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 scalar 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 and Hewlett, 1982; VanShaar et al., 2002; Mango et al., 2011; Farley et al., 2005; Liu and Zhong, 65

Hibbert, 2001), where some catchment studies indicated baseflow of forests was lower due to their
high evapotranspiration rates (Lørup et al., 1998; Lorup and Hansen, 1997; Smith and Scott, 1992),
while other studies indicated the baseflow increased in the dry season due to higher infiltration
and recharge of subsurface storage (the "sponge-effect hypothesis") (Price, 2011; Lørup et al.,
1998; Ogden et al., 2013). In contrast, other studies showed that vegetation has a positive impact
on streamflow (Tobella et al., 2014; Li et al., 2001) or no impact on streamflow (Wang, 2000;
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 since the 1980s (Fig. 1). Meanwhile the annual
99	streamflow of the Wei River has decreased significantly since the 1980s (Liu and Hu, 2006; Lin
100	and Li, 2010; Wang et al., 2011). Since the 1990s, the streamflow has sharply dropped and the
101	observed streamflow of Linjiacun station in the 1990s was less than one third of that before 1990s.
102	The terrace and check dam both had a negative effect on annual streamflow which was a result of
103	the balance between the streamflow reducing in the flood season and baseflow increasing in
104	non-flood season on the Loess Plateau (Shao et al., 2013a; Xu et al., 2013). But the impacts of
105	vegetation on streamflow are controversial and complicated. Meanwhile on the Loess Plateau, it
106	was found that there is a drying layer of soil underneath forest with a depth of over 1 m to 3 m
107	from the soil surface owing to serious soil desiccation in water-limited ecosystems (Li, 2001;
108	Wang, 2010a). The land use, rainfall, soil type and slope gradient had a significant impact on dried
109	soil layer thickness (Wang, 2010b). And the great water deficit prevents gravitational infiltration
110	of rainfall and replenishment of groundwater. So forests on the Loess Plateau reduced streamflow

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

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

# 124 **2. Study area and data**

### 125 **2.1 Study area**

Wei River is the largest tributary of the Yellow River, which originates from the north of the Wushu mountain at an altitude of 3495 m (involving Gansu, Ningxia and Shaanxi Provinces), and runs across 818 km through into the Yellow River at Tongguan County, Shaanxi Province. In this study, we choose the basin of the upper and middle reaches  $(4.68 \times 10^4 \text{ km}^2)$  of the Wei River basin  $(103.97^{\circ} \sim 108.75^{\circ} \text{ E}, 33.69^{\circ} \sim 36.20^{\circ} \text{ N}, 13.48 \times 10^4 \text{ km}^2)$ . And the Linjiacun, Weijiabu and Xianyang hydrological stations are used from upstream to midstream in this study (Fig. 2), which divided the study area into 3 regions. Linjiacun station locates at the control section of the

133 upstream and Xianyang station is the control station of middle reaches.

Geologically, the basin consists of the Loess Plateau and Qinling Mountain in the respective north and south of the Wei River (Fig. 2). In the north, there are fewer tributaries, whose lengths are further and the gradient is smaller. While in the south, abundant tributaries originate from Qinling Mountain which is steep and close to the river. So the tributaries are shorter and the flows are swifter. And there distribute lots of earth-rock mountain landscape and gravel riverbed in the piedmont.

## 140 **2.2 Land Use and land Cover Change (LUCC) data**

141 We obtained observed LUCC data from National Science & Technology Infrastructure of 142 China, National Earth System Science Data Sharing Infrastructure 3) (Fig. 143 (http://www.geodata.cn). Land use maps for the years of 1980 and 2005 were interpreted based on 144 the corresponding national land use survey data (1:100,000), satellite image, the MODIS data, 145 250-meter space resolution data and combined with pasture resources map (1:500,000), soil type 146 map (1:1,000,000), vegetation type map (1:1,000,000) and other auxiliary data. The LUCC data 147 were divided into six types and further 25 subtypes. And the six types included forest, shrubland, 148 pasture, cropland, water bodies and residential areas: ① The forest type includes Range-Brush 149 (RNGB), Forest-Mixed (FRST), Forest-Deciduous (FRSD), Pine (PINE) and Forest-Evergreen 150 (FRSE); 2 The pasture type includes Pasture (PAST), Winter Pasture (WPAS) and Range-Grasses (RNGE); ③ The cropland means Agricultural Land (AGRL); ④ Water includes 151 152 water (WATR) and Wetlands-Mixed (WETL); ⑤ The residential areas include area of 153 Residential-High Density (URHD) and Residential-Medium Density (URMD); ⑥ The code of 154 bare type is BARE. The area of agricultural land decreased about 7.26% and forest area increased 155 0.81% in 2005 compared with 1980 for the study area.

