

1 **Impact of LUCC on Streamflow Based on the SWAT Model over the**
2 **Wei River Basin on the Loess Plateau of China**

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23 **Abstract:** Under the Grain for Green project in China, vegetation recovery constructions have
24 been widely implemented on the Loess Plateau for the purpose of soil and water conservation.
25 Now it becomes controversial whether the recovery constructions of vegetation, particularly forest,
26 is reducing streamflow in rivers of the Yellow River Basin. In this study, we choose the Wei River,
27 the largest branch of the Yellow River and implemented with revegetation constructions, as the
28 study area. To do that, we apply the widely used Soil and Water Assessment Tool (SWAT) model
29 for the upper and middle reaches of the - Wei River basin. The SWAT model was forced with daily
30 observed meteorological forcings (1960-2009), calibrated against daily streamflow for 1960-1969,
31 validated for the period of 1970-1979 and used for analysis for 1980-2009. To investigate the
32 impact of the LUCC (Land Use and land Cover Change) on the streamflow, we firstly use two
33 observed land use maps of 1980 and 2005 that are based on national land survey statistics emerged
34 with satellite observations. We found that the mean streamflow generated by using the 2005 land
35 use map decreased in comparison with that using the 1980 one, with the same meteorological
36 forcings. Of particular interest here, we found the streamflow decreased in agricultural land but
37 increased in forest area. More specifically, the surface runoff, soil flow and baseflow all decreased
38 in agricultural land, while the soil flow and baseflow of forest were increased. To investigate that,
39 we then designed five scenarios including (S1) the present land use (1980), (S2) 10%, (S3) 20%,
40 (S4) 40% and (S5) 100% of agricultural land was converted into mixed forest. We found that the
41 streamflow consistently increased with agricultural land converted into forest by about 7.4 mm per
42 10%. Our modeling results suggest that forest recovery constructions have positive impact on both
43 soil flow and base flow compensating reduced surface runoff, which leads to a slight increase in
44 streamflow in the Wei River with mixed landscapes of Loess Plateau and earth-rock mountain.

45 **1. Introduction**

46 Since 1999, China's Grain for Green project has greatly increased the vegetation cover
47 (Chen et al., 2015) and the total conversion area reaches 29.9 million ha until 2014 (Li, 2015).
48 And the proposals are to further return another 2.83 million ha farmland to forest and grassland by
49 2020 (NDRC, 2014). The establishment of either forest or grassland on degraded cropland has
50 been proposed as an effective approach to mitigating climate change because these types of land
51 use can increase soil carbon stocks (Yan et al., 2012; Deng et al., 2013). Implementation of large
52 scale Grain for Green project is undoubtedly one type of geoengineering which not only mitigates
53 climate change but also is expected to alter hydrological cycle (Lacombe et al., 2016; Lacombe et
54 al., 2008).

55 Some researchers have urged a cessation on Grain for Green expansion on the Loess Plateau
56 of China and argued that continued expansion of revegetation would cause more harm than good
57 to communities and the environment (Chen et al., 2015). One important reason was that the Grain
58 for Green project lead to annual streamflow of the Yellow River declining (Chen et al., 2015; Li,
59 2001). Land use change can disrupt the surface water balance and the partitioning of precipitation
60 into evapotranspiration, runoff, and groundwater flow (Sriwongsitanon and Taesombat, 2011;
61 Foley et al., 2005; Wagner et al., 2013). Large scale revegetation constructions change hydrologic
62 cycle process and distribution of water resources. There are three controversial points of view
63 about the impact of vegetation on streamflow as a whole. Quite a few catchment studies indicated
64 that annual streamflow decreased with revegetation increasing (Zhang and Hiscock, 2010; Bosch
65 and Hewlett, 1982; VanShaar et al., 2002; Mango et al., 2011; Farley et al., 2005; Liu and Zhong,
66 1978) or increased with vegetation destruction (Bosch and Hewlett, 1982; Woodward et al., 2014;

67 Hibbert, 2001), where some catchment studies indicated baseflow of forests was lower due to their
68 high evapotranspiration rates (Lørup et al., 1998; Lørup and Hansen, 1997; Smith and Scott, 1992),
69 while other studies indicated the baseflow increased in the dry season due to higher infiltration
70 and recharge of subsurface storage (the “sponge-effect hypothesis”) (Price, 2011; Lørup et al.,
71 1998; Ogden et al., 2013). In contrast, other studies showed that vegetation has a positive impact
72 on streamflow (Tobella et al., 2014; Li et al., 2001) or no impact on streamflow (Wang, 2000;
73 Beck et al., 2013).

74 To interpret the controversial results, it was argued that the impact of vegetation on annual
75 streamflow depends on watershed area and the relationship between them was negative in smaller
76 watershed and positive in larger watershed (Huang et al., 2009; Zhang, 1984). Some of them
77 thought it was probably the large amount of transpiration water played the main function in
78 hydrological process when the watershed was smaller. And some thought that the different impacts
79 of area probably because the forest of larger watershed could increase precipitation and vegetation
80 was also conducive for the infiltration of precipitation, which increased the proportion of the
81 underground flow of streamflow in forest region. Some researchers indicated tree planting has
82 both negative and positive effects on water resources and the overall effect was the result of a
83 balance between them, which were strongly dependant on tree density (Tobella et al., 2014).
84 Lacombe et al. (2016) found soil infiltrability was an important factor for explaining two modes of
85 afforestation (natural regeneration vs. planting) led to opposite changes in streamflow regime.
86 Huang (1982) analyzed Soviet research results found that 48% runoff coefficients increased, 32%
87 has no change, and 20% decreased with watershed forest increasing. The increased regions were
88 located at high latitude and humid areas. Under this condition, the total evaporation in wooded

89 areas and woodless area are equal. The speculation was that snow may be blown away or to
90 wooded areas from woodless area, which could enhance the coefficient of streamflow but these
91 factors would be weaker over low to middle latitude than that in high latitude (Huang, 1982).
92 Further, vegetation may change hydrological cycle as follows (Le Maitre et al., 1999): redirection
93 of precipitation by the canopy; branches, stem and litter tends to intercept more water into the soil;
94 roots may provide channels for the flow infiltrating to groundwater and extract soil water as
95 evaporation. Hence different results have led to contentious relationship between vegetation and
96 streamflow (Bradshaw et al., 2007; Dijk et al., 2009).

97 The Wei River is one main branch of the Yellow River and has been widely implemented
98 measures of soil and water conservation, including forestation, terraces, grass and check dam,
99 since the 1980s. Meanwhile the annual streamflow of the Wei River has decreased significantly
100 since the 1980s (Liu and Hu, 2006; Lin and Li, 2010; Wang et al., 2011). Since the 1990s, the
101 streamflow has sharply dropped and the observed streamflow of Linjiacun station in the 1990s was
102 less than one third of that before 1990s. The terrace and check dam both had a negative effect on
103 annual streamflow which was a result of the balance between the streamflow reducing in the flood
104 season and baseflow increasing in non-flood season on the Loess Plateau (Shao et al., 2013a; Xu
105 et al., 2012). But the impacts of vegetation on streamflow are controversial and complicated.
106 Meanwhile on the Loess Plateau, it was found that there is a drying layer of soil underneath forest
107 with a depth of over 1 m to 3 m from the soil surface owing to serious soil desiccation in
108 water-limited ecosystems (Li, 2001; Wang, 2010a). The land use, rainfall, soil type and slope
109 gradient had a significant impact on dried soil layer thickness (Wang, 2010b). And the great water
110 deficit prevents gravitational infiltration of rainfall and replenishment of groundwater. So forests

111 on the Loess Plateau reduced streamflow as the results of increased retention of rainfall and
112 reduced recharge into ground water (Li, 2001; Tian, 2010). But for earth-rock mountain landscape,
113 vegetation grows on thinner soil layer of rock mountain, which is apt to be saturated and produce
114 soil flow on relatively impermeable rock. So the streamflow in wooded areas might be larger than
115 that in adjacent woodless areas. Under this situation, forests may have positive impact for
116 producing streamflow (Liu and Zhong, 1978).

117 To investigate that, we develop hydrological experiments based on the widely used SWAT
118 model and observed hydrological/ meteorological data and land use data in the Wei River. We aim
119 at understanding possible impact of revegetation constructions, especially the forest restoration on
120 streamflow and its components in the Wei River, which is not only the largest branch of the
121 Yellow river but also with very mixed landscape with the loess plateau and earth-rock mountain.
122 In Sect. 2, we describe the study area and data. In Sect. 3, we set up, calibrate, and validate the
123 SWAT model in the Wei River. Section 4 reports the numerical experiment results, which is then
124 followed by the conclusion in Sect. 5.

125 **2. Study area and data**

126 **2.1 Study area**

127 Wei River is the largest tributary of the Yellow River, which originates from the north of the
128 Wushu Mountain at an altitude of 3495 m (involving Gansu, Ningxia and Shaanxi Provinces), and
129 runs across 818 km through into the Yellow River at Tongguan County, Shaanxi Province. In this
130 study, we choose the basin of the upper and middle reaches ($4.68 \times 10^4 \text{ km}^2$) of the Wei River basin
131 ($103.97^\circ \sim 108.75^\circ \text{ E}$, $33.69^\circ \sim 36.20^\circ \text{ N}$, $13.48 \times 10^4 \text{ km}^2$). And the Linjiacun, Weijiabu and
132 Xianyang hydrological stations are used from upstream to midstream in this study (Fig. 1), which

133 divided the study area into 3 regions. Linjiacun station locates at the control section of the
134 upstream and Xianyang station is the control station of middle reaches.

