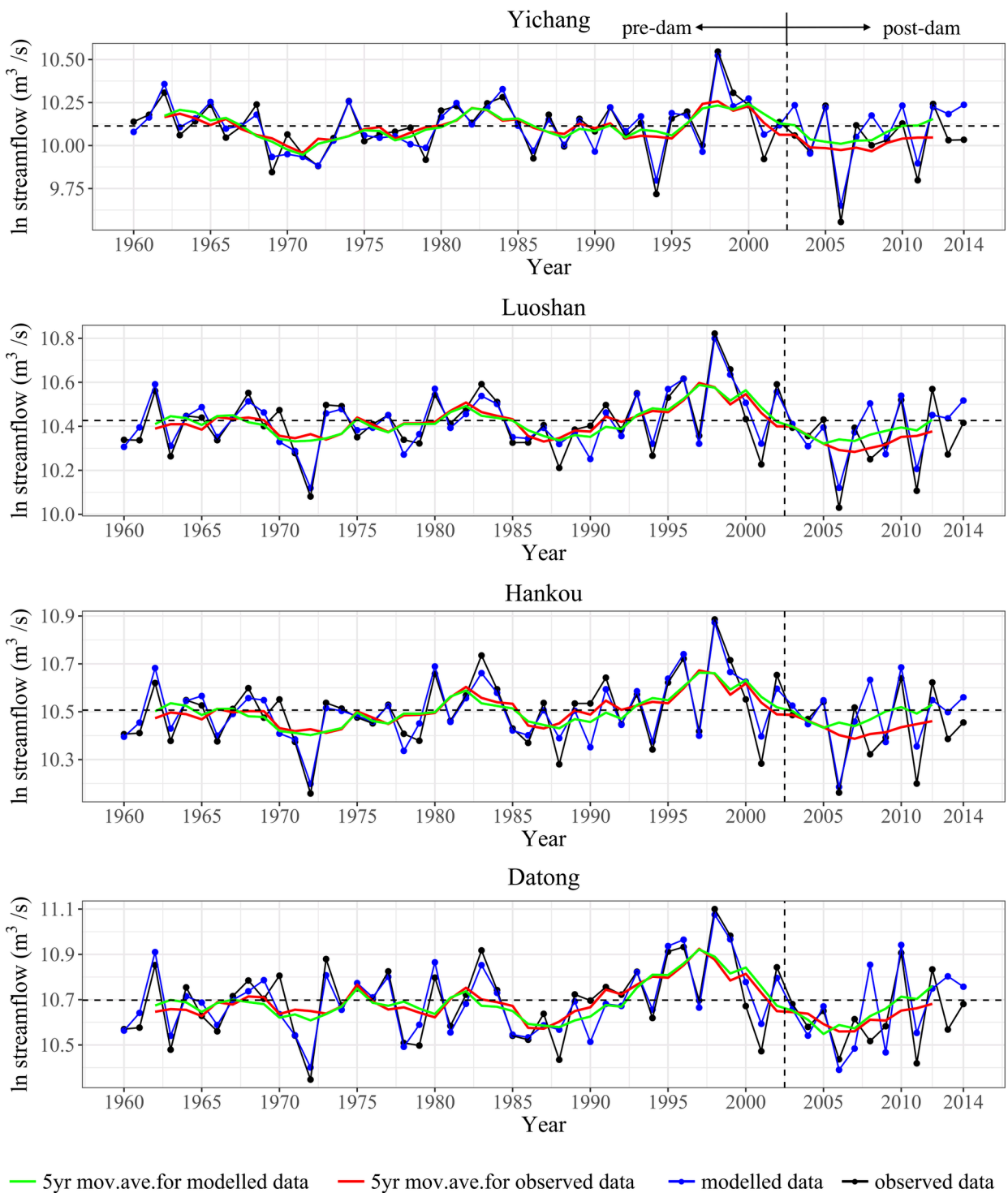
HYP_13619_Figure 3_final.tif



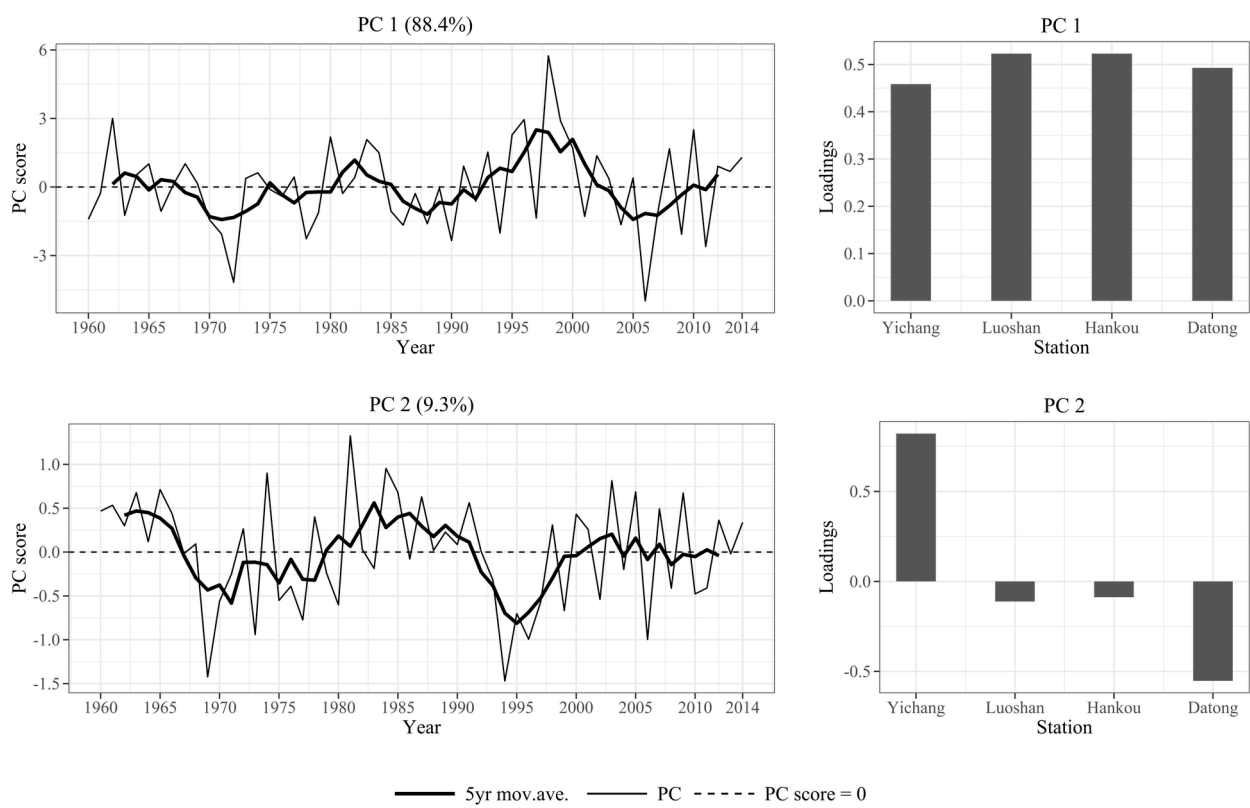
HYP_13619_Figure 4_final.tif



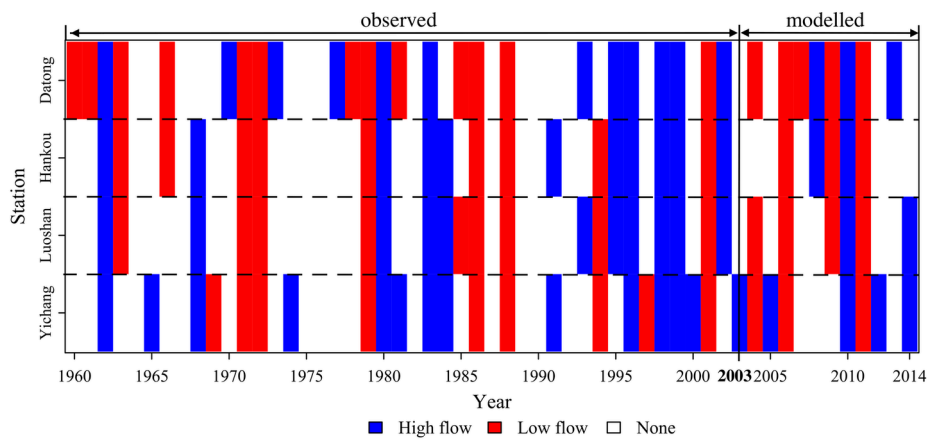
