

Research of Assessment Model Framework of Rainwater Resource Utilization and Driving Force in Arid and Semi-Arid Areas

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1 **Research of assessment model framework of rainwater resource utilization and**
2 **driving force in arid and semi-arid areas**

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7 **Abstract:** The arid and semi-arid areas generally face the degradation of natural
8 resources, poor vegetation conditions, and serious soil erosion. In this paper, a spatio-
9 temporal assessment model for the rainwater resources utilization and its driving forces
10 is studied and constructed, and a typical watershed is selected for application and
11 verification. The model is based on SWAT (Soil & Water Assessment Tool) and GTWR
12 (Geographically Temporal Weighted Regression) for secondary development. Two
13 indicators are proposed to characterize the utilization of rainwater: rainwater resources
14 utilization quantity (RRU) and rainwater resources utilization rate (RRUR). To further
15 verify the performance of the model, the scenarios of different watershed management
16 measures were simulated and evaluated. The results show that the model can accurately
17 evaluate the RRU, RRUR, and driving force. The annual average value of the RRUR
18 in the typical watershed is 0.44, and the utilization rate of the middle reaches is higher
19 than that of the upper and lower reaches. Meanwhile, the model can evaluate the RRUR
20 and driving force of different land management scenarios. The driving effects of
21 meteorological, topographical, vegetation, and other factors on the regional RRUR
22 exhibits certain spatiotemporal differences, among which the leaf area index (LAI) is
23 the key driving factor. The framework can provide scientific guidance for improving
24 the rainwater resources utilization rate (RRUR) and alleviating ecological problems.

25 **Key words:** Rainwater resource utilization; Assessment model; Spatio-temporal

26 driving force; Watershed management

27 **1 Introduction**

28 The degradation of natural resources is one of the main problems faced by arid
29 and semi-arid areas (Liu et al. 2005). For a long period of time, improving the utilization
30 ratio of rainwater has been an important method by which to alleviate the problems of
31 soil drought and ecological degradation (Hatibu 2002), yet achieving this remains a
32 challenge. Rainwater in the arid and semi-arid areas does not penetrate deeply into the
33 soil profile, and most of it evaporates directly into the atmosphere after rainfall and lose
34 (Zhang et al. 2013). Rainwater utilization (or rainwater harvesting, RWH) in watershed
35 can reduce slope runoff, alleviates soil erosion, increases groundwater recharge (Page
36 et al. 2016), improves ecosystem function (Aladenola and Adeboye 2010).

37 Rainwater harvesting and management (RWHM) technology has great potential to
38 improve rainwater utilization efficiency and maintain rain fed agriculture in the region.
39 For agricultural areas, rainwater utilization is more reflected in the underlying surface
40 of the basin in addition to the form of point harvesting. Rainwater is used for
41 agricultural production, vegetation growth and wetland improvement in these areas
42 (Oweis 2017; Singh 2012). Many scholars are exploring the best management practices
43 to achieve better utilization of rainwater resources.

44 Rainwater utilization involves many aspects, and it is necessary to fully consider
45 a series of social and natural conditions and cooperate with multiple stakeholders (Gain
46 et al. 2017). There is inadequate knowledge and scientific assessment on hydrological
47 impacts and limits of up-scaling rainwater harvesting at a river basin scale (Ngigi 2003;
48 Glendenning et al. 2012). Ngigi (2003) assessed the potential of RWH in improving
49 food and water supplies. At the same time, there is a knowledge gap in the limitation of
50 watershed expansion of RWH, especially in the semi-arid areas of East Africa, what is

51 the limit of expanding rainwater utilization?

52 Considering the difficulty, cost and time required for physical measurements of
53 hydrological elements, modelling often provides a cheaper, faster way to consider larger
54 scale watershed hydrological impacts of RWH (Glendenning et al. 2012). Kadam et al.
55 (2012) used SCS-CN method to identify potential rainwater harvesting sites in the semi-
56 arid basaltic region of Western India with considering the hydrological, geomorphic,
57 and geological parameter. Based on a physical, spatially distributed TRAIN-ZIN model,
58 Shadeed and Lange (2010) calculated evapotranspiration to assess the rainwater
59 harvesting potential and two different rainwater harvesting technologies in the arid and
60 semi-arid Faria catchment in the West Bank of Palestine. Ghimire and Johnston (2013)
61 used the SWAT model to simulate the baseline of urban and agricultural land use and
62 the RWH scenarios to better understand the impact of urban domestic water and
63 agricultural rainwater utilization on regional hydrology, and applied the method to three
64 different watersheds in the southeastern United States. Ngigi et al. (2007) presented
65 hydrological assessment of up-scaling rainwater harvesting (HASR) conceptual
66 framework, which assesses the impacts of land use changes on hydrological regime in
67 the upper Ewaso Ng'iro river basin in Kenya. Mahmoud et al. (2016) evaluated the
68 suitability of RWH in the study area and selected five factors: soil type, land cover and
69 land slope (topography), runoff coefficient, and remaining precipitation to determine
70 the suitable area for RWH. This is a multi-objective, multi-standard problem. Terêncio
71 et al. (2017) constructed a RWH suitability assessment model based on physical, socio-
72 economic, and ecological variables to guide the implementation of rainwater utilization
73 agricultural and forestry management measures in the catchment area.

74 Scholars have used various methods to study the potential and hydrological impact
75 of rainwater utilization in the basin. In arid and semi-arid areas where land management

76 measures are commonly implemented, it is still complicated and cumbersome how to
77 finely evaluate the rainwater utilization of different measures and clarify the main
78 influencing factors of regional rainwater utilization. RHW will change the water
79 balance in the basin, and the hydrological cycle of agricultural basins will be
80 continuously affected by natural factors such as climate change, topography, and
81 vegetation in arid and semi-arid areas. These hydrological response includes changes
82 in runoff, groundwater supply, evaporation, etc. (Beganskas et al. 2019; Vanlooche et
83 al. 2010), and exhibits obvious characteristics of spatial heterogeneity and time non-
84 stationarity (Beven 1989). For these reasons, it is also important to guide the utilization
85 of rainwater in the watershed to explore the natural influencing factors of rainwater
86 utilization on the spatial and temporal scales in addition to the assessment of rainwater
87 resources.

88 Therefore, we aim to propose a flexible model framework for rainwater utilization
89 and driving force assessment on a temporal and spatial scale. This paper constructs the
90 assessment model for the rainwater resources utilization and driving forces, takes a
91 typical watershed in arid and semi-arid areas as an example to verify the feasibility of
92 the model, and use the model evaluate the scenarios of different watershed management
93 measures to verify the wider applicability. The specific goals are as follows: i) Propose
94 evaluation methods and indicators for rainwater resources utilization in arid and semi-
95 arid areas; ii) Construct rainwater resources utilization evaluation model based on
96 SWAT model and GTWR model; iii) Verify the applicability of the model under
97 different watershed management scenarios, and study the impact of different
98 management practices on changes in hydrological elements in agricultural basins. This
99 model framework may have more practical significance in arid and semi-arid areas
100 where management measures such as large-scale terraced construction and

101 afforestation are more common. The research can provide scientific tools for the
102 evaluation and utilization of rainwater resources in these areas.