#### 156 **2.3 Soil data**

157 Soil data were obtained from National Science & Technology Infrastructure of China,

- 158 National Earth System Science Data Sharing Infrastructure (Fig. 4(a)) (http://www.geodata.cn).
- 159 This soil data map reflects the distribution and characteristics of different soil type and digitized

160 based on 1:500,000 remote sensing digital figures of environment on Loess Plateau.

161 Based on the soil data, the distribution of earth-rock mountain in study area is drawn as Fig.

- 162 4(b). There were 83 soil types in the study area and 15 of them are composed of earth and rock
- involving 70 hydrological response units (HRUs) (Table 1). At the same time, these 15 soil types
  distribute mainly in the Qinling Mountain and Liupan Mountain (Fig. 2). And the earth-rock
  mountain area accounts for 24% of study area.

#### 166 **2.4 Meteorological and hydrological data**

167 The meteorological data were obtained from the China Meteorological Data Sharing Service 168 System (http://www.escience.gov.cn/metdata/page/index.html) and some local rainfall stations. 169 The data include atmospheric pressure, mean (minimum and maximum) temperature, vapor 170 pressure, relative humidity, rainfall, wind speed, wind direction, sunshine time. Figure 5 (a) shows 171 the distribution of meteorological stations and the annual average precipitation over Wei River 172 basin, which was calculated using kriging interpolation method of ArcGIS 9.3 based on annual 173 average precipitation of 34 meteorological stations. Then the time series of annual average 174 precipitation for the three regions of the study area were calculated respectively using elevation 175 bands method of ArcSWAT (Soil and Water Assessment Tool) 2009.93.7b, which can account for 176 orographic effects on precipitation (Neitsch et al., 2011). SWAT allows the subbasin to be split into a maximum of ten elevation bands. Precipitation is calculated for each elevation band as a function of the respective lapse rate and the difference between the gage elevation and the average elevation specified for the band. Once the precipitation values have been calculated for each elevation band in the subbasin, new average subbasin precipitation value is calculated based on the fraction of subbasin area within the elevation band (Neitsch et al., 2011).

- 182 And the daily streamflow data of three hydrological stations were obtained from Ecological
- 183 Environment Database of Loess Plateau (http://www.loess.csdb.cn/pdmp/index.action) and the
- 184 Hydrological Year books of China. Figure 6 shows the time-series of average precipitation, annual
- 185 streamflow and runoff coefficients for the 3 regions of study area. And the runoff coefficients were
- 186 0.13, 0.35 and 0.17 on average for region 1, 2 and 3 over the past 50 years (1960-2009).
- 90-meter resolution digital elevation model (DEM) (Fig. 5 (b)) was used to define the
  topography and delineate the watershed boundary. It was obtained from the Computer Network
  Information Center, Chinese Academy of Sciences (http://srtm.datamirror.csdb.cn/), based on the
  Shuttle Radar Topography Mission (SRTM) version 4.1.

## 191 **3. Methods**

#### 192 **3.1 The SWAT model**

The SWAT model is developed by the USDA Agricultural Research Service (ARS). It is a physically based and distributed hydrological model. The SWAT model has been widely applied to understand the impact of land management practices on water, sediment and agricultural yields over large complex watersheds with varying soils, land use and management conditions over long periods (Arnold et al., 2009). It is forced with meteorological data, and input with soil properties, topography, land use, and land management practices in the catchment. The physical processes associated with hydrological cycle and sediment movement etc. are directly modeled by SWAT
using these input data (Arnold et al., 2009). In addition, the ArcSWAT extension (ArcSWAT
2009.93.7b version) is used as the graphical user interface for the SWAT model (Gassman et al.,
2007; Arnold et al., 1998).