135 Geologically, the basin consists of the Loess Plateau and Qinling Mountain in the respective
136 north and south of the Wei River (Fig. 1). In the north, there are fewer tributaries, whose lengths
137 are further and the gradient is smaller. While in the south, abundant tributaries originate from
138 Qinling Mountain which is steep and close to the river. So the tributaries are shorter and the flows
139 are swifter. And there distribute lots of earth-rock mountain landscape and gravel riverbed in the
140 piedmont.

141 **2.2 Land Use and land Cover Change (LUCC) data**

142 We obtained observed LUCC data from National Science & Technology Infrastructure of
143 China, National Earth System Science Data Sharing Infrastructure (Fig.2) (<http://www.geodata.cn>).
144 Land use maps for the years of 1980 and 2005 were interpreted based on the corresponding
145 national land use survey data (1:100,000), satellite image, the MODIS data, 250-meter space
146 resolution data and combined with pasture resources map (1:500,000), soil type map (1:1,000,000),
147 vegetation type map (1:1,000,000) and other auxiliary data. The LUCC data were divided into six
148 types and further 25 subtypes. And the six types included forest, shrubland, pasture, cropland,
149 water bodies and residential areas: ① The forest type includes Range-Brush (RNGB),
150 Forest-Mixed (FRST), Forest-Deciduous (FRSD), Pine (PINE) and Forest-Evergreen (FRSE); ②
151 The pasture type includes Pasture (PAST), Winter Pasture (WPAS) and Range-Grasses (RNGE);
152 ③ The cropland means Agricultural Land (AGRL); ④ Water includes water (WATR) and
153 Wetlands-Mixed (WETL); ⑤ The residential areas include area of Residential-High Density
154 (URHD) and Residential-Medium Density (URMD); ⑥ The code of bare type is BARE. The

155 area of agricultural land decreased about 7.26% and forest area increased 0.81% in 2005 compared
156 with 1980 for the study area.

157 **2.3 Soil data**

158 Soil data were obtained from National Science & Technology Infrastructure of China,
159 National Earth System Science Data Sharing Infrastructure (Fig. 3(a)) (<http://www.geodata.cn>).
160 This soil data map reflects the distribution and characteristics of different soil type and digitized
161 based on 1:500,000 remote sensing digital figures of environment on Loess Plateau.

162 Based on the soil data, the distribution of earth-rock mountain in study area is drawn as Fig.
163 3(b). There were 83 soil types in the study area and 15 of them are composed of earth and rock
164 involving 70 hydrological response units (HRUs) (Table 1). At the same time, these 15 soil types
165 distribute mainly in the Qinling Mountain and Liupan Mountain (Fig. 1). And the earth-rock
166 mountain area accounts for 24% of study area.

167 **2.4 Meteorological and hydrological data**

168 The meteorological data were obtained from the China Meteorological Data Sharing Service
169 System (<http://www.eservice.gov.cn/metdata/page/index.html>) and some local rainfall stations.
170 The data include atmospheric pressure, mean (minimum and maximum) temperature, vapor
171 pressure, relative humidity, rainfall, wind speed, wind direction, sunshine time. Figure 4 (a) shows
172 the distribution of meteorological stations and the annual average precipitation over Wei River
173 basin, which was calculated using kriging interpolation method of ArcGIS 9.3 based on annual
174 average precipitation of 34 meteorological stations. Then the time series of annual average
175 precipitation for the three regions of the study area were calculated respectively using elevation
176 bands method of ArcSWAT (Soil and Water Assessment Tool) 2009.93.7b, which can account for

177 orographic effects on precipitation (Neitsch et al., 2011). SWAT allows the subbasin to be split
178 into a maximum of ten elevation bands. Precipitation is calculated for each elevation band as a
179 function of the respective lapse rate and the difference between the gage elevation and the average
180 elevation specified for the band. Once the precipitation values have been calculated for each
181 elevation band in the subbasin, new average subbasin precipitation value is calculated based on
182 the fraction of subbasin area within the elevation band (Neitsch et al., 2011). Figure 5 (b), (c) and
183 (d) show the time-series of average precipitation calculated through elevation bands method of
184 ArcSWAT from 1960 to 2009. The average of precipitation of region 1, 2 and 3 were 489.71
185 493.25 and 566.60 mm/yr and the trend analysis showed that the precipitation of them decreased
186 with an average decreasing rate of 0.57, 0.55 and 0.21 mm/yr, whereas the decreasing tendencies
187 were not significant at the 0.05 level.

188 And the daily streamflow data of three hydrological stations were obtained from Ecological
189 Environment Database of Loess Plateau (<http://www.loess.csdb.cn/pdmp/index.action>) and the
190 Hydrological Year books of China. Figure 5 (b), (c) and (d) show the time-series of annual
191 streamflow and runoff coefficients in the three regions of study area. The trend analysis showed
192 that streamflow of region 1 and 2 decreased extremely significantly ($P < 0.01$), with an average
193 decreasing rate of 1.74 and 5.38 mm/yr. The streamflow of region 3 did not decrease significantly.
194 And the average runoff coefficients were 0.13, 0.34 and 0.17 in region 1, 2 and 3 over the past 50
195 years (1960-2009). The trend analysis of runoff coefficients showed that the tendencies of region 1
196 and 2 decreased extremely significantly ($P < 0.01$), with an average decreasing rate of 0.34%, and
197 1.09 % per year. The runoff coefficient of region 3 decreased significantly ($P < 0.01$) too, with an
198 average decreasing rate of 0.2% per year.

199 90-meter resolution digital elevation model (DEM) (Fig. 4 (b)) was used to define the
200 topography characteristics (such as elevation, slope and aspect) and delineate the watershed
201 boundary. It was obtained from the Computer Network Information Center, Chinese Academy of
202 Sciences (<http://srtm.datamirror.csdb.cn/>), based on the Shuttle Radar Topography Mission (SRTM)
203 version 4.1.

204 **3. Methods**

205 **3.1 The SWAT model**

206 The SWAT model is developed by the USDA Agricultural Research Service (ARS). It is a
207 physically based and distributed hydrological model. The SWAT model has been widely applied to
208 understand the impact of land management practices on water, sediment and agricultural yields
209 over large complex watersheds with varying soils, land use and management conditions over long
210 periods (Arnold et al., 2009). It is forced with meteorological data, and input with soil properties,
211 topography, land use, and land management practices in the catchment. The physical processes
212 associated with hydrological cycle and sediment movement etc. are directly modeled by SWAT
213 using these input data (Arnold et al., 2009). In addition, the ArcSWAT extension (ArcSWAT
214 2009.93.7b version) is used as the graphical user interface for the SWAT model (Gassman et al.,
215 2007; Arnold et al., 1998). For the streamflow, surface runoff, soil and baseflow are considered.
216 Soil flow is streamflow contribution which originates below the surface but above the zone where
217 rocks are saturated with water. Base flow is the volume of streamflow originates from
218 groundwater (Arnold et al. 1993).

219 **3.2 The SWAT Model setup**

220 The SWAT model setup includes four steps: watershed delineation, hydrological response

221 unit (HRU) analyst, input database building and modification and model operation. Based on
 222 research of the Wei River (Shao, 2013b; Wang, 2013), the extraction threshold, which is the
 223 minimum drainage area required to form the origin of a stream, of subbasin area was 80 km². The
 224 Linjiacun, Weijiabu and Xianyang hydrological stations were loaded manually as subbasin outlets
 225 and one whole watershed outlet was defined. The study area was divided into 308 subbasins (Fig.
 226 1). The land area in a subbasin can be further divided into the HRUs, which is the basic computing
 227 element of the SWAT model. In this study, a subbasin was subdivided into only one HRU that was
 228 characterized by dominant land use and soil type. Then the daily meteorological data, including
 229 temperature, relative humidity, sunshine duration, wind speed, rainfall, were input and all data
 230 were written into database building and modification to force the SWAT model.

231 For evaluating the performance in the model calibration and validation, we use the R² and NS
 232 coefficient to evaluate the performance rating of the model (Nash and Sutcliffe, 1970) (Equation
 233 (1) & (2)).

$$234 \quad R^2 = \frac{\left[\sum_{i=1}^n \left(O_i^{obs} - \overline{O_i^{obs}} \right) \left(O_i^{sim} - \overline{O_i^{sim}} \right) \right]^2}{\sum_{i=1}^n \left(O_i^{obs} - \overline{O_i^{obs}} \right)^2 \left(O_i^{sim} - \overline{O_i^{sim}} \right)^2} \quad \text{Eq. (1)}$$

$$235 \quad NS = 1 - \frac{\sum_{i=1}^n \left(O_i^{obs} - O_i^{sim} \right)^2}{\sum_{i=1}^n \left(O_i^{obs} - \overline{O_i^{obs}} \right)^2} \quad \text{Eq. (2)}$$

236 where n is the number of observations, O^{obs} is the observed value, O^{sim} is the simulated value, and
 237 the overbar means the average of the variable. The R² describes the proportion of the variance in
 238 measured data explained by the model and typically 0.5 is considered an acceptable threshold
 239 (Santhi et al., 2001; Van Liew and Garbrecht, 2003). The SWAT model simulation can be judged
 240 as “satisfactory” if the NS > 0.50 for a monthly time step simulation and the performance rating of

241 the SWAT model was very good when the $NS > 0.75$, and the model performed good when the
242 $NS > 0.65$ (Moriassi et al., 2007).