103 **2 Methodology**

104 **2.1 Assessment model of rainwater resources utilization**

105 To assess the utilization of rainwater, it is necessary firstly to clarify the utilization
106 and non-utilization components of rainwater resources. In arid and semi-arid areas, the
107 discrete rainfall events are important components of water resources and aquifers
108 developed in closed basins where lateral inflow is nearly absent on the long time scale
109 (Miranda et al. 2011; Van Camp et al. 2010). At the same time, evapotranspiration (E),
110 includes both productive water use, i.e. transpiration (E_t), and non-productive water use,
111 i.e. evaporation from soil surface (E_s) and evaporation from canopy interception (E_i)
112 (Cheng et al. 2017). This paper holds that the rainfall which may be used to improve
113 the ecological environment in the basin space is considered utilization of rainwater, and
114 the evaporation and runoff from the basin which are not involved in the vegetation
115 growth process are considered non-utilization of rainwater. Therefore, we propose two
116 indexes of rainwater utilization evaluation, namely, rainwater resources utilization
117 quantity (RRU) and rainwater resources utilization rate (RRUR), and the calculation
118 formula of the two indexes is as follows:

$$119 \quad RRU_t = P_t - E_{s,t} - E_{i,t} - Q_t - \Delta W_t \quad (1)$$

$$120 \quad RRUR = \frac{RRU_t}{P_t} \quad (2)$$

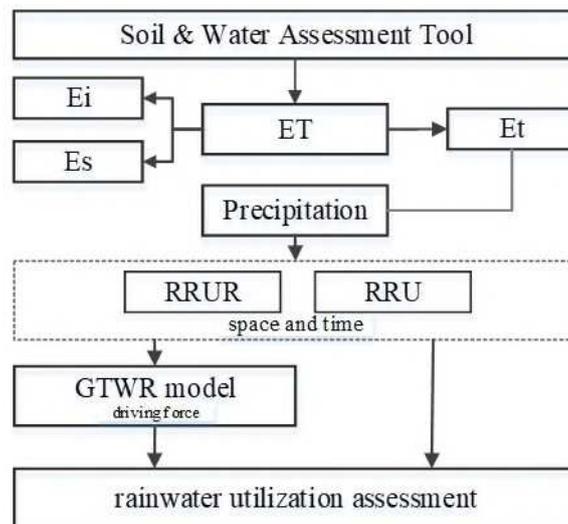
121 where t is the time step; P_t is the precipitation, mm; $E_{s,t}$ is the soil evaporation, mm;

122 $E_{i,t}$ is the vegetation intercepted evaporation, mm; Q_t is the basin discharge, mm;

123 ΔW_t is the precipitation loss such as water surface evaporation, and when there are

124 water collection facilities such as reservoirs in the basin, ΔW_r includes the river water
125 surface evaporation and reservoir evaporation, mm.

126 This paper constructs an assessment model of rainwater utilization and driving
127 force based on SWAT model and GTWR model. The SWAT model is used to calculate
128 hydrological elements such as evaporation. Then, RRUR and RRU can be calculated
129 by combining precipitation, and GTWR model is used to analyze driving factors. The
130 technical route of the assessment model is shown in Fig. 1. The rainfall runoff and
131 evaporation process in the basin have a certain lag. Therefore, in order to avoid errors
132 on a short time scale, the model is best to evaluate the rainwater utilization on a monthly
133 or annual scale. The following studies confirm the accuracy and reliability of the model
134 on an annual scale.



135

136 Fig.1 Technical route of rainwater utilization assessment model

137 2.2 Soil & Water Assessment Tool (SWAT)

138 As a basin modeling tool, SWAT has been used by many scholars to quantify
139 ecosystem services (Francesconi et al. 2016). In the SWAT model, the hydrological
140 process of the basin includes the land part of water cycle - runoff generation and slope
141 confluence, and the water part of water cycle - river confluence. The sub-basin module

142 controls the input of water, sediment, nutrient load, and chemical substances in the main
143 river channel of each sub-basin. SWAT can divide several sub-basins according to the
144 characteristics of the basin water system. The physical parameters of the sub-basin are
145 different, to represent the spatial difference of runoff yield and concentration of the
146 basin (Neitsch et al. 2011).

147 The E_t , E_i and E_s cannot be calculated directly by the SWAT, only E can be
148 calculated. After obtaining E , the components need to be calculated. Among them, E_i
149 is calculated according to canopy leaf area index (LAI) during rainfall. Partitioning of
150 transpiration and soil evaporation can be evaluated via Beer's Law as:

$$151 \quad \frac{E_t}{E_t + E_s} = 1 - \exp(-kL) \quad (3)$$

152 where L is the canopy leaf area index, and k is the radiation extinction coefficient.
153 Partitioning of transpiration and soil evaporation is much more complex than Beer's
154 law in reality (Jarvis and McNaughton 1986; Wang et al. 2014), especially at short
155 times scales. However, this study is focused on monthly or annual time scales, at which
156 Beer's law can provide reasonable and accurate partitioning between transpiration and
157 soil evaporation (Kelliher et al. 1995; Wang et al. 2014).

158 **2.3 Geographically Temporal Weighted Regression (GTWR) model**

159 In order to consider the characteristics of rainwater utilization in time and space,
160 and to comprehensively analyze the driving effects of weather, terrain, vegetation
161 conditions and other factors on the RRUR, the Geographically Temporal Weighted
162 Regression (GTWR) model is added to the assessment model. The GTWR model splits
163 the spatial position coordinates and time coordinates into three coordinates, and
164 considers the influence of space and time on the regression coefficient of each
165 explanatory variable (Huang et al. 2010; Wu et al. 2019). In the time coordinate, the

166 coordinate of spatiotemporal position i is (u_i, v_i, t_i) , and the GTWR model can be
167 expressed as follows (Fotheringham et al. 2015; Huang et al. 2010):

$$168 \quad Y_i = \alpha_0(u_i, v_i, t_i) + \sum_{j=1}^m \alpha_j(u_i, v_i, t_i) X_{ij} + \varepsilon_i \quad (4)$$