**3.2 The SWAT Model setup** 

204 The SWAT model setup includes four steps: watershed delineation, hydrological response unit (HRU) analyst, input database building and modification and model operation. Based on 205 research of the Wei River (Shao, 2013b; Wang, 2013), the extraction threshold, which is the 206 minimum drainage area required to form the origin of a stream, of subbasin area was 80 km<sup>2</sup>. The 207 208 Linjiacun, Weijiabu and Xianyang hydrological stations were loaded manually as subbasin outlets 209 and one whole watershed outlet was defined. The study area was divided into 308 subbasins (Fig. 210 2). The land area in a subbasin can be further divided into the HRUs, which is the basic computing 211 element of the SWAT model. In this study, a subbasin was subdivided into only one HRU that was 212 characterized by dominant land use and soil type. Then the daily meteorological data, including 213 temperature, relative humidity, sunshine duration, wind speed, rainfall, were input and all data 214 were written into database building and modification to force the SWAT model.

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

218 
$$R^{2} = \frac{\left[\sum_{i=1}^{n} \left(\boldsymbol{O}_{i}^{obs} - \overline{\boldsymbol{O}_{i}^{obs}}\right) \left(\boldsymbol{O}_{i}^{sim} - \overline{\boldsymbol{O}_{i}^{sim}}\right)\right]^{2}}{\sum_{i=1}^{n} \left(\boldsymbol{O}_{i}^{obs} - \overline{\boldsymbol{O}_{i}^{obs}}\right)^{2} \left(\boldsymbol{O}_{i}^{sim} - \overline{\boldsymbol{O}_{i}^{sim}}\right)^{2}}$$
Eq. (1)

219 
$$NS = 1 - \frac{\sum_{i=1}^{n} \left( O_{i}^{obs} - O_{i}^{sim} \right)^{2}}{\sum_{i=1}^{n} \left( O_{i}^{obs} - \overline{O_{i}^{obs}} \right)^{2}}$$
Eq. (2)

where n is the number of observations,  $O^{obs}$  is the observed value,  $O^{obs}$  is the simulated value, and the overbar means the average of the variable. The R<sup>2</sup> describes the proportion of the variance in measured data explained by the model and typically 0.5 is considered an acceptable threshold (Santhi et al., 2001; Van Liew and Garbrecht, 2003). The SWAT model simulation can be judged as "satisfactory" if the NS > 0.50 for a monthly time step simulation and the performance rating of the SWAT model was very good when the NS > 0.75, and the model performed good when the NS > 0.65 (Moriasi et al., 2007).

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

228 We setup the SWAT-CUP procedure for the sensitivity analysis, calibration and validation in 229 our study (Abbaspour, 2007). The sensitivity analysis is carried out by keeping all parameters 230 constant to realistic values, while varying each parameter within the range assigned in step one. 231 The sensitive parameters were calibrated using LH-OAT (Latin-Hypercube-One Factor-At-a-Time) 232 method of the Sequential Uncertainty Fitting (SUFI2) program (Abbaspour, 2007; Xu et al., 2012). 233 And the t-stat and p-value were used to evaluate the sensitivity of parameters. The t-stat is the 234 coefficient of a parameter divided by its standard error and the larger values are more sensitive. 235 And the p-value determines the significance of the sensitivity and a value close to zero means 236 more significant. The most sensitive (seven) parameters were selected by the SWAT-CUP module. Combined with previous research in Wei River, two additional parameters (SOL K and 237 GW DELAY) with the seven parameters were selected in this study (Table 2). 238