243 **3.3 Calibration and validation of the SWAT model**

244 We setup the SWAT-CUP procedure for the sensitivity analysis, calibration and validation in
245 our study (Abbaspour, 2007). The sensitivity analysis is carried out by keeping all parameters
246 constant to realistic values, while varying each parameter within the range assigned in step one.
247 The sensitive parameters were calibrated using LH-OAT (Latin-Hypercube-One Factor-At-a-Time)
248 method of the Sequential Uncertainty Fitting (SUFI2) program (Abbaspour, 2007; Xu et al., 2012).
249 And the t-stat and p-value were used to evaluate the sensitivity of parameters. The t-stat is the
250 coefficient of a parameter divided by its standard error and the larger values are more sensitive.
251 And the p-value determines the significance of the sensitivity and a value close to zero means
252 more significant. The most sensitive (seven) parameters were selected by the SWAT-CUP module.
253 Combined with previous research in Wei River, two additional parameters (SOL_K and
254 GW_DELAY) with the seven parameters were selected in this study (Table 2).

255 The initial value and the range of relevant parameters were derived from previous research in
256 study area (Wang, 2014; Shao, 2013b; Zuo et al., 2015). Vegetation construction changes
257 undelaying surface and affects quantity of surface runoff and recharge of both soil and ground
258 water. It has a significant impact on infiltration by providing canopy and litter cover to protect the
259 soil surface from raindrop impacts and producing organic matter which can bind soil particles and
260 increase soil porosity (Le Maitre et al., 1999). These impacts of vegetation on hydrological
261 process are epitomized and reflect by CN and management operation in the SWAT model. the Soil
262 Conservation Service (SCS) curve number equation is the model for computing the amounts of

263 streamflow in SWAT model and its comprehensive parameter is CN which relates to the soil's
264 permeability, land use and antecedent soil water conditions. We have done some research on the
265 impacts of LUCC changes on runoff, infiltration and groundwater under different soil, slope and
266 rainfall intensity in Wei River basin based on simulated rainfall experiments before (Wang, 2014).
267 Based on the experiments, the SCS model and the three-dimensional finite-difference groundwater
268 flow model (MODFLOW) were calibrated and applied also. So values of parameters related to
269 runoff, infiltration and groundwater, such as the initial CN values and recharge rates for different
270 LUCC, specific yield of soil layer etc. were gotten based on experiments and mathematical
271 simulation (Wang, 2014). Meanwhile in the SWAT model, agricultural land and forest have
272 different heat units required for plant maturity and different management operations. The
273 agricultural land includes plant, harvest/ kill and auto-fertilizer operation and the forest only has
274 plant operation. And the management operation of forest involves leaf area index (LAT_INIT),
275 plant biomass (BIO_INIT), age of trees (CURYR_MAT).

276 The revegetation was mainly implemented in the study area after the 1980s. Hence we
277 choose 1960-1969 and 1970-1979 for the model calibration and validation respectively and used
278 the daily streamflow data of the Linjiacun, the Weijiabu and the Xianyang hydrological stations
279 from the upper to middle reaches (the data of 1965 and 1968-1971 are missing in the Weijiabu
280 station). The parameters were calibrated for hydrological stations by the order of upstream to
281 midstream using the daily streamflow of 1960-1969. Firstly, the parameters against the streamflow
282 at the Linjiacun control station were calibrated. Secondly, based on the premise of the calibrated
283 parameter values of the Linjiacun station, the parameters were calibrated for the subbasin
284 controlled by the Weijiabu station. In that way, the parameters for the subbasin controlled by the

285 Xianyang station were then calibrated. Then the SWAT model was validated for the three
286 hydrological stations respectively against the streamflow from 1970 to 1979 (Fig. 6).

287 **4. Results and discussions**

288 The corresponding statistic results of three hydrological stations showed that the ranges of
289 NS and R^2 were 0.59~0.66 and 0.63~0.68 respectively in the calibration period for a daily time
290 step. And they were 0.57~0.62 and 0.61~0.65 respectively in the validation period. At a monthly
291 time step, the results of the NS and R^2 were 0.82~0.84 and 0.79~0.86 respectively in the
292 calibration period. And they were 0.70~0.76 and 0.74~0.79 respectively in the validation period
293 demonstrating good performance of the model. In addition, the time-series and the patterns of the
294 simulated and observed streamflow during the calibration and validation periods showed similar
295 trends (Fig. 6). Our conclusion is that the SWAT model can be used in upper and middle reaches
296 of the Wei River basin.

297 **4.1 Impact of the observed LUCC on streamflow**

298 Analysis above (Fig. 5) showed that the observed precipitation of study area did not
299 decreased significantly from 1960 to 2009, while the annual streamflow (except region 3) and
300 runoff coefficients decreased significantly ($P < 0.05$) under this meteorological conation. This
301 discrepancy could attribute to LUCC changes mostly (Lacombe et al., 2016; Lacombe et al., 2008).
302 In order to analyze the impact of the LUCC on streamflow, the land use data of the 1980 and 2005
303 were used in the validated SWAT model and the DEM and soil data remained constant. Firstly,
304 the daily streamflow from 1980 to 2009 were simulated using observed daily meteorological
305 forcing data and topography, soil data in study area. Secondly, the LUCC data of 1980 was
306 replaced by that of 2005 and their relevant parameters of corresponding land use type were also

307 replaced. We used the LUCC data of 2005 but the same meteorological data to simulate the daily
308 streamflow from 1980 to 2009.

309 The change of annual streamflow based on LUCC data of 2005 compared with LUCC data of
310 1980 showed that annual streamflow of Xianyang hydrological station decreased during 20-year in
311 30-year ((1980-2009)) and the annual average reduction was 2.0 mm/yr for these 20-year in study
312 area. This result was consistent with the decreasing tendencies of the observed streamflow of
313 Xianyang station, which decreased significantly ($P < 0.05$), with an average decreasing rate of
314 2.45 mm/yr from 1980 to 2009. The modelled streamflow represent the impacts of constant LUCC
315 data of 1980 and 2005, whereas observations are based on dynamic LUCC data, which could
316 explain the discrepancy. Yin et al. (2017) studied the impact of LUCC changes on streamflow in
317 Jinghe River basin, which is the largest tributary of the Wei River basin, found that the streamflow
318 increasingly influenced by LUCC changes, which contributed to 44% of the streamflow changes
319 between the 1980s and 1990s and 71% of the streamflow changes between the 1990s and 2000s.
320 At the same time, different land use types hydrological responds differently even to the same
321 meteorological forcings, i.e., rainfall intensity was of great importance influencing to hydrological
322 process in semi-dry and semi-humid region (Lacombe et al., 2008; Wang, 2014). Results of
323 rainfall experiments showed when the rainfall intensity was smaller or larger, the rainfall would
324 infiltrate into soil or flow away as surface runoff mainly on both grass land and bare slope, while
325 when the rainfall intensity was medium, the rainfall would infiltrate into grass land and flowed
326 away as surface runoff on bare slope (Tobella et al., 2014; Wang, 2014). To reduce influence of
327 meteorological conditions and isolate the impact of the LUCC on streamflow, the 30-year
328 (1980-2009) values of the streamflow for forest and agricultural land were averaged respectively.

329 For period of 1980-2009, we just used their measured and long-term daily meteorological data in
330 the study area to drive the validated model for the designed hydrological experiments. Figure 7
331 shows the changes of streamflow, surface runoff, soil flow and baseflow between agricultural land
332 and forest. The surface runoff, soil flow and baseflow all decreased for agricultural land, while the
333 soil flow and baseflow of forest increased. Overall, the streamflow decreased in agricultural land
334 and increased in forest area. When the LUCC data are classified and re classified in SWAT model,
335 the tree types are summarized as Range-Brush (RNGB), Forest-Mixed (FRST) and
336 Forest-Deciduous (FRSD). Different types have different hydrological responses for their leaf,
337 roots and so on. We also analyzed the streamflow generation of the main types of forest (RNGB,
338 FRST and FRSD) in study area further. Results showed that the streamflow yield of FRST and
339 FRSD were about 1.20 and 1.60 times of that of RNGB respectively.

340 **4.2 Hydrological experiments on the impact of conversion of** 341 **agricultural land to forests on streamflow**

342 Because the LUCC data involves various land use interconversions, of particular interest here
343 the impact of conversion of cropland to forest on streamflow cannot be distinguished. Starting
344 from the LUCC data of 1980 as (S1) the present land use, we design other four scenarios (Table 3)
345 that (S2) 10%, (S3) 20%, (S4) 40% and (S5) 100% of the agricultural land was converted into
346 Forest-Mixed (FRST) respectively. And all experiments carried out based on the same the DEM,
347 soil data and meteorological conditions.

348 Based on the five scenarios, the SWAT simulations were conducted to analyze the effect of
349 forest constructions on the streamflow in upper and middle reaches of the Wei River basin. Firstly,
350 the converted agricultural land area was controlled proportionately as same as the variational area

351 ratios of set scenarios in 3 regions divided by Linjiacun, Weijiabu and Xianyang hydrological
352 stations (Fig. 5(a)). Secondly, lands with the same soil type and similar slope were the priorities
353 choosing as the converted land. Thirdly, the converted lands were distributed evenly as much as
354 possible in 3 regions. The simulation period was from 1980 to 2009.

355 We present the distribution of average streamflow change under S2 ~ S5 scenarios compared
356 with S1 scenario in Fig. 8. It shows that the streamflow generally increased when the land use
357 converted from agricultural land into forest in the upstream. And Fig. 9 shows the change rate of
358 streamflow at the Linjiacun, Weijiabu and Xianyang stations correspondingly for its annual
359 average and annual average over non-flood season (Jan - Jun and Nov - Dec). Compared with the
360 S1 scenario, the annual average streamflow increases in the non-flood season were 12.70 %,
361 11.21 % and 9.11% for the Linjiacun, Weijiabu and Xianyang stations with per 10% area of
362 agricultural land converted into forest. Interestingly the average annual streamflow increases were
363 11.61%, 21.63%, 42.51% and 109.25% for S2, S3, S4 and S5 scenario respectively (Fig. 9 (b)),
364 which almost consistently suggested about 1.1% per 1% change of the agricultural land. The
365 results are important in that one can expect that for a 0.8% increase in the forest in the observed
366 LUCC would lead to less than 1% change in the streamflow, which is negligible.