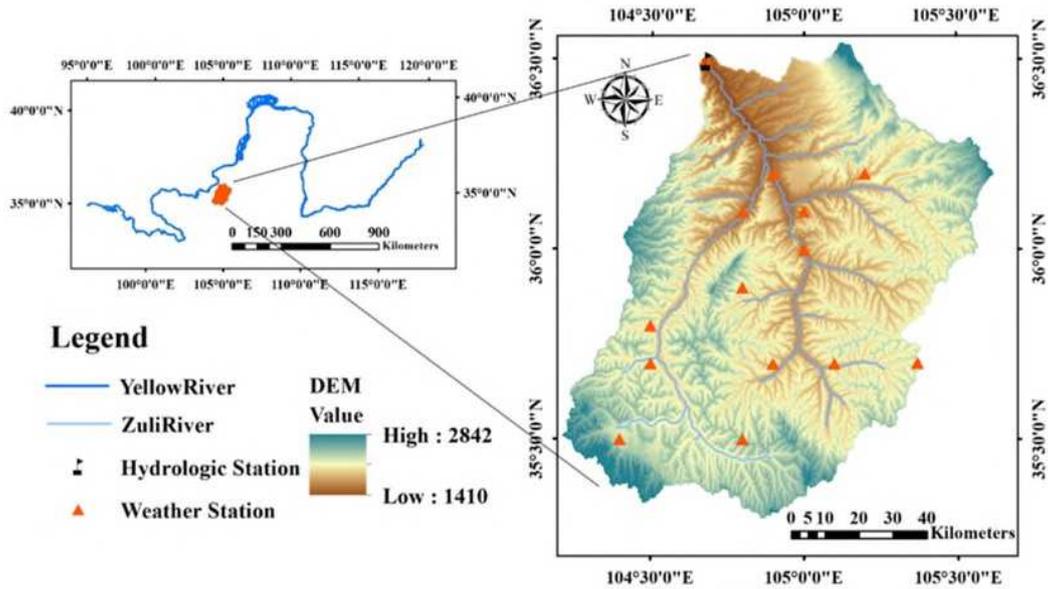
169 where Y_i is the interpreted variable value of sample point i ($i=1,2,\dots,n$); n is the number
170 of sample points; m is the number of explanatory variables; t_i is the time coordinate at
171 sample point i ; $\alpha_0(u_i, v_i, t_i)$ is the time-space intercept term of sample point i ; X_{ij} is
172 the j^{th} explanatory variable value of sample point i ; and $\alpha_j(u_i, v_i, t_i)$ is the regression
173 coefficient of the j^{th} variable at sample point i .

174 **3 Research Area and Model Application**

175 **3.1 Research Area and Data**

176 In this paper, the Zuli River Basin is selected as the research object. The Zuli River
177 is a first-class tributary of the upper reaches of the Yellow River in China, and it is
178 located in the western part of the Loess Plateau. It is a typical area for the development
179 of agriculture and animal husbandry in arid and semi-arid areas of China (Li 2018),
180 with a drainage area of 10,584 km² and an average annual runoff of 145 million m³.
181 The rainfall in the upper reaches is high. In the middle reaches, rainfall began to
182 decrease and the salinity of the river water began to increase. The annual precipitation
183 is about 300 mm, and the salinity is more than 10 g/L, making the runoff there difficult
184 to use in the lower reaches. Jingyuan Hydrological Station is the control station at the
185 outlet of the basin. See Fig. 2 for a topographic map of the basin.

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Fig. 2 Overview of the Zuli River Basin

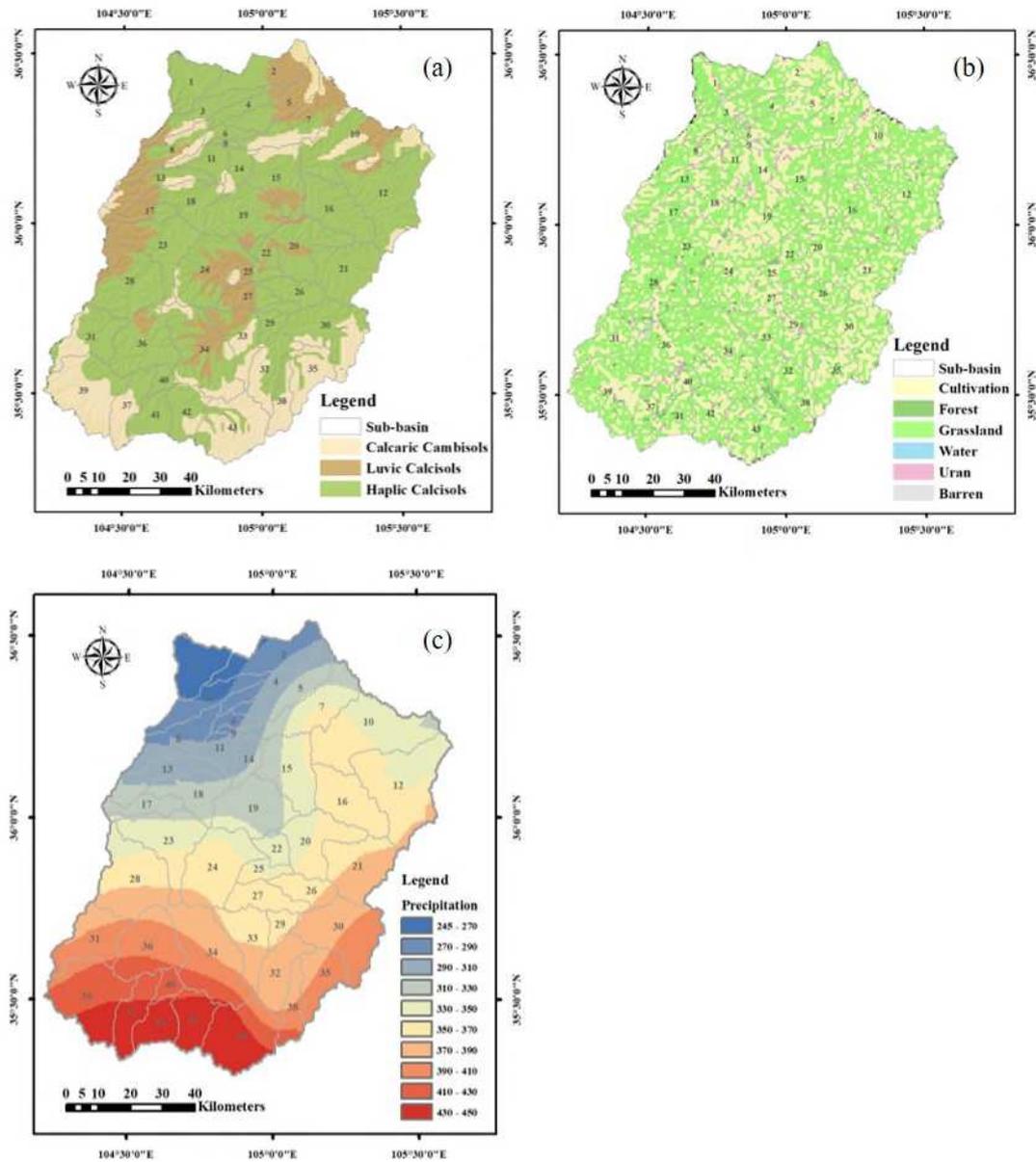
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In the Zuli River Basin, the loess deposits are thick, and the main soil types are calcium soil. The main natural vegetation is temperate semi-arid grasslands and arid grasslands, and forests are scarce. The overall vegetation condition in the upper reaches is superior to that in the middle and lower reaches, and it is the worst near the estuary area.

193

The hydrological and meteorological data that support the findings of this study are from the Yellow River Committee. Evaporation data of 1981-1990 is available from CSIRO at the following url: <https://www.csiro.au/en/Research/LWF>. Geographic data comes from Geospatial Data Cloud (<http://www.gscloud.cn/home>) and Resource and Environment Science and Data Center (<https://www.resdc.cn/Default.aspx>).