239	The initial value and the range of relevant parameters were derived from simulated rainfall
240	experiments, regional monitoring data and previous research in study area (Wang, 2014; Shao,
241	2013b; Zuo et al., 2015). Vegetation construction changes undelaying surface and affects quantity
242	of surface runoff and recharge of both soil and ground water. It has a significant impact on
243	infiltration by providing canopy and litter cover to protect the soil surface from raindrop impacts
244	and producing organic matter which can bind soil particles and increase soil porosity (Le Maitre et
245	al., 1999). These impacts of vegetation on hydrological process are epitomized and reflect by CN
246	and management operation in the SWAT model. the Soil Conservation Service (SCS) curve
247	number equation is the model for computing the amounts of streamflow in SWAT model and its
248	comprehensive parameter is CN which relates to the soil's permeability, land use and antecedent
249	soil water conditions. We have done some research on the impacts of LUCC changes on runoff,
250	infiltration and groundwater under different soil, slope and rainfall intensity in Wei River basin
251	based on simulated rainfall experiments before (Wang, 2014). Based on the experiments, the SCS
252	model and the three-dimensional finite-difference groundwater flow model (MODFLOW) were
253	calibrated and applied also. So values of parameters related to runoff, infiltration and groundwater,
254	such as the initial CN values and recharge rates for different LUCC, specific yield of soil layer etc.
255	were gotten based on experiments and mathematical simulation (Wang, 2014). Meanwhile in the
256	SWAT model, agricultural land and forest have different heat units required for plant maturity and
257	different management operations. The agricultural land includes plant, harvest/ kill and
258	auto-fertilizer operation and the forest only has plant operation. And the management operation of
259	forest involves leaf area index (LAT_INIT), plant biomass (BIO_INIT), age of trees
260	(CURYR_MAT).

261 According to Fig. 1, we could see the revegetation was mainly implemented in the study area after the 1980s. Hence we choose 1960-1969 and 1970-1979 for the model calibration and 262 263 validation respectively and used the daily streamflow data of the Linjiacun, the Weijiabu and the Xianyang hydrological stations from the upper to middle reaches (the data of 1965 and 1968-1971 264 265 are missing in the Weijiabu station). The parameters were calibrated for hydrological stations by 266 the order of upstream to midstream using the daily streamflow of 1960-1969. Firstly, the 267 parameters against the streamflow at the Linjiacun control station were calibrated. Secondly, 268 based on the premise of the calibrated parameter values of the Linjiacun station, the parameters 269 were calibrated for the subbasin controlled by the Weijiabu station. In that way, the parameters for 270 the subbasin controlled by the Xianyang station were then calibrated. Then the SWAT model was 271 validated for the three hydrological stations respectively against the streamflow from 1970 to 1979 272 (Fig. 7).

273

# 4. Results and discussions

274 The corresponding statistic results of three hydrological stations showed that the ranges of NS and R<sup>2</sup> were 0.59~0.66 and 0.63~0.68 respectively in the calibration period for a daily time 275 276 step. And they were  $0.57 \sim 0.62$  and  $0.61 \sim 0.65$  respectively in the validation period. At a monthly time step, the results of the NS and  $R^2$  were 0.82~0.84 and 0.79~0.86 respectively in the 277 278 calibration period. And they were 0.70~0.76 and 0.74~0.79 respectively in the validation period 279 demonstrating good performance of the model. In addition, the time-series and the patterns of the 280 simulated and observed streamflow during the calibration period and validation period showed 281 similar trends. Our conclusion is that the SWAT model can be used in upper and middle reaches of 282 the Wei River basin.

### 4.1 Impact of the observed LUCC on streamflow

In order to analyze the impact of the LUCC on streamflow, the land use data of the 1980 and 2005 were used in the validated SWAT model. Firstly, the daily streamflow from 1980 to 2009 were simulated using observed daily meteorological forcing data and topography, soil data in study area. Secondly, the LUCC data of 1980 was replaced by that of 2005 and their relevant parameters of corresponding land use type were also replaced. We used the LUCC data of 2005 but the same meteorological data to simulate the daily streamflow from 1980 to 2009.