367 To be more comparable, Fig. 10 show the distribution of the annual runoff coefficients with
368 the scenario changed from S1 to S5. The spatial variability in mean runoff coefficient was large,
369 which ranges from 0.03 to 0.68 and increased with more forest converted from agricultural land.
370 The annual average runoff coefficient of study area increased from 0.21 to 0.37 with forest area
371 increasing from S1 to S5 (Fig. 11). On average, the runoff coefficient increased about 0.014 (i.e.,
372 1.4% of rainfall transformed into streamflow) with per 10% area of agricultural land converted

373 into forest.

374 The landscape of the Wei River is mixed with the Loess Plateau and earth-rock mountain
375 landscapes, which induce different mechanisms of transforming rainfall into streamflow. The
376 earth-rock mountain area accounts for 24.03% of study area (Fig. 3 (b)). In earth-rock mountain
377 area, vegetation grows on much thinner soil layer over the earth-rock mountain. And the soil has
378 high infiltration ability for high stone fragment content. The thin soil is apt to be saturated and
379 produce more soil flow on relatively impermeable rock, hence the streamflow in wooded areas is
380 larger than that in adjacent woodless areas favoring streamflow production (Liu and Zhong, 1978).
381 On the contrary, in Loess Plateau there is existing a drying layer of soil underneath forestland in
382 great water deficit. When the agricultural land converted into forest, the precipitation, intercepted
383 by vegetation, infiltrated into soil and supplied the drying layer of soil, vegetation growth, etc.
384 Together with much thicker soil layer on the Loess Plateau, it usually prevents gravitational
385 infiltration into groundwater and reduces streamflow recharge (Li, 2001; Tian, 2010). The
386 observed results of precipitation and streamflow in study area also showed the runoff coefficients
387 had obviously positive correlation with rates of earth-rock mountain area. The regional annual
388 averages of runoff coefficient were 0.13, 0.17 and 0.34 for Fig. 5 (b), (d) and (c), while the rates of
389 earth-rock mountain area were opposite correspondingly (Fig. 3 (b)). The complication is that the
390 overall effect of forest on the streamflow is in fact a balance between earth-rock mountain positive
391 and Loess Plateau negative effects on the streamflow.

392 Combined with the spatial distribution of precipitation (Fig. 4 (a)), we can see earth-rock
393 mountain landscapes are mainly distributed in regions with more rainfall. To be precise, the whole
394 earth-rock mountain area located where rainfall was greater than 500 mm/yr and over 62% of the

395 study area where the annual rainfall is greater than 600 mm was in earth-rock mountain.
396 Meanwhile, the river network over the earth-rock mountain is denser and most of tributaries in the
397 earth-rock mountain are close to the main stream of the Wei River. Moreover, there distribute a lot
398 of developed gravel riverbed in piedmont, sandy soil along the river and its groundwater level is
399 shallow, which facilitate rainfall infiltration and recharging streamflow. Therefore although the
400 area of earth-rock mountain accounts for 24% of the study area, its distribution areas are
401 concentrated in the main regions of streamflow yield of the study area. Therefore the overall result
402 of balance among all factors was that the forest constructions have a little positive effect on
403 streamflow in study area.

404 Seemingly, this result was not consistent with the significant decreasing tendencies of
405 streamflow in study area. The combined effects of LUCC, including forestation, terraces, grass,
406 and dam, could explain the discrepancy. Under the same meteorological conditions, the
407 streamflow is mainly a result of combined effects of these measures. Results showed the terrace in
408 the main Weihe River basin could delay the flood and add the drought season streamflow, which
409 reduced the annual streamflow in general. The terrace in 2000 could decrease about 37 million m³
410 annual water and increased the most dry month streamflow by 3.5% in Xianyang station (*Shao,*
411 *2013b*). Zhang et al (*2014a, 2014b*) studied the terrace measures of Yanhe River basin, typical
412 basin of the Loess Plateau, and results showed that the terrace measures could reduce the runoff in
413 the flood season and increased the baseflow. Results showed that 1 m³ water could be supplied to
414 the river when 5~ 6 m³ water stored by the terrace. This meant the water reducing effect of terrace
415 was larger than 80% in Yanhe River basin and. Xu et al. (2012) applied the SWAT model to
416 simulate the streamflow in the Yanhe basin and results showed that the check dams had a

417 regulation effect on streamflow. From 1984 to 1987, the streamflow in rainy season (from May to
418 October) decreased by $1.54 \text{ m}^3\text{s}^{-1}$ (14.7 %) to $3.13 \text{ m}^3\text{s}^{-1}$ (25.9 %) due to the check dams; while in
419 dry season (from November to the following April), streamflow increased by $1.46 \text{ m}^3\text{s}^{-1}$ (60.5%)
420 to $1.95 \text{ m}^3\text{s}^{-1}$ (101.2 %); From 2006 to 2008, the streamflow in rainy season decreased by 0.79
421 m^3s^{-1} (15.5 %) to $1.75 \text{ m}^3\text{s}^{-1}$ (28.9 %), and the streamflow in dry season increased by $0.51 \text{ m}^3\text{s}^{-1}$
422 (20.1 %) to $0.97 \text{ m}^3\text{s}^{-1}$ (46.4 %). Lots of results showed that the terrace and check dam both had a
423 negative effect on annual streamflow which was a result of the balance between the streamflow
424 reducing in the flood season and baseflow increasing in non-flood season on the Loess Plateau
425 (Shao, et al., 2012, 2013a, 2013b; Zhang, et al., 2014a, 2014b; Xu, et al., 2012). The observed
426 streamflow was a result of the balance among forestation, terraces, grass, and dam.

427 **4.3 Impact of conversion of agricultural land to forests on baseflow**

428 In Fig. 9 (a), one important point is that the average increase in the non-flood season was
429 about 1.41 times larger than the annual increase of the streamflow. To understand that, Fig. 12
430 shows distribution of the baseflow index, i.e., the ratio between baseflow and streamflow, under
431 S1~S5 scenarios. We can see that the baseflow index also increased with land use converted from
432 agricultural land into forest, which means that groundwater contribution to the streamflow
433 increased with the overall increase of forest area. Putting the pictures together, Fig. 13 shows the
434 changes of the streamflow and the baseflow under the S2~S5 scenarios minus those results under
435 the S1 scenario in the non-flood season. The average increases of streamflow and baseflow were
436 1.14 and 0.98 mm/yr with per 1% increase of forest area respectively. For the non-flood season,
437 they were 0.60 and 0.53 mm/yr. The increase of the streamflow contributed by the increased
438 baseflow was about 88.33% in the non-flood season. So the increasing streamflow was mainly

439 contributed by groundwater with increasing of forest area overall.

440 Although some researchers have urged a cessation on Grain for Green expansion on the
441 Loess Plateau of China for it lead to annual streamflow of the Yellow River declining (Chen et al.,
442 2015; Li, 2001), our modeling results suggest that forest recovery constructions have a little
443 positive impact on both soil flow and base flow compensating reduced surface runoff, which leads
444 to a slight increase in streamflow in the Wei River with mixed landscapes of Loess Plateau and
445 earth-rock mountain. And rainfall patter also has great effect on streamflow, particularly the
446 extremes rainfall, i.e., Lacombe et al. (2008) found no streamflow change was found for when the
447 precipitation was larger than 40 mm. Results showed that the daily precipitation extremes seem to
448 be consistent with the 7% increase per degree of warming (Allen and Ingram, 2002; Pall et al.,
449 2007) and one-hour precipitation extremes increase twice as fast with rising temperatures as
450 expected when daily mean temperatures exceed 12 °C (Lenderink and Meijgaard, 2008; Westra,
451 2014). The streamflow is the combined effects of LUCC (forestation, terraces, grass, and dam and
452 so on) and climate changes. The impact of Grain for Green project on streamflow should be
453 thoughtfully studied according to the characteristics of the basin.

454 At the same time, there are some uncertainties in SWAT model simulations. First, the SWAT
455 model could offer the comprehensive parameters for subbasin and detailed parameters for
456 different HRU according to their slopes, soli type and LUCC. The comprehensive parameters were
457 calibrated according to observed streamflow of subbasin, while the different parameters of HRU
458 could not be calibrated individually. Second, the model could not tell the impact of short-duration
459 rainfall on streamflow which has great effect on streamflow. In addition, watershed size,
460 generalization and data accuracy all can lead to uncertainty in the simulations (Yin et al., 2017).

461 To reduce the uncertainty of simulation influence, the 30-year (1980-2009) values of streamflow
462 were averaged to analyze the impacts.

463 **5. Conclusion**

464 The large scale implementation of Grain for Green project in China is expected to alter
465 hydrological cycle, in particular on the Loess Plateau, within the Yellow River Basin. The
466 scientific question is how large the impact of the LUCC on the streamflow and its components in
467 that area. We choose the Wei River as the study area, in that it has been widely implemented
468 revegetation constructions since the 1980s. Of particular interest here, the landscape of the upper
469 and middle reaches of the Wei River basin is mixed with the Loess Plateau and rocky mountain,
470 which would induce different mechanisms of generating surface runoff, soil flow, base flow and
471 therefore streamflow.