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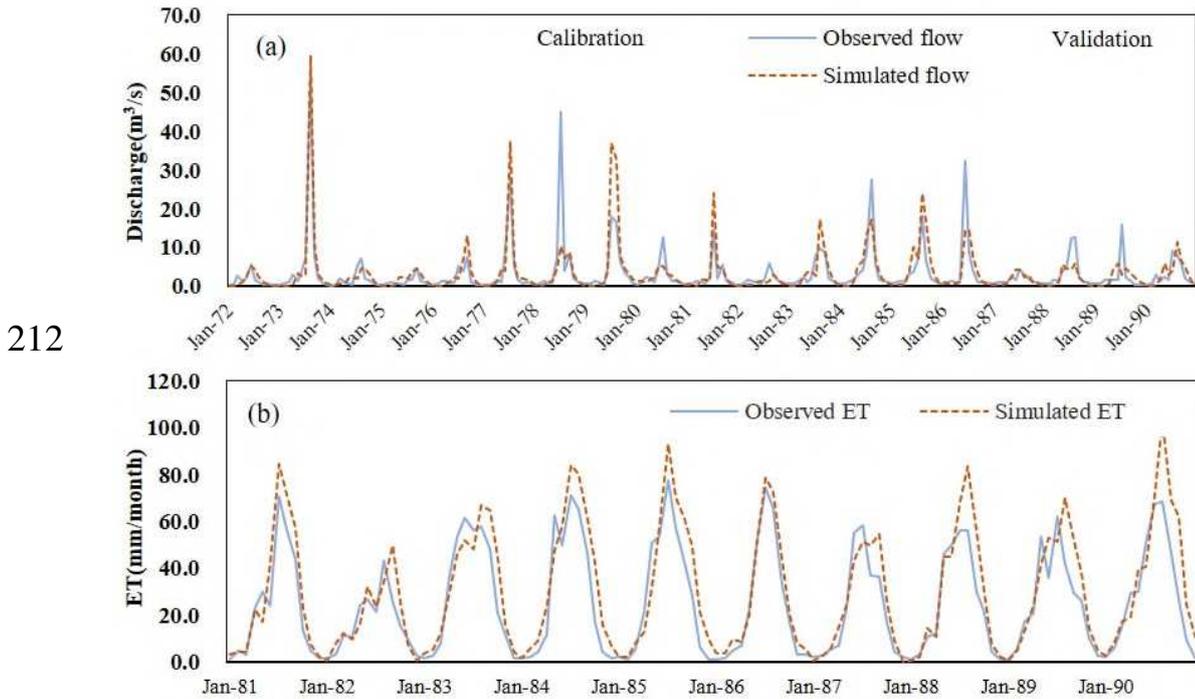
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199 Fig. 3 Distribution of soil types(a), land use(b), precipitation(c) and sub-basin division
 200 in the research area

201 **3.2 Construction of the Assessment Model**

202 In this paper, the period of 1972-1990 (1972-1980: calibration period; 1981-1990:
 203 verification period) is selected as the research period, and the land cover data of 1980
 204 are used as the benchmark. We constructed a complete assessment model of rainwater
 205 utilization in Zuli River Basin, and analyzed the characteristics and applicability of the
 206 model framework. First, based on the characteristics of topographic confluence, the

207 research area is divided into 43 sub-basins (Fig. 3), and 1,297 hydraulic response units
208 (HRUs) are subdivided according to soil type, land use, and topographic slope. There
209 are 14 single rainfall stations in the basin, and rainfall daily data comes from these
210 stations (see Fig. 2 for the distribution of rainfall stations). Other weather data come
211 from four comprehensive weather stations adjacent the basin.



213 Fig. 4 Calibration and validation of runoff results(a) and validation of evaporation(b)

214 The runoff results of the assessment model are calibrated and verified by the
215 monthly measured runoff series of Jingyuan Hydrological Station for the periods of
216 1972-1990 (Fig. 4), and the evaporation is verified by the PML estimates 1981-1990
217 data (Zhang et al. 2016) (Fig. 4). NSE, PBIAS and R^2 are selected as the evaluation
218 indexes (Table 1), and the calibrated parameters of the model are shown in Table 2.

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220
221
222

Table 1 Model calibration and validation evaluation index value

	Discharge Calibration (1972-1980)	Discharge Validation (1981-1990)	ET Validation (1981-1990)
NSE	0.64	0.56	0.75
PBIAS	-6.62%	-2.78%	20.27%
R ²	0.69	0.56	0.85

224

Notes: Due to some data limitations, the model is difficult to achieve better flow

225

calibration and verification results. However, the model can well reflect the trend of the

226

hydrological cycle and does not affect the further study of the model.

227

Table 2 Model calibrated parameters

ID	Parameter	Range	Value
1	v__ALPHA_BF.gw	0-1	0.148333
2	v__ALPHA_BNK.rte	0-1	0.304333
3	v__CH_K2.rte	0.01-500	3.509930
4	v__CH_N2.rte	0.01-0.3	0.077183
5	r__CN2.mgt	-0.5-0.5	-0.224333
6	v__ESCO.hru	0-1	0.805667
7	v__GW_DELAY.gw	0-500	165.833344
8	v__GW_REVAP.gw	0.02-0.2	0.078380
9	v__OV_N.hru	0.01-30	8.077310
10	r__SOL_AWC().sol	-0.9-0.9	0.159000
11	r__SOL_BD().sol	-0.9-0.9	-0.743400
12	r__SOL_K().sol	-0.9-0.9	0.103800
13	v__SURLAG.bsn	0.05-24	8.983350
14	v__CANMX.hru	3-15	10.388000

229 Then, the dependent variable and explanatory variables are selected to construct
 230 GTWR model. The RRU is directly related to rainfall, so it is difficult to reflect the
 231 deep driving force of rainwater resources utilization by analyzing its influencing factors.
 232 The RRUR is the utilization rate (the ratio of RRU to rainfall), which can better reflect
 233 the impact of temporal and spatial factors on the utilization of rainwater as a dependent
 234 variable. The selected explanatory variables are: temperature (T_{mp}), precipitation
 235 (Precip), area (Area), slope (Slope), elevation (Elev), soil infiltration coefficient (Sol),
 236 potential evaporation (PET), and leaf area index (LAI), and the explanatory variables
 237 were normalized. The selection of explanatory variables is very flexible and can be
 238 determined based on analysis of the actual conditions of the river basin, which is also
 239 important for studying the temporal and spatial effects of rainwater utilization. The 43
 240 sub-basins were selected as spatial samples, and the time scale was 18 years from 1972
 241 to 1990, so there are 817 spatiotemporal samples. The RRUR is the dependent variable,
 242 and the year and spatial position are taken as the time and spatial coordinates to form
 243 the three-dimensional coordinates of the sample points. In this way, the model is
 244 completed (Equation 5), which can analyze the driving effects of different factors on
 245 RRUR and explore its temporal and spatial variation law.