290 The change of annual streamflow based on LUCC data of 2005 compared with LUCC data of 291 1980 showed that annual streamflow decreased during 20-year in 30-year ((1980-2009)) and the 292 annual average reduction was 2.0 mm/yr for these 20- year in study area. This is mainly because 293 over different land use types hydrological responds differently even to the same meteorological 294 forcings. For example, rainfall intensity was of great importance influencing to hydrological 295 process of the Wei River, which locates in semi-dry and semi-humid region (Lacombe et al., 2008; 296 Wang, 2014). Results of rainfall numerical experiments showed when the rainfall intensity was 297 smaller or larger, the rainfall would infiltrate into soil or flow away as surface runoff mainly on 298 both grass land and bare slope, while when the rainfall intensity was medium, the rainfall would 299 infiltrate into grass land and flowed away as surface runoff on bare slope (Tobella et al., 2014; 300 Wang, 2014). To reduce influence of meteorological conditions and isolate the impact of the 301 LUCC on streamflow, the 30-year (1980-2009) values of the streamflow for forest and agricultural 302 land were averaged respectively. For period of 1980-2009, we just used their measured and 303 long-term daily meteorological data in the study area to drive the validated model for the designed 304 hydrological experiments. Figure 8 shows the changes of streamflow, surface runoff, soil flow and

305 baseflow between agricultural land and forest. The surface runoff, soil flow and baseflow all 306 decreased for agricultural land, while the soil flow and baseflow of forest increased. Overall, the 307 streamflow decreased in agricultural land and increased in forest area. When the LUCC data are classified and re classified in SWAT model, the tree types are summarized as Range-Brush 308 309 (RNGB), Forest-Mixed (FRST) and Forest-Deciduous (FRSD). Different types have different 310 hydrological responses for their leaf, roots and so on. We also analyzed the streamflow generation 311 of the main types of forest (RNGB, FRST and FRSD) in study area further. Results showed that 312 the streamflow yield of FRST and FRSD were about 1.20 and 1.60 times of that of RNGB 313 respectively.

#### 314

# 4.2 Hydrological experiments on the impact of conversion of agricultural land to forests on streamflow 315

316 Because the LUCC data involves various land use interconversions, of particular interest here 317 the impact of conversion of cropland to forest on streamflow cannot be distinguished. Starting 318 from the LUCC data of 1980 as (S1) the present land use, we design other four scenarios (Table 3) 319 that (S2) 10%, (S3) 20%, (S4) 40% and (S5) 100% of the agricultural land was converted into 320 Forest-Mixed (FRST) respectively.

321 Based on the five scenarios, the SWAT simulations were conducted to analyze the effect of 322 forest constructions on the streamflow in upper and middle reaches of the Wei River basin. Firstly, 323 the converted agricultural land area was controlled proportionately as same as the variational area 324 ratios of set scenarios in 3 regions divided by Linjiacun, Weijiabu and Xianyang hydrological 325 stations (Fig. 6(a)). Secondly, lands with the same soil type and similar slope were the priorities 326 choosing as the converted land. Thirdly, the converted lands were distributed evenly as much as

327 possible in 3 regions. The simulation period was from 1980 to 2009.

328	We present the distribution of average streamflow change under $S2 \sim S5$ scenarios compared
329	with S1 scenario in Fig. 9. It shows that the streamflow generally increased when the land use
330	converted from agricultural land into forest in the upstream. And Fig. 10 shows the change rate of
331	streamflow at the Linjiacun, Weijiabu and Xianyang stations correspondingly for its annual
332	average and annual average over non-flood season (Jan - Jun and Nov - Dec). Compared with the
333	S1 scenario, the annual average streamflow increases in the non-flood season were 12.70 %,
334	11.21 % and 9.11% for the Linjiacun, Weijiabu and Xianyang stations with per 10% area of
335	agricultural land converted into forest. Interestingly the average annual streamflow increases were
336	11.61%, 21.63%, 42.51% and 109.25% for S2, S3, S4 and S5 scenario respectively (Fig. 10 (b)),
337	which almost consistently suggested about 1.1% per 1% change of the agricultural land. The
338	results are important in that one can expect that for a 0.8% increase in the forest in the observed
339	LUCC would lead to less than 1% change in the streamflow, which is negligible.
340	To be more comparable, Fig. 11 shows the distribution of the annual runoff coefficients with
341	the scenario changed from S1 to S5. The spatial variability in mean runoff coefficient was large,
342	which ranges from 0.03 to 0.68 and increased with more forest converted from agricultural land.
343	The annual average runoff coefficient of study area increased from 0.21 to 0.37 with forest area
344	increasing from S1 to S5 (Fig. 12). On average, the runoff coefficient increased about 0.014 (i.e.,
345	1.4% of rainfall transformed into streamflow) with per 10% area of agricultural land converted
346	into forest.