472 To investigate it, we setup the SWAT model for the upper and middle reaches of the Wei
473 River basin with the inputs of long term observed meteorological forcing data, hydrological data,
474 and observed land use data. We use daily and monthly streamflow of the Linjiacun, Weijiabu and
475 Xianyang hydrological stations from upper to middle reaches during 1960-1969 and 1970-1979
476 respectively for the model calibration and model validation. The results showed that the
477 Nash-Sutcliffe (NS) coefficients and the coefficients of determination (R^2) were > 0.57 and 0.61
478 for daily streamflow and 0.70 and 0.74 for monthly streamflow respectively demonstrating that
479 the SWAT model can be used in this study.

480 We analyze the impact of the LUCC on streamflow based on the observed LUCC data of
481 1980 and 2005. The daily streamflow from 1980 to 2009 were simulated using observed daily
482 meteorological data with the two different land use data. The results showed that two-thirds of

483 annual streamflow decreased and the change of streamflow was different among different land use.
484 On the overall average, the 30-year averages of the streamflow decreased in agricultural land but
485 increased in forest. To interpret the overall result, we design five scenarios in this study including
486 (S1) the present land use of 1980 and the scenarios where agricultural land was converted into
487 forest by 10% (S2), 20% (S3), 40% (S4) and 100% (S5) respectively. Based on the five scenarios,
488 we use the calibrated and validated SWAT model to analyze the effect of forest constructions on
489 the streamflow in detail. The results confirm that annual streamflow consistently increased with
490 more forest converted from the agricultural land. Interestingly, the rate is almost consistently 7.41
491 mm/yr per 10% increase of forest converted from the agricultural land. Based on detailed analysis
492 of each component of streamflow, we found it was most attributed by the baseflow. The overall
493 effect of LUCC on the streamflow in the Wei River basin, the largest branch of the Yellow River is
494 the result of the balance between Loess Plateau negative and earth-rock mountain positive effects.
495 Our results here are not only of great importance in understanding the impact of LUCC on
496 streamflow for a catchment with much complicated and mixed landscape, but also of significance
497 for water resources managing practice.

498 **Data availability**

499 The data used in this manuscript were obtained from reliable public data repositories. The
500 LUCC and soil data were obtained from the National Science & Technology Infrastructure of
501 China, the National Earth System Science Data Sharing Infrastructure (<http://www.geodata.cn>).
502 The DEM data were obtained from the Computer Network Information Center, the Chinese
503 Academy of Sciences (<http://srtm.datamirror.csdb.cn/>). The meteorological data were obtained
504 from the China Meteorological Data Sharing Service System

505 (<http://www.escience.gov.cn/metdata/page/index.html>). The daily streamflow data were from the
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665 **Figure Captions:**

666 **Fig.1** The study area: the Wei river basin on the Loess Plateau.

667 **Fig. 2** The observed land use data of the year 1980 and the year 2005 in study area.

668 **Fig. 3** The Soil data and the distribution of earth-rock mountain in study area.

669 **Fig. 4** The spatial distribution of annual average precipitation in Wei River basin over the past 55 years
670 (1956-2010) and the DEM of study area.

671 **Fig. 5** The time-series of precipitation, annual streamflow and runoff coefficients for the regions of
672 study area.

673 **Fig.6** The time-series graphs of calculated vs. observed values during calibration period and
674 verification period for hydrological stations.

675 **Fig. 7** The changes of 30-year (1980-2009) averages of streamflow, surface runoff, soil flow and
676 baseflow between agricultural land and forest.

677 **Fig. 8** The watershed distribution of average streamflow change under S2~S5 scenarios compared with
678 S1 scenario.

679 **Fig. 9** The corresponding proportional change rate of streamflow at Linjiacun, Weijiabu and Xianyang
680 station for annual average and annual average in non-flood season.

681 **Fig. 10** The distribution of annual runoff coefficient with the scenario changed from S1 to S5.

682 **Fig. 11** The annual average runoff coefficient of study area with forest area increasing from S1 to S5.

683 **Fig. 12** The distribution of baseflow index under S1~S5 scenarios.

684 **Fig. 13** The corresponding change of streamflow and baseflow under S2~S5 scenarios compared with
685 S1 for annual average of year and non-flood season.

686

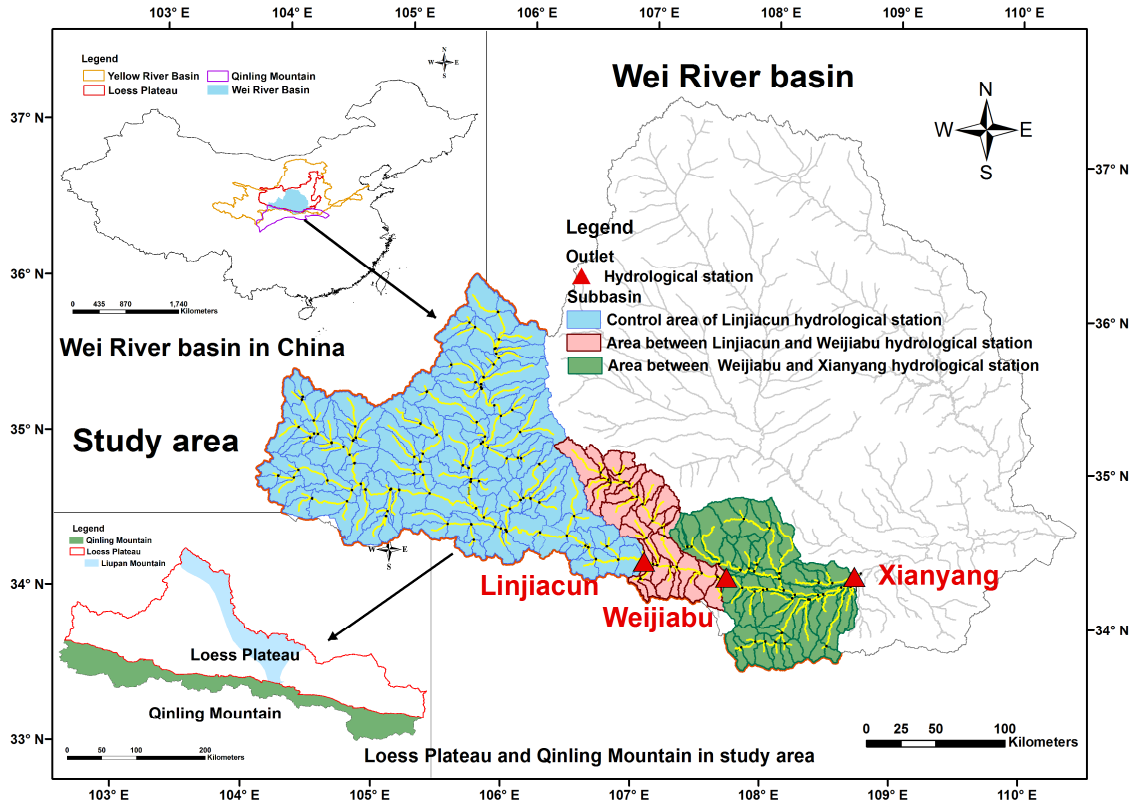


Fig. 1 The study area: the Wei river basin on the Loess Plateau.

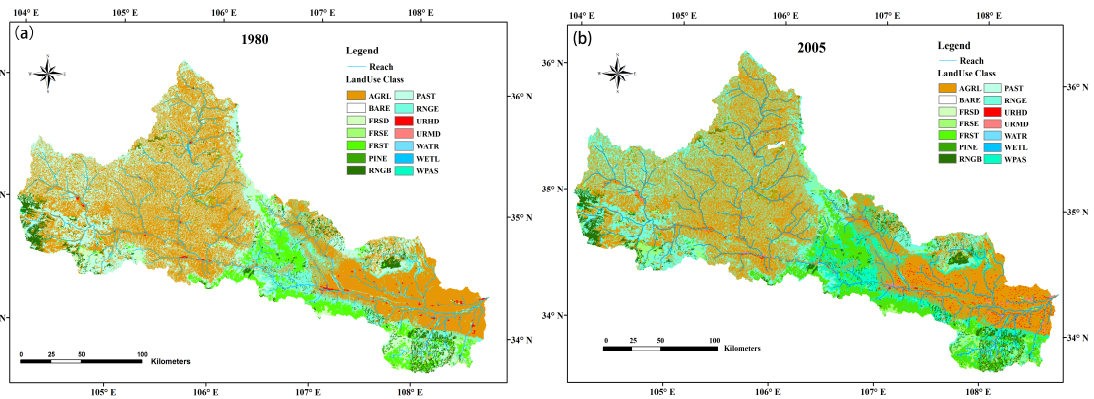


Fig. 2 The observed land use data of the year 1980 and the year 2005 in study area