$$246 \quad R_{STi} = \alpha_{0i} + \alpha_{TMi} TM_i + \alpha_{PRi} PR_i + \alpha_{ARi} AR_i + \alpha_{SLi} SL_i + \alpha_{EIi} EI_i + \quad (5)$$

$$\alpha_{SOi} SO_i + \alpha_{PEi} PE_i + \alpha_{LAIi} LA_i + \varepsilon_i$$

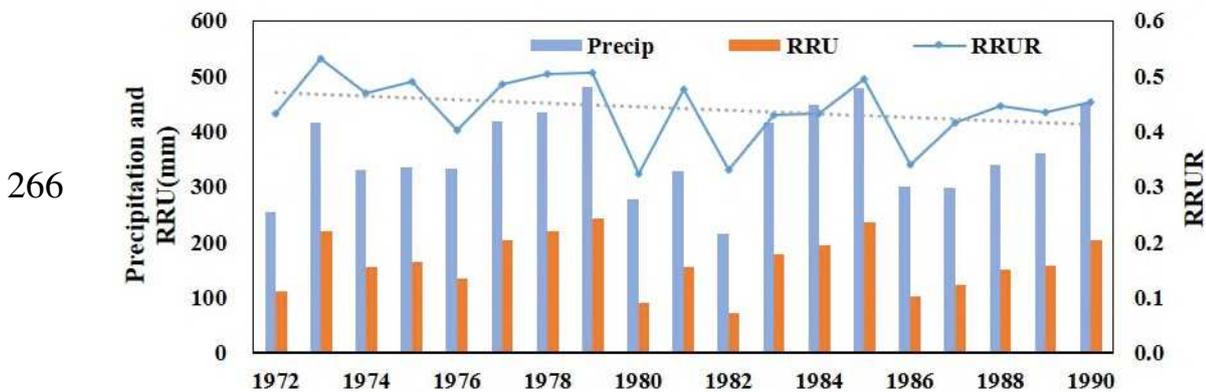
247 where R_{STi} is the RRUR of the i^{th} sample, and all regression coefficient α changes
 248 with time and space. The weight function is the Gaussian function. The space-time
 249 distance ratio of the model is determined by automatic optimization, while the optimal
 250 bandwidth is determined by the AICc criterion.

251 **4 Results and Discussion**

252 **4.1 Temporal and spatial situation of rainwater utilization**

253 Through the calculation of the assessment model, the RRU and RRUR can be
254 obtained, and the rainwater utilization changes in the basin can be analyzed. The spatial
255 distribution of RRU and RRUR can also be obtained to further clarify the spatial
256 difference of rainwater utilization.

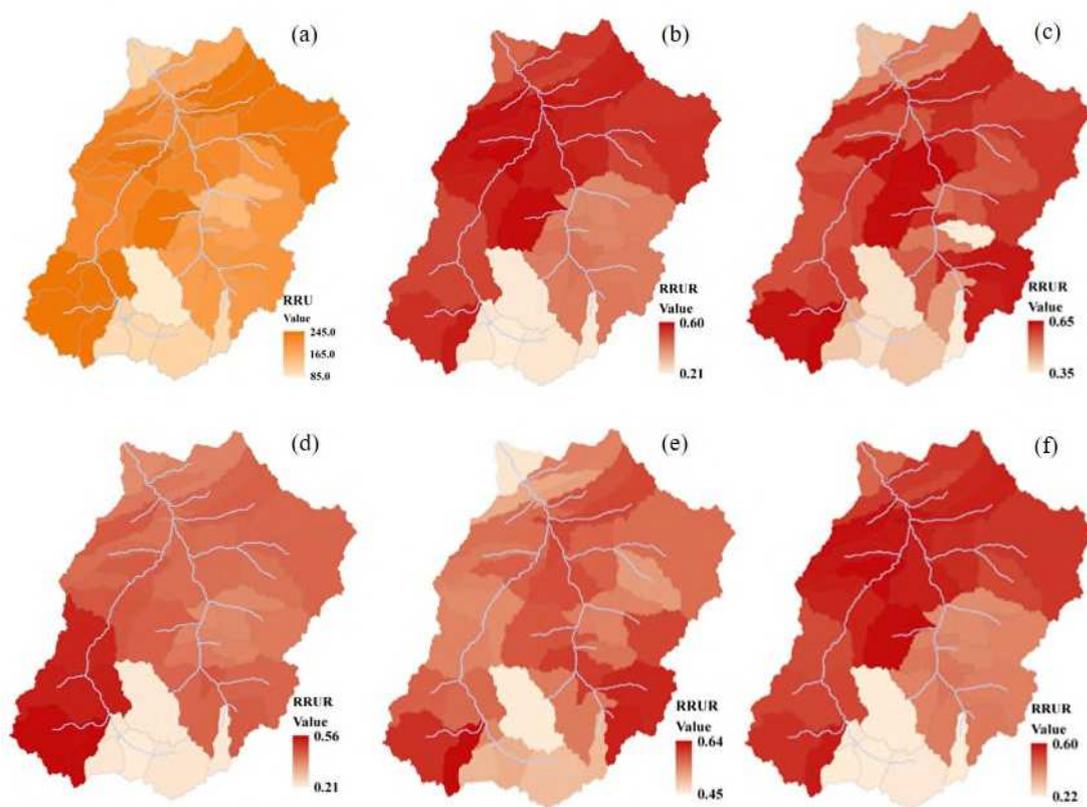
257 The results show that the average value of the RRU is 164 mm in study period,
258 and the increase of precipitation will lead to the increase of RRU. The RRU in the years
259 with higher precipitation may exceed 200 mm. For the RRUR, the annual average value
260 is 0.44. The higher the precipitation is, the larger the RRUR will be. The highest RRUR
261 was 0.53 and the lowest was 0.32, and it is within the range of 0.4-0.5 in most years. In
262 the overall trend, the RRUR of the Zuli River Basin shows a decreasing trend during
263 the study period. Greater amounts of rainwater resources are lost, thereby leading to the
264 aggravation of the basin drought, the deterioration of the ecological environment, and
265 the further reduction of the water storage and water conservation capacity of the basin.



267 Fig. 5 Current rainwater utilization situation in the Zuli River Basin

268 At the same time, the average RRU in the southwest and middle reaches of the
269 basin is relatively high, the spatial distribution of RRUR from assessment model show
270 that the annual average RRUR in the western area of the upper and middle reaches of
271 the basin are relatively high, while those in the eastern areas of the middle and upper
272 reaches are relatively low (Fig. 6). In a concentrated area in the upper reaches of the

273 basin, more than 70% of rainwater resources have been lost. The spatial difference of
274 the RRUR is related to the spatial distribution of land use types. The RRUR is relatively
275 high in the areas with greater amounts of cultivated land and less grassland. The reason
276 for this phenomenon is that the Zuli River Basin is mainly cultivated land and grassland,
277 among which the grassland vegetation condition is poor. For the research basin, the
278 upstream rainfall is relatively large, and the RRUR is small, which theoretically has
279 greater rainwater utilization potential.

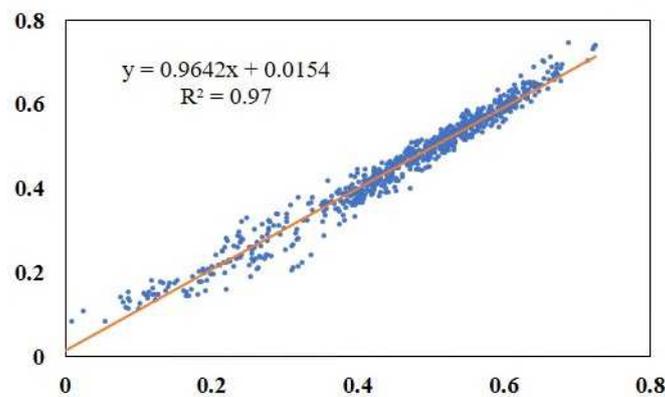


280
281 Fig. 6 Spatial distribution of RRU under current conditions (a), and RRUR under
282 current conditions(b), forest rehabilitation (c), grassland rehabilitation (d), cropping
283 rotations (e), parallel terrace (f).