347 The landscape of the Wei River is mixed with the Loess Plateau and earth-rock mountain348 landscapes, which induce different mechanisms of transforming rainfall into streamflow. The

349 earth-rock mountain area accounts for 24.03% of study area (Fig. 4 (b)). In earth-rock mountain 350 area, vegetation grows on much thinner soil layer over the earth-rock mountain. And the soil has 351 high infiltration ability for high stone fragment content. The thin soil is apt to be saturated and 352 produce more soil flow on relatively impermeable rock, hence the streamflow in wooded areas is 353 larger than that in adjacent woodless areas favoring streamflow production (Liu and Zhong, 1978). 354 On the contrary, in Loess Plateau there is exiting a drying layer of soil underneath forestland in great water deficit. When the agricultural land converted into forest, the precipitation, intercepted 355 356 by vegetation, infiltrated into soil and supplied the drying layer of soil, vegetation growth, etc. 357 Together with much thicker soil layer on the Loess Plateau, it usually prevents gravitational infiltration into groundwater and reduces streamflow recharge (Li, 2001; Tian, 2010). The 358 359 observed results of precipitation and streamflow in study area also showed the runoff coefficients 360 had obviously positive correlation with rates of earth-rock mountain area. The regional annual 361 averages of runoff coefficient were 0.13, 0.17 and 0.35 for Fig. 6 (b), (d) and (c), while the rates of 362 earth-rock mountain area were opposite correspondingly (Fig. 4 (b)). The complication is that the 363 overall effect of forest on the streamflow is in fact a balance between earth-rock mountain positive 364 and Loess Plateau negative effects on the streamflow.

Combined with the spatial distribution of precipitation (Fig. 5 (a)), we can see earth-rock mountain landscapes are mainly distributed in regions with more rainfall. To be precise, the whole earth-rock mountain area located where rainfall was greater than 500 mm/yr and over 62% of the study area where the annual rainfall is greater than 600 mm was in earth-rock mountain. Meanwhile, the river network over the earth-rock mountain is denser and most of tributaries in the earth-rock mountain are close to the main stream of the Wei River. Moreover, there distribute a lot of developed gravel riverbed in piedmont, sandy soil along the river and its groundwater level is shallow, which facilitate rainfall infiltration and recharging streamflow. Therefore although the area of earth-rock mountain accounts for 24% of the study area, its distribution areas are concentrated in the main regions of streamflow yield of the study area. Therefore the overall result of balance among all factors was that the forest constructions have positive effect on streamflow.

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

In Fig. 10 (a), one important point is that the average increase in the non-flood season was 377 378 about 1.41 times larger than the annual increase of the streamflow. To understand that, Fig. 13 379 shows distribution of the baseflow index, i.e., the ratio between baseflow and streamflow, under S1~S5 scenarios. We can see that the baseflow index also increased with land use converted from 380 381 agricultural land into forest, which means that groundwater contribution to the streamflow 382 increased with the overall increase of forest area. Putting the pictures together, Fig. 14 shows the 383 changes of the streamflow and the baseflow under the S2~S5 scenarios minus those results under 384 the S1 scenario in the non-flood season. The average increases of streamflow and baseflow were 385 1.14 and 0.98 mm/yr with per 1% increase of forest area respectively. For the non-flood season, 386 they were 0.60 and 0.53 mm/yr. The increase of the streamflow contributed by the increased 387 baseflow was about 88.33% in the non-flood season. So the increasing streamflow was mainly 388 contributed by groundwater with increasing of forest area overall.