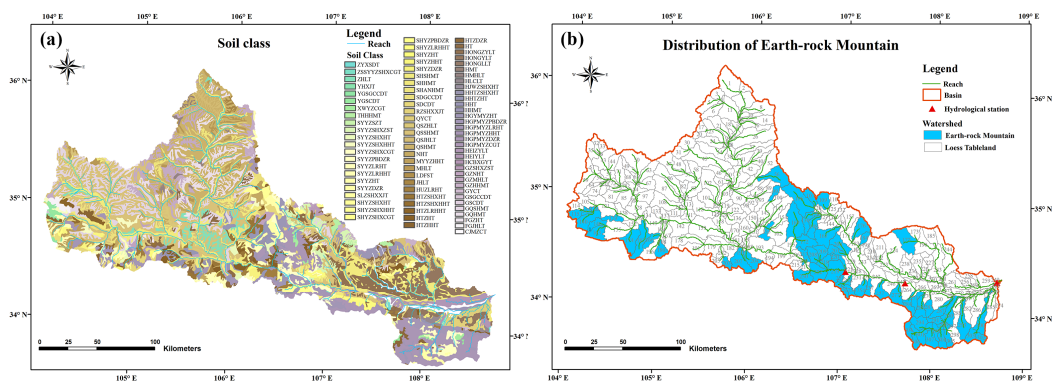
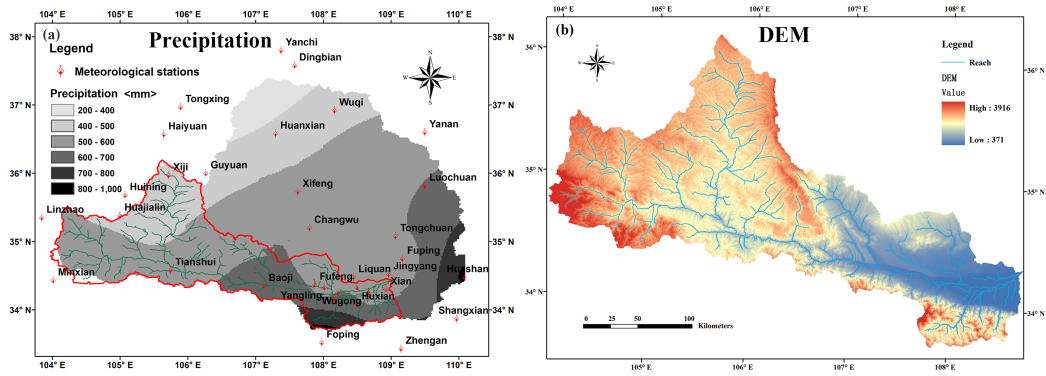
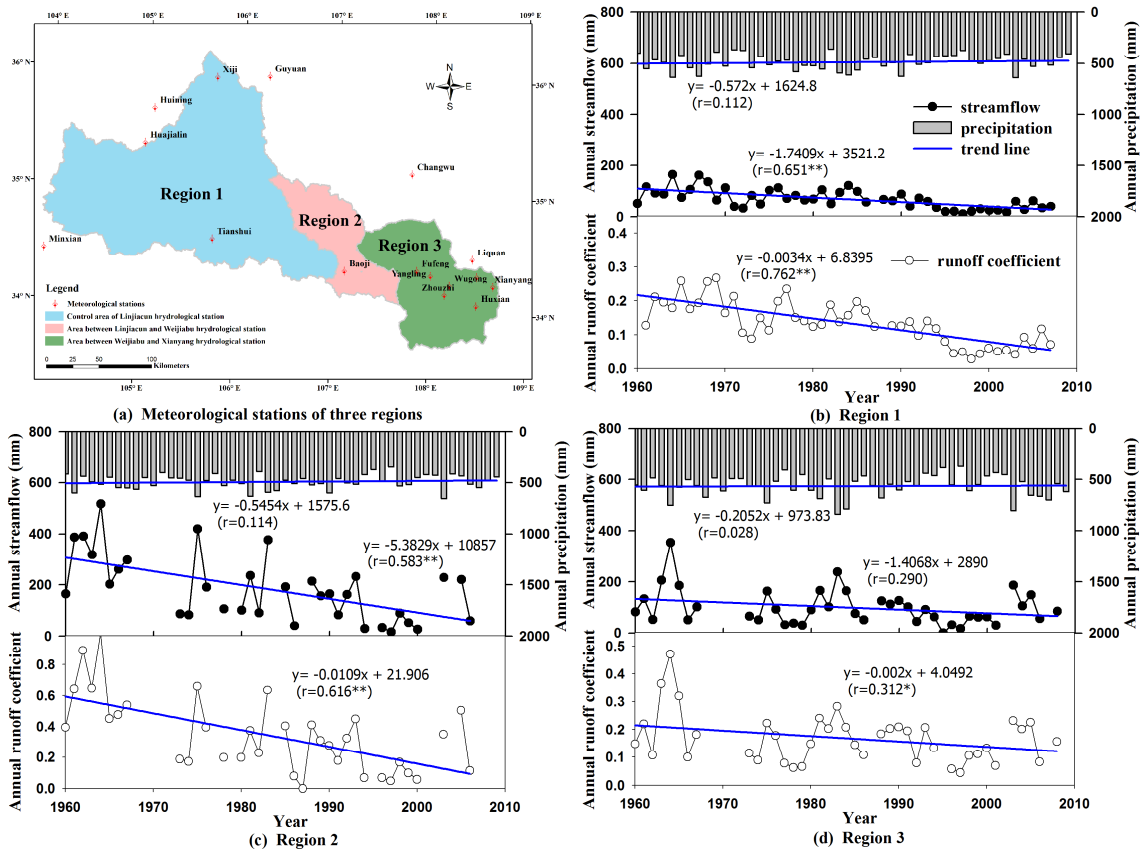


Fig. 3 The Soil data and the distribution of earth-rock mountain in study area



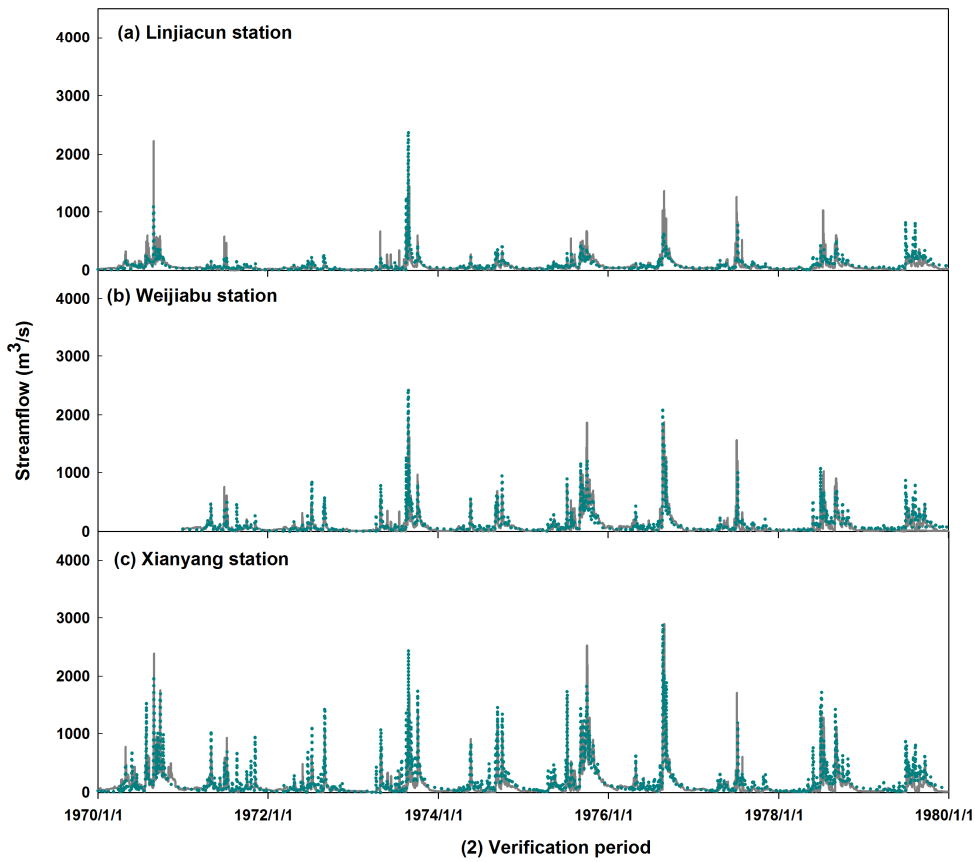
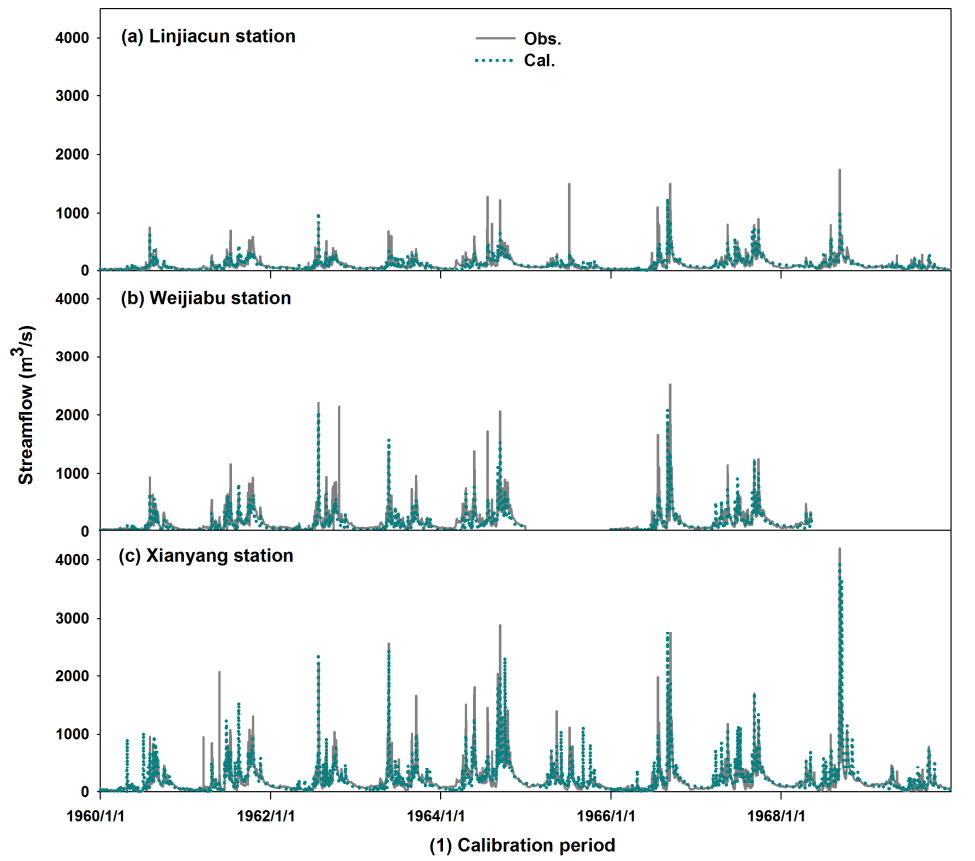
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694 **Fig. 4** The spatial distribution of annual average precipitation in Wei River basin over the past 55 years
 695 (1956-2010) and the DEM study area