284 4.2 Temporal and spatial situation of driving force

285 Global Moran's I is a coefficient that characterizes spatial autocorrelation. The
286 Global Moran's I of the RRUR is 0.44 and the z-score is 3.98, which indicates that the
287 distribution of the RRUR has a high positive spatial correlation, thus the assessment

288 model based on GTWR is applicable for regression analysis. Fig. 7 shows the regression
289 distribution of the sample. The regression result R^2 is 0.97, AICc is -3,417.29, and RSS
290 is 0.46. The R^2 is the determining coefficient between the actual value and simulated
291 value of the sample point. AICc is an important criterion for the goodness of model
292 fitting. The smaller the value is, the higher the model accuracy will be. The RSS is the
293 sum of the squares of the residuals. The results show that the driving force evaluation
294 ability of the model is strong, and the model calculation results are reliable and can
295 reflect the influence of each explanatory variable on the dependent variable.



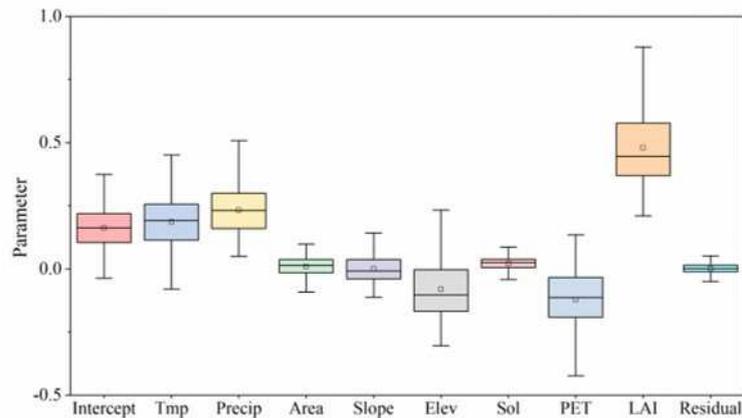
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Fig. 7 Regression distribution of the sample points

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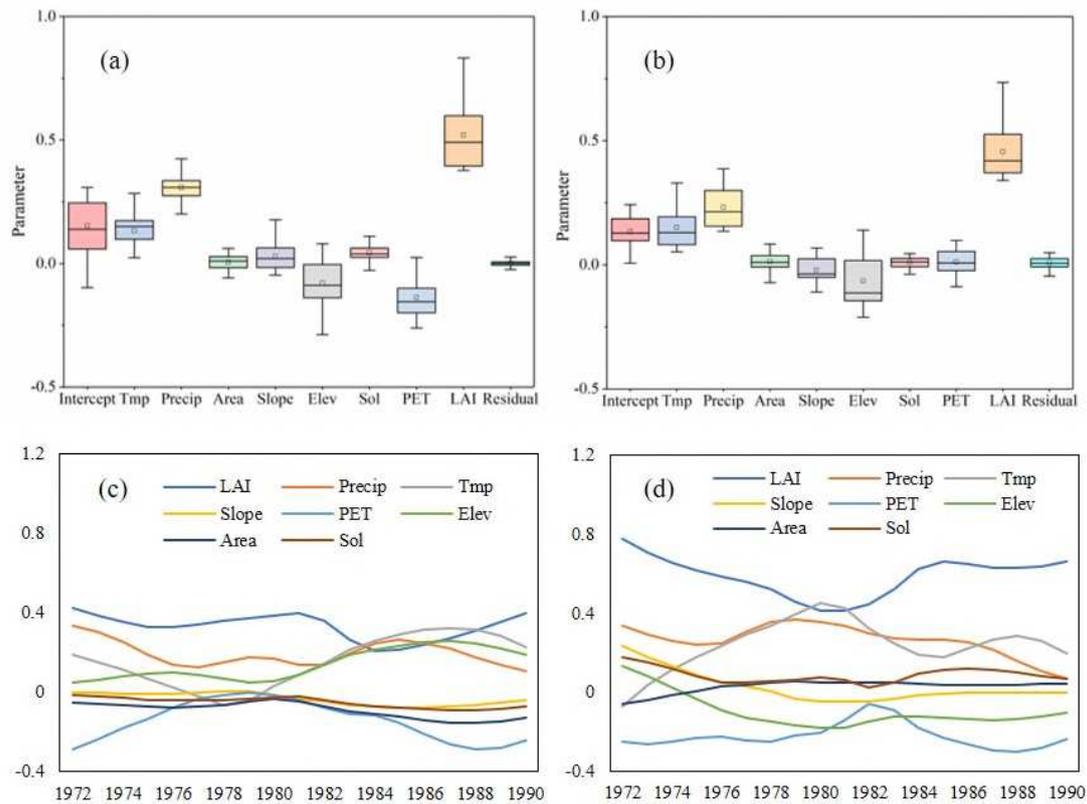
298 It can be seen from Fig. 8 that the regression coefficient values of explanatory
299 variables such as Elev, Tmp, PET, and LAI are widely distributed, which shows that
300 the driving effect of these factors on rainwater utilization exhibits significant spatial
301 heterogeneity and time non-stationarity. The LAI is the most important factor affecting
302 the utilization of rainwater. The mean value of regression coefficient is 0.49, and the
303 value of regression coefficient of some sample points exceeds 0.8. This shows that the
304 driving effect of vegetation conditions on the rainwater utilization in the research basin
305 is clear, exhibiting a positive correlation. The increase of the LAI will significantly
306 affect the water cycle process, while the evaporation will increase with the increase of
307 the LAI (Zhang et al. 2016,2015; Scott et al. 2013). With the increase of LAI, the

308 transpiration of vegetation will be enhanced, the evaporation of soil will be reduced,
309 and finally RRUR will be increased. The vegetation canopy can also effectively
310 intercept rainfall, reduce surface runoff, affect the circulation of atmospheric water, and
311 significantly alter the precipitation characteristics (Spracklen et al. 2012), further
312 increasing rainwater utilization.



313
314 Fig. 8 Value distribution of the regression coefficient of the explanatory variables

315 PET can characterize regional evaporation capacity. The average regression
316 coefficient in the research basin is -0.14, and the values are mainly distributed
317 downward. This shows that PET has a weak negative correlation with the utilization of
318 rainwater, and the areas with strong evaporation capacity may have a low RRUR. The
319 mean value of the regression coefficient of the Area, Slope, and Sol are very close to 0,
320 and the distribution is relatively concentrated. This reveals that the three explanatory
321 variables have no obvious driving effect on the rainwater utilization in the spatial scale.
322 The RRUR is basically not affected by elevation, and the average regression coefficient
323 of Elevation is -0.09. The regression coefficient distribution of Precip and Tmp in this
324 area is relatively consistent, and the mean values of the regression coefficient are 0.24
325 and 0.19 respectively. As an important aspect of the hydrological cycle of a basin,
326 meteorological factors have a positive driving effect on the utilization of rainwater. To
327 a certain extent, the increase of precipitation and temperature will increase the RRUR.