**5.** Conclusion

390 The large scalar implementation of Grain for Green project in China is expected to alter 391 hydrological cycle, in particular on the Loess Plateau, within the Yellow River Basin. The 392 scientific question is how large the impact of the LUCC on the streamflow and its components in that area. We choose the Wei River as the study area, in that it has been widely implemented revegetation constructions since the 1980s. Of particular interest here, the landscape of the upper and middle reaches of the Wei River basin is mixed with the Loess Plateau and rocky mountain, which would induce different mechanisms of generating surface runoff, soil flow, base flow and therefore streamflow.

398 To investigate it, we setup the SWAT model for the upper and middle reaches of the Wei 399 River basin with the inputs of long term observed meteorological forcing data, hydrological data, 400 and observed land use data. We use daily and monthly streamflow of the Linjiacun, Weijiabu and 401 Xianyang hydrological stations from upper to middle reaches during 1960-1969 and 1970-1979 402 respectively for the model calibration and model validation. The results showed that the Nash-Sutcliffe (NS) coefficients and the coefficients of determination ( $\mathbb{R}^2$ ) were > 0.57 and 0.61 403 404 for daily streamflow and 0.70 and 0.74 for monthly streamflow respectively demonstrating that 405 the SWAT model can be used in this study.

406 We analyse the impact of the LUCC on streamflow based on the observed LUCC data of 407 1980 and 2005. The daily streamflow from 1980 to 2009 were simulated using observed daily 408 meteorological data with the two different land use data. The results showed that two-thirds of 409 annual streamflow decreased and the change of streamflow was different among different land use. 410 On the overall average, the 30-year averages of the streamflow decreased in agricultural land but 411 increased in forest. To interpret the overall result, we design five scenarios in this study including 412 (S1) the present land use of 1980 and the scenarios where agricultural land was converted into 413 forest by 10% (S2), 20% (S3), 40% (S4) and 100% (S5) respectively. Based on the five scenarios, 414 we use the calibrated and validated SWAT model to analyze the effect of forest constructions on

415 the streamflow in detail. The results confirm that annual streamflow consistently increased with 416 more forest converted from the agricultural land. Interestingly, the rate is almost consistently 7.41 417 mm/yr per 10% increase of forest converted from the agricultural land. Based on detailed analysis 418 of each component of streamflow, we found it was most attributed by the baseflow. The overall 419 effect of LUCC on the streamflow in the Wei River basin, the largest branch of the Yellow River is 420 the result of the balance between Loess Plateau negative and earth-rock mountain positive effects. 421 Our results here are not only of great importance in understanding the impact of LUCC on 422 streamflow for a catchment with much complicated and mixed landscape, but also of significance 423 for water resources managing practice.

# 424 **Data availability**

425 The data used in this manuscript were obtained from reliable public data repositories. The 426 LUCC and soil data were obtained from the National Science & Technology Infrastructure of 427 China, the National Earth System Science Data Sharing Infrastructure (http://www.geodata.cn). 428 The DEM data were obtained from the Computer Network Information Center, the Chinese 429 Academy of Sciences (http://srtm.datamirror.csdb.cn/). The meteorological data were obtained 430 from the China Meteorological Data Sharing Service System 431 (http://www.escience.gov.cn/metdata/page/index.html). The daily streamflow data were from the 432 Ecological Environment Database of Loess Plateau (http://www.loess.csdb.cn/pdmp/index.action) and the Hydrological Year books of China. 433

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# 567 **Figure Captions:**

- 568 Fig. 1 The development of soil and water conservation measures in the main stream basin of Wei River
- over last 50 years.
- 570 Fig. 2 The study area: the Wei river basin on the Loess Plateau.
- 571 Fig. 3 The observed land use data of the year 1980 and the year 2005 in study area.
- 572 Fig. 4 The Soil data and the distribution of earth-rock mountain in study area.
- 573 Fig. 5 The spatial distribution of annual average precipitation in Wei River basin over the past 55 years
- 574 (1956-2010) and the DEM of study area.
- 575 Fig. 6 The time-series of precipitation, annual streamflow and runoff coefficients for the regions of
- 576 study area.
- 577 Fig.7 The time-series graphs of calculated vs. observed values during calibration period and
- 578 verification period for hydrological stations.
- 579 Fig. 8 The changes of 30-year (1980-2009) averages of streamflow, surface runoff, soil flow and
- 580 baseflow between agricultural land and forest.
- 581 Fig. 9 The watershed distribution of average streamflow change under S2~S5 scenarios compared with
- 582 S1 scenario.
- 583 Fig. 10 The corresponding proportional change rate of streamflow at Linjiacun, Weijiabu and Xianyang
- 584 station for annual average and annual average in non-flood season.
- 585 Fig. 11 The distribution of annual runoff coefficient with the scenario changed from S1 to S5.
- 586 Fig. 12 The annual average runoff coefficient of study area with forest area increasing from S1 to S5.