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697 **Fig. 5** The time-series of precipitation, annual streamflow and runoff coefficients for the regions of
 698 study area



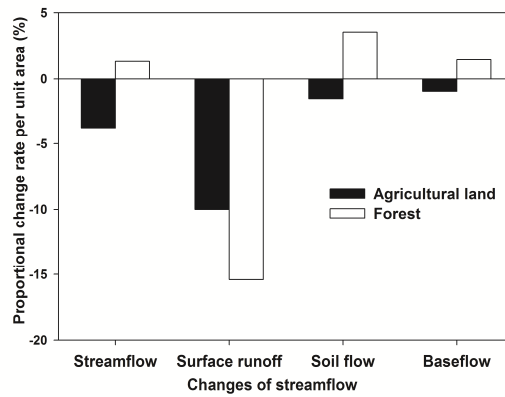
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Fig. 6 The time-series graphs of calculated vs. observed values during calibration period and verification

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period for hydrological stations



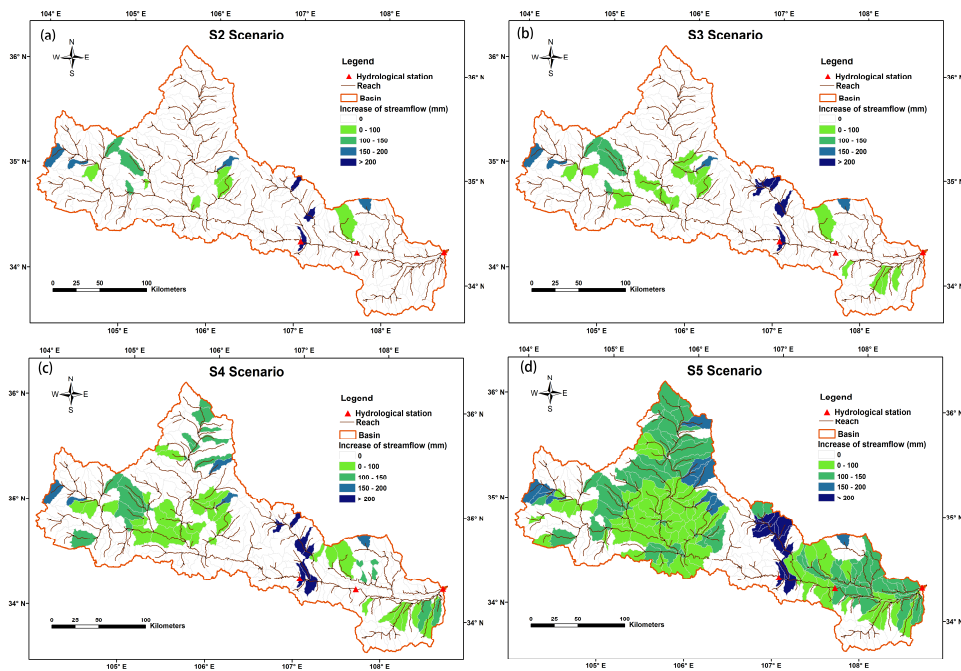
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Fig. 7 The changes of 30-year (1980-2009) averages of streamflow, surface runoff, soil flow and

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baseflow between agricultural land and forest



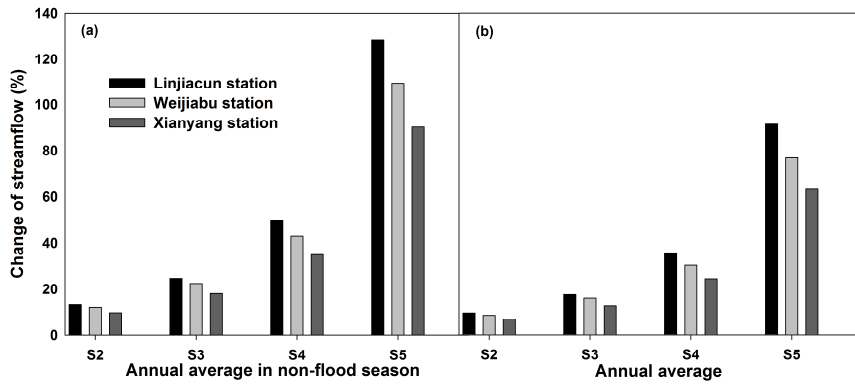
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Fig. 8 The watershed distribution of average streamflow change under S2~S5 scenarios compared with

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S1 scenario



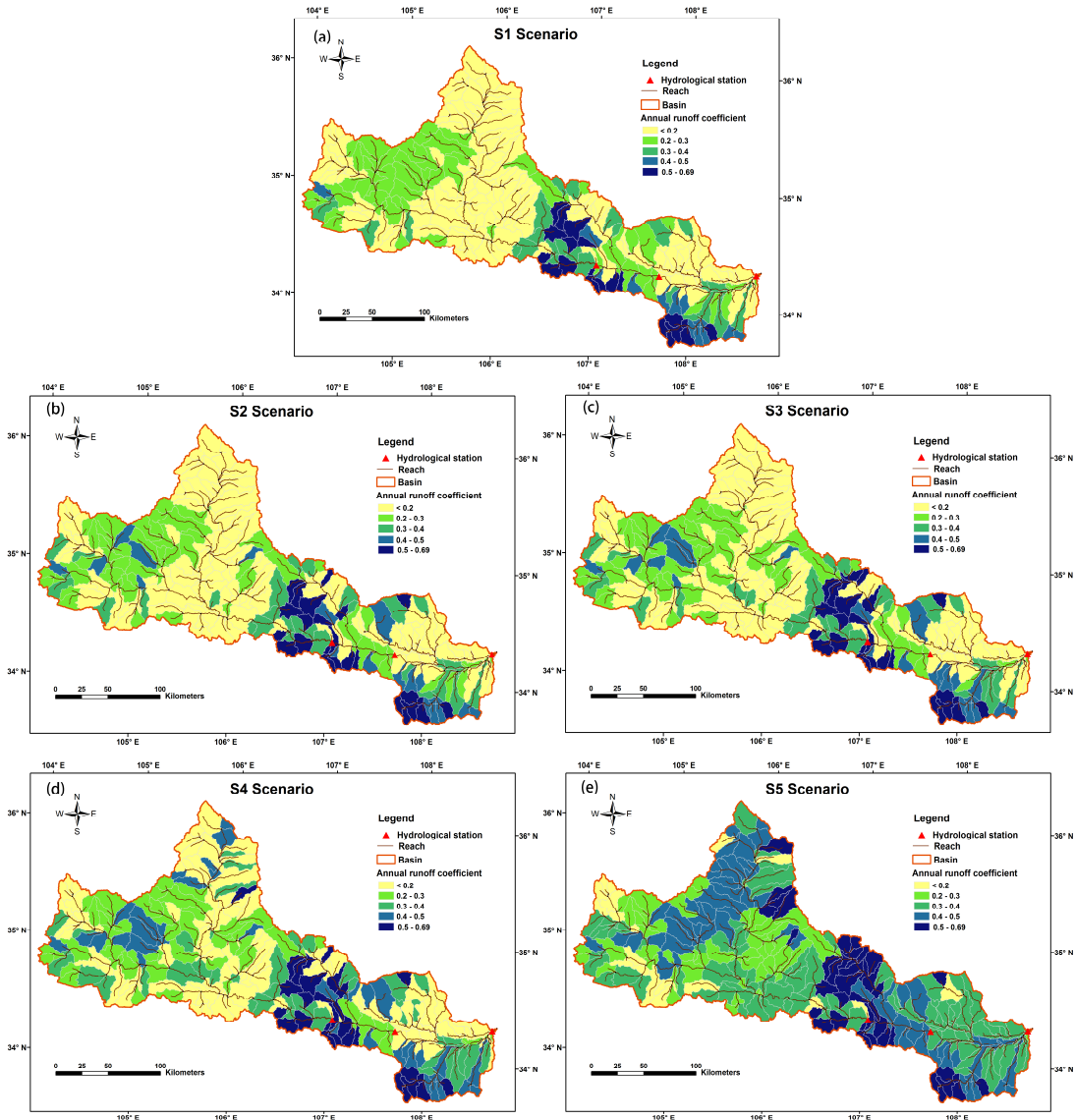
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Fig.9 The corresponding proportional change rate of streamflow at Linjiacun, Weijiabu and Xianyang

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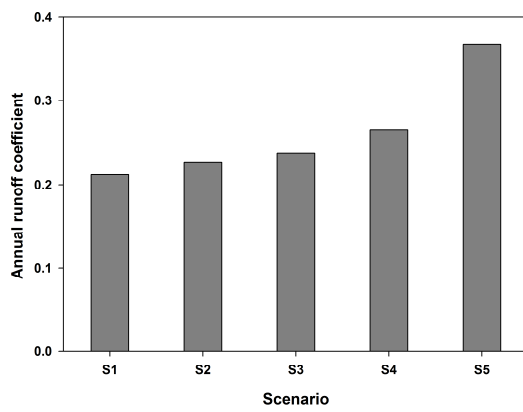
station for annual average and annual average in non-flood season