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Fig. 9 Distribution of regression coefficient of each explanatory variable in wet years(a) and dry years(b), and change of the regression coefficient values of each explanatory variable in sub-basin 1 (c) and sub-basin 39 (d)

The results of the assessment model show the driving effect of various variables on RRUR, and can be reasonably explained and analyzed. At the same time, the model can further compare the inter-annual and spatial differences of driving factors to support more detailed analysis and research. Fig. 9 shows the respective regression coefficient distribution of the explanatory variables in the wet (1979) and dry (1982) years. The mean values of the regression coefficient of the Prep and LAI in the wet years are greater than those in the dry years, which has a greater driving effect on the RRUR, and the Prep coefficient values are distributed intensively. The mean regression coefficient of the PET in wet years is smaller than that in dry years, and has a more substantial negative driving effect. The RRUR in dry years is less affected by factors other than meteorological and vegetation. Fig. 9 shows the comparison results of driving forces

343 between lower sub-basin (1) and upper sub-basin (39) of Zuli River Basin. Sub-basin
344 39 has more annual rainfall and better vegetation conditions than sub-basin 1. The
345 results show that RRUR in the upper reaches is affected by a greater number of factors,
346 while there are fewer major factors influencing the driving effect in the lower reaches.
347 In the upper reaches, the influence of LAI, Prep, and Tmp on RRUR is greater, while
348 in the lower reaches it is mainly affected by the LAI. Therefore, LAI is the most critical
349 driving force for rainwater utilization in arid and semi-arid regions. In the planning and
350 implementation of rainwater utilization measures in the basin, changes in LAI should
351 be properly considered.

352 Since the human behavior in this period did not have much sustained impact on
353 land use in research basin, human behavior is not included in the analysis of driving
354 forces. And the regression results also show that the choice of explanatory variables is
355 reasonable. In many areas, human intervention is the main culprit leading to climate
356 change and land degradation. It is necessary to consider the impact of human
357 intervention on rainwater utilization, and it can also be reflected indirectly through
358 natural factors.

359 **4.3 Application of rainwater utilization scenario**

360 Under global climate change and ecological protection, many regions are
361 implementing large-scale watershed management measures. The Loess Plateau where
362 the research basin is located is a typical arid and semi-arid region, and the most typical
363 area of loess distribution (Wei et al. 2017). In order to restrain the soil degradation,
364 China has taken continuous measures to strengthen the utilization of rainwater and the
365 restoration of the water environment in this area (Li and Qian 2018). Therefore, the
366 assessment model should be flexible, and scientific assessment can be carried out for
367 different management measures of the basin, to play a more positive role in practical.

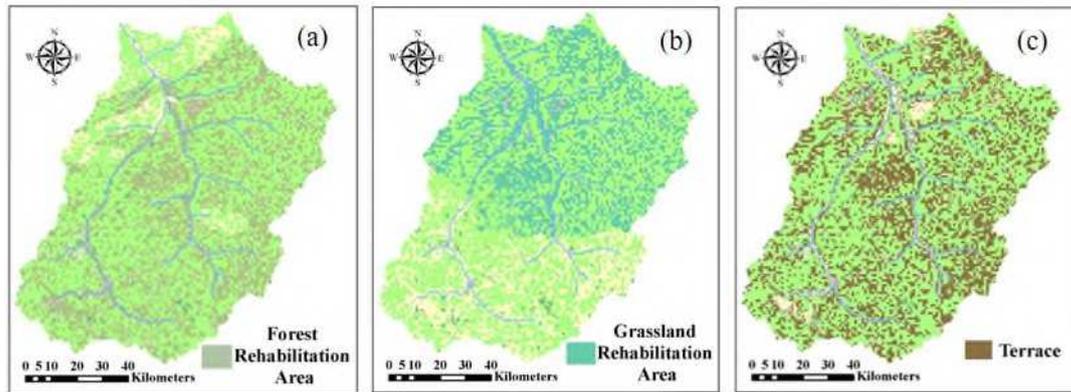
368 In order to verify the wide applicability of the evaluation model, a variety of land
 369 management measures were designed and evaluated in the research basin, including
 370 four scenarios of forest rehabilitation (Dan 2016), grassland rehabilitation (LIAN and
 371 HUANG 2019), cropping rotations (Lautenbach et al. 2013) and parallel terraces (Park
 372 et al. 2014), and the temporal and spatial changes of RRU and RRUR were analyzed.
 373 Fig. 10 shows the schematic diagram for the implementation of the forest rehabilitation,
 374 grassland rehabilitation and parallel terrace scenarios. And the specific implementation
 375 scheme is shown in Table 3.

376 Table 3 Specific evaluation scenarios

	Implementation scope	Change	Area (km ²)	Proportion
Forest rehabilitation	cultivated land of sub-basins with forest land growth under the natural conditions	mixed forest (Maximum LAI value is 4)	3,653	35%
Grassland rehabilitation	cultivated land in the middle and lower reaches	shrub bluestem (Maximum LAI value of 2)	2,550	24%
Parallel terraces	3°~25° gentle slope farmland	slope-to-terrace	3,593	34%
Cropping rotations	all the cultivated lands	corn-winter wheat-soybean-fallow	4,160	39%

377 **Note:** Since the scenarios research mainly focuses on the performance of the
 378 assessment model, rather than the design of scenarios, this scheme is just a hypothesis
 379 based on actual conditions.

380



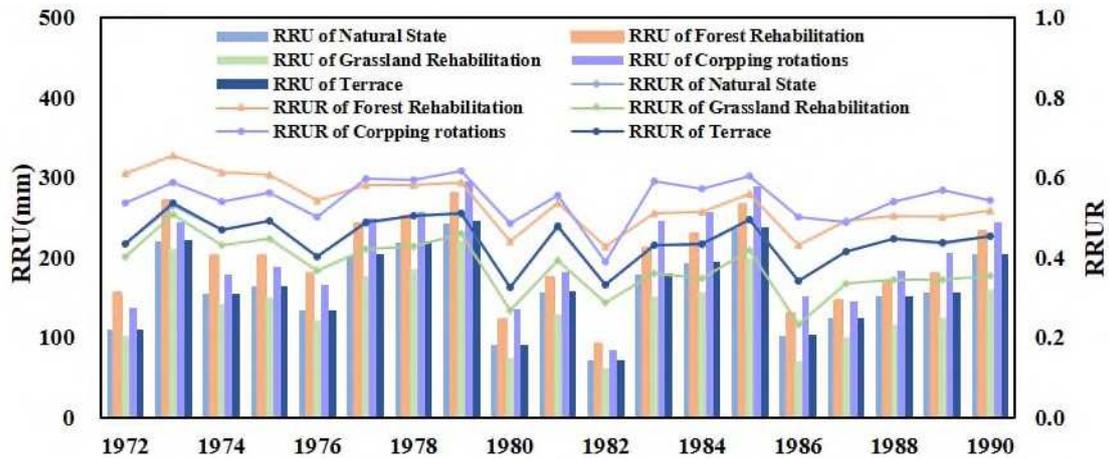
381 Fig. 10 Implementation area of rainwater utilization scenarios in the research basin,
382 forest rehabilitation (a), grassland rehabilitation (b), and parallel terrace (c).