- 587 Fig. 13 The distribution of baseflow index under S1~S5 scenarios.
- 588 Fig. 14 The corresponding change of streamflow and baseflow under S2~S5 scenarios compared with
- 589 S1 for annual average of year and non-flood season.





592 Fig. 1 The development of soil and water conservation measures in the main stream basin of Wei River



#### over last 50 years

594 Figure 1 (a) is the area developing of forestation, terraces, grass and dam land separately. Figure

595 1(b) is the sum area of the forestation, terraces, grass and dam land in upstream, midstream

### 596 and downstream.







Fig. 3 The observed land use data of the year 1980 and the year 2005 in study area





Fig. 4 The Soil data and the distribution of earth-rock mountain in study area



604 Fig. 5 The spatial distribution of annual average precipitation in Wei River basin over the past 55 years

605 (1956-2010) and the DEM study area





study area





610 Fig. 7 The time-series graphs of calculated vs. observed values during calibration period and verification



613 Fig. 8 The changes of 30-year (1980-2009) averages of streamflow, surface runoff, soil flow and

612

baseflow between agricultural land and forest





616 Fig. 9 The watershed distribution of average streamflow change under S2~S5 scenarios compared with

617 S1 scenario



**Fig.10** The corresponding proportional change rate of streamflow at Linjiacun, Weijiabu and Xianyang

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





Fig. 11 The distribution of annual runoff coefficient with the scenario changed from S1 to S5



Fig. 12 The annual average runoff coefficient of study area with forest area increasing from S1 to S5 624









Fig. 13 The distribution of baseflow index under S1~S5 scenarios



629 Fig.14 The corresponding change of streamflow and baseflow under S2~S5 scenarios compared with

<b>S</b> 1	for annual	average of	vear and	non-flood	season
			J		

# 631 Tables

#### 632

628

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

			1	1
No.	Code of Soil type	Physical meaning of the code	HRU	Area
				(km <sup>2</sup> )
1	SHYZHT	Limestone Cinnamon soil	220, 257	26316.90
2	SHYZSHXHT	Limestone Calcic cinnamon soil	153	11471.22
3	SYYZLRHT	Sandstone—shale Luvie cinnamon soil	166, 203, 207	50065.29
4	HGPMYZLRHT	Granite—gneiss Luvie 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						
Table 2 Calibrated values of model parameters							
			Calibration result				
Parameters	Physical meaning	Calibration	Linjiac	Weijia	Xianya		
		Tange	un	bu	ng		
r_CN2	Initial SCS runoff curve number for	0202	0.27	0.05	0.17		
	moisture condition II	-0.5~0.5	-0.27	0.05	-0.17		
rSOL_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	05.05	0.5	0.2	0.5		
	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		
vALPHA_BF	Baseflow alpha factor	0~1.0	0.48	0.61	0.61		
vGW_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		
Notes: vmeans the existing parameter value is to be replaced by the given value; rmeans the existing parameter value is multiplied							
Iable 3 Scenarios for simulation       The average simulated streamf							
Scenario	Description	Area (km <sup>2</sup> )	(1980-2009) $(10^8 \text{ m}^3/\text{yr})$				
S 1	present situation	0		50.44			

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

639 Notes: ① Agricultural land refers to the land for crops planting, including cultivated land, newly cultivated soil, fallow field,

640 rotation plot, pasture-crop rotation and land used for agro-fruit, agro-mulberry, agroforestry (The code in model is AGRL). ②

641 Forest refers to the natural forest and plantation, which canopy density is larger than 30%, including timberland, economic forest,

642 protection forest (The code in model is FRST).