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Fig. 10 The distribution of annual runoff coefficient with the scenario changed from S1 to S5

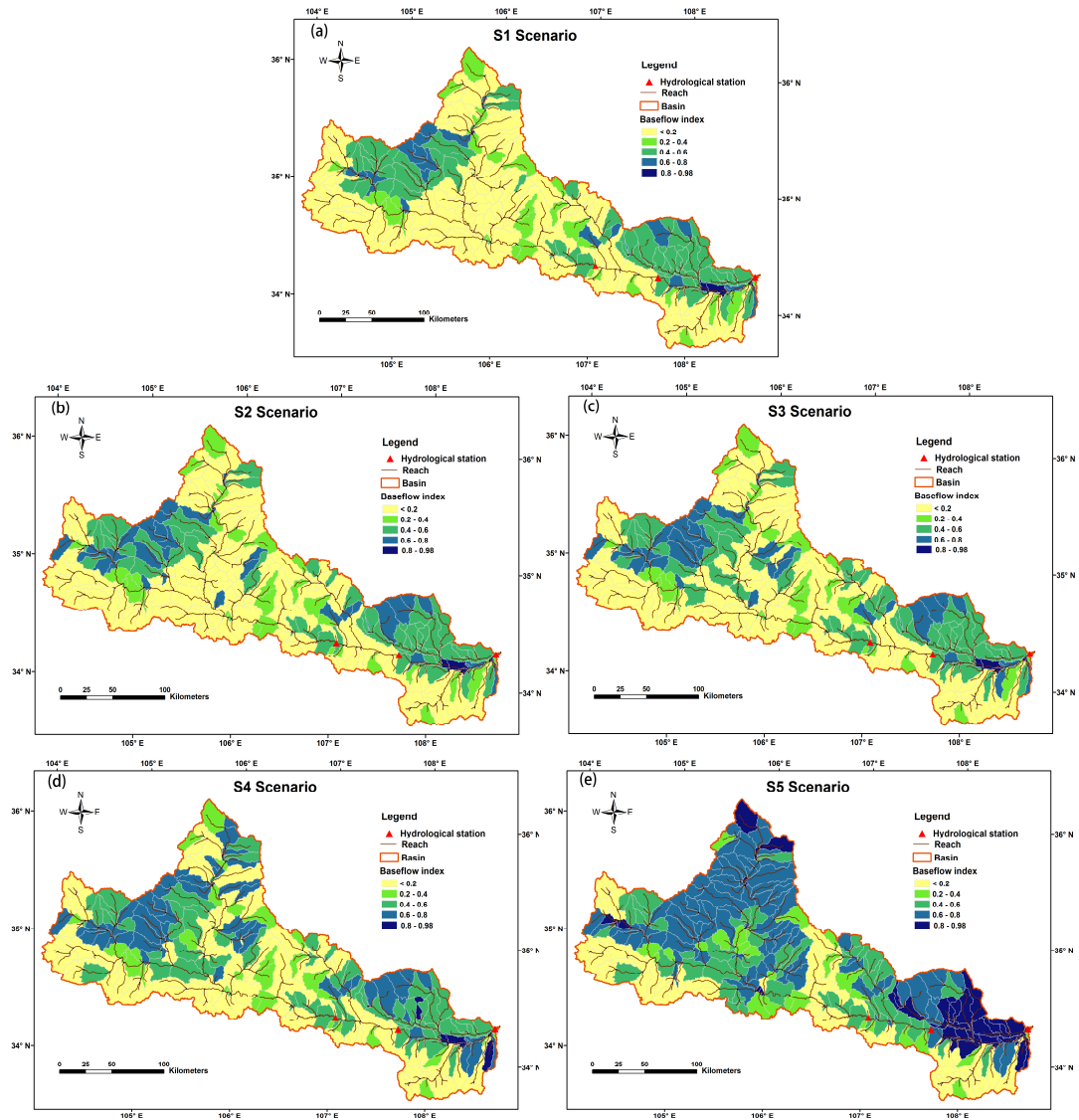


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Fig. 11 The annual average runoff coefficient of study area with forest area increasing from S1 to S5

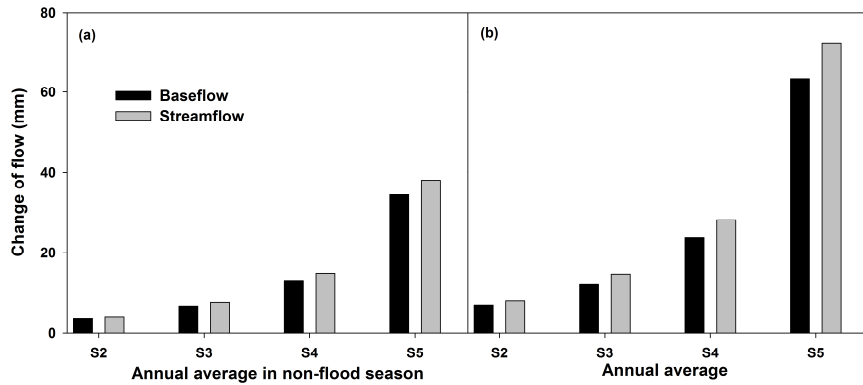
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Fig. 12 The distribution of baseflow index under S1~S5 scenarios



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719 **Fig.13**The corresponding change of streamflow and baseflow under S2~S5 scenarios compared with

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S1 for annual average of year and non-flood season

721 **Tables**

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Table 1 The soil type and its distribution of earth-rock mountain in study area

No.	Code of Soil type	Physical meaning of the code	HRU	Area (km ²)
1	SHYZHT	Limestone Cinnamon soil	220, 257	26316.90
2	SHYZSHXHT	Limestone Calcic cinnamon soil	153	11471.22
3	SYYZLRHT	Sandstone—shale Luvic cinnamon soil	166, 203, 207	50065.29
4	HGPMYZLRHT	Granite—gneiss Luvic cinnamon soil	174, 180, 187, 201, 204, 221, 277, 283, 287	158397.93
5	SYYZDZR	Sandstone—shale Light brown earth	106, 169, 299	103955.40
6	HGPMYZDZR	Granite—gneiss Light brown earth	130, 148, 172, 209, 252, 284, 289, 290, 291, 293, 294, 300, 301, 302, 303, 305, 306, 307, 308	299737.26
7	HGPMYZPBDZR	Granite—gneiss Light brown earth	253	8739.90
8	MYYZHHT	Sandstone—shale Grey cinnamon soil	115, 117, 146, 163	51204.96
9	SYYZSHXHHT	Sandstone—shale Calcic grey cinnamon soil	99, 129	19392.21
10	SHYZSHXHHT	Limestone Calcic Grey cinnamon soil	56	33885.54
11	SYYZSHXZST	Sandstone—shale Purple soil	109, 176, 177, 184, 200	106159.41
12	HGPMYZCGT	Granit—gneiss Rhogosol	165, 230, 237, 254, 271, 292, 295, 296, 297, 304	112136.40
13	SYYZSHXCGT	Sandstone—shale Rhogosol	107, 208, 213, 216, 218, 219, 248	87612.84
14	SHYZSHXCGT	Limestone Rhogosol	222	23375.79
15	SYYZLRHHT	Sandstone—shale Luvic grey	116, 140	30320.73

		cinnamon soil		
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Table 2 Calibrated values of model parameters

Parameters	Physical meaning	Calibration range	Calibration result		
			Linjiac un	Weijia bu	Xianya ng
r_CN2	Initial SCS runoff curve number for moisture condition II	-0.3~0.3	-0.27	0.05	-0.17
r_SOL_AWC	Available water capacity of soil layer	-0.6~0.6	0.01	-0.01	-0.01
r_SOL_K	Saturated hydraulic conductivity of soil layer (mm/hr)	-0.5~0.5	0.5	0.3	0.5
r_HRU_SLP	Average slope stepness (m/m)	-0.5~1.5	1.5	0.41	0.52
r_SLSUBBSN	Average slope length (m)	-0.5~1.5	1.17	0.70	1.20
v_ALPHA_BF	Baseflow alpha factor	0~1.0	0.48	0.61	0.61
v_GW_DELAY	Groundwater delay (days)	0~500	220	38	62
v_ESCO	Soil evaporation compensation factor	0~1.0	0.65	0.90	0.80
v_CH_K2	Effective hydraulic conductivity in main channel alluvium	0~130	5	30	30

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Notes: v__ means the existing parameter value is to be replaced by the given value; r__ means the existing parameter value is multiplied by (1+ a given value).

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Table 3 Scenarios for simulation

Scenario	Description	Area (km ²)	The average simulated streamflow
			(1980-2009) (10 ⁸ m ³ /yr)
S 1	present situation	0	50.44

S 2	10% agricultural land → forest	2937.63	53.92
S 3	20% agricultural land → forest	5875.26	56.83
S 4	40% agricultural land → forest	11750.53	62.73
S 5	100% agricultural land → forest	29376.32	82.28

729 Notes: ① Agricultural land refers to the land for crops planting, including cultivated land, newly cultivated soil, fallow field,
730 rotation plot, pasture-crop rotation and land used for agro-fruit, agro-mulberry, agroforestry (The code in model is AGRL). ②
731 Forest refers to the natural forest and plantation, which canopy density is larger than 30%, including timberland, economic forest,
732 protection forest (The code in model is FRST).
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