383 4.3.1 Forest rehabilitation

384 Forest rehabilitation seeks to end the cultivation of sloping farmland that is prone
385 to soil erosion in a step-by-step and systematic manner, and to plant trees according to
386 local conditions and restore forest vegetation in accordance with the principle of
387 suitable land for suitable trees (Hai et al. 2015; Chen et al. 2015). To ensure the growth
388 environment of forest land, the cultivated land of sub-basins with forest land under the
389 current conditions is selected for forest rehabilitation (Fig. 10).

390 From the results of the assessment model, the RRUR has been significantly
391 improved after the implementation of forest rehabilitation scenario in the research basin,
392 the annual average value of the RRUR is 0.54, which is 0.1 higher than the original
393 situation (Fig. 11). The annual increase of RRU is 36.42mm, and the water volume is
394 $3.85 \times 10^8 \text{ m}^3$. In the years with more precipitation, the increase in RRUR is greater.
395 The increase in 1979 was 0.11 in comparison to the original situation. In 1983, with the
396 less rainfall and RRUR, the increase was 0.07. After afforestation, the annual average
397 value of soil evaporation in the basin was reduced by 38.64 mm, which effectively
398 reduced the loss of rainwater resources; however, the vegetation intercepted
399 evaporation increased slightly. The soil water content and groundwater recharge both

400 changed. The runoff at the outlet of the basin decreased by 10%. In terms of spatial
 401 distribution, the utilization rate of the area where the scenario has been implemented
 402 has been obviously improved, and the change in the upper reaches of the basin is
 403 relatively significant. And the area with high rainfall is more suitable for afforestation.
 404 The average RRUR in the sub-basin 43 has increased from 0.22 to 0.44.



405
 406 Fig. 11 RRU and RRUR of the Zuli River basin after implementation of each scenario

407 4.3.2 Grassland rehabilitation

408 Grassland rehabilitation aims to end the farming of cultivated land, plant grass to
 409 prevent and control water and soil loss, restore vegetation. According to the climate and
 410 grass growth characteristics in arid and semi-arid areas, the utilization scheme of
 411 rainwater resources dominated by grassland rehabilitation is generally implemented in
 412 areas with annual precipitation of less than 400 mm (Jinren 2004). Therefore, the
 413 grassland rehabilitation scheme chooses to convert the cultivated land in the middle and
 414 lower reaches of the Zuli River Basin (Fig. 10).

415 After the implementation of the scenario, the annual average value of the RRUR
 416 in the research basin is 0.38, which is 15% lower than the original situation, and the
 417 reduction of RRU is 54.63 mm. In the areas where the grassland rehabilitation measures
 418 have been taken, neither the RRU nor the RRUR reach a high level (Fig. 11). The

419 selected species of returning grass in the scheme is bluestem, the leaf surface and
420 canopy conditions are poor during the growth period, and the implementation area is
421 relatively dry, none of which effectively increase the vegetation transpiration or soil
422 water content. The original cultivated land crop is grain, and the vegetation condition
423 is better than grassland overall. Therefore, this led to the model evaluation results of
424 the scheme, which also verifies the results of the driving force analysis.

425 4.3.3 Crop rotations

426 Crop rotation refers to the planting pattern of sequential rotation of different crops,
427 or multiple cropping combinations between seasons and years in the same field. The
428 plan is selected to be implemented in all of the cultivated lands of the Zuli River Basin,
429 which changes from original grain of one season a year to "corn-winter wheat-soybean-
430 fallow" of three seasons in two years.

431 The annual average value of the RRUR under the cropping rotations scenario is
432 0.54, which is 0.10 higher than the original situation (Fig. 11). In 1979, which had a
433 rather high precipitation, the RRUR reached 0.62, with an increase of 0.18 compared
434 with the original situation, and in 1982, which had less rainfall, the increase was 0.06.
435 The overall leaf area index is shown to have increased by 30% on average, the annual
436 average vegetation transpiration has increased by 26%, and the soil evaporation has
437 decreased by 31%. Particularly during the growth period of winter wheat, rainwater
438 resources have been effectively utilized. The runoff, soil water content and groundwater
439 recharge of the basin are shown to have decreased. In terms of spatial distribution,
440 RRUR has increased even more in the sub-basins with more cultivated land and greater
441 rainfall.

442 4.3.4 Parallel terrace

443 Parallel terraces are terraced fields that change gentle slopes into horizontal ones

444 for the purpose of soil and water conservation and planting of crops or trees. In China's
445 arid and semi-arid areas, parallel terraces are widely used to control soil erosion,
446 increase soil water storage, and regulate rainwater resources. In this study, the 3°~25°
447 gentle slope farmland in the Zuli River Basin is selected to implement the slope-to-
448 terrace measure.

449 After the implementation of scenario, the model results show that the annual
450 growth rate of RRUR is very small (Fig. 11). The annual average value of the RRUR is
451 the same as the original situation. In the years with high rainfall (1976, 1979, 1985),
452 the RRUR is shown to have increased by 0.01 compared with the original situation.
453 There is almost no change in RRUR in years with low rainfall. Analyzing the reasons,
454 terrace is the transformation of microtopography of sloping farmland, which mainly
455 lags the process of yield and confluence of sloping land, reduces runoff in the basin
456 (Arnáez et al. 2015; Yang et al. 2009), but has little impact on vegetation water cycle.
457 In addition, the changes in the hydrological cycle process are not so obvious on the
458 annual scale. On the scale of space, the increase in the RRUR in the upper reaches of
459 the basin with higher rainfall is greater than that in the middle and lower reaches with
460 lower rainfall.

461 **4.4 Discussion**

462 Through the application of various scenarios, the assessment model shows wide
463 applicability and guiding value of regional rainwater utilization. We use this model to
464 evaluate the utilization of rainwater resources area and analyze the driving factors of
465 the study on the spatial and temporal scales. In the process of evaluation of various
466 scenarios, the above analysis conclusion is confirmed and the evaluation results are also
467 consistent with some of our previous understandings of hydrology. Due to the large
468 evaporation and low runoff coefficient in the research areas, rainwater is mainly lost

469 through the evaporation of soil and vegetation. The RRU and RRUR are more related
470 to vegetation. Land cover is an important factor that affects the hydrology of the basin
471 (Caja et al. 2018). The main driving factor is LAI, and the variation of RRUR was
472 consistent with that of LAI. In other arid climate areas, studies have also shown that
473 strategies to improve ecosystem services should rely on adequate management of
474 vegetation to reduce surface flow, and this reduction contributes to higher soil moisture
475 contents, less intense erosive processes, and higher stocked soil carbon (Andrade et al.
476 2020). The model can also quantitatively evaluate the temporal and spatial changes of
477 rainwater resources and hydrological elements in a watershed, which is of great
478 significance for guiding agricultural river basins to take rainwater utilization and land
479 management measures (