HYP_13619_Figure 5_final.tif



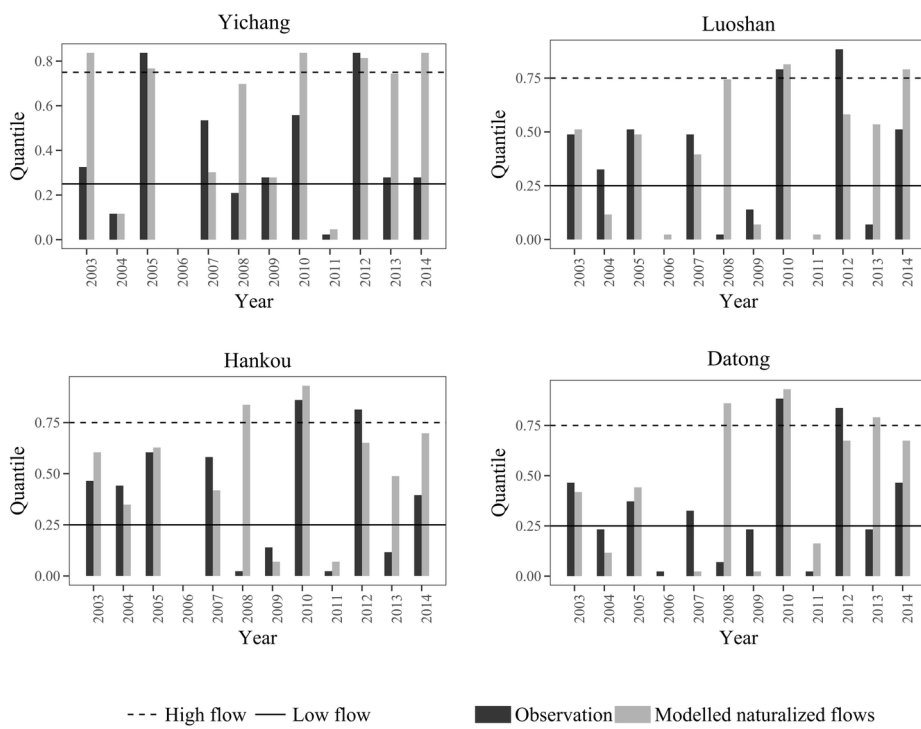
HYP_13619_Figure 6_final.tif



HYP_13619_Figure 7_final.tif



HYP_13619_Figure 8_final.tif



HYP_13619_Figure 9_final.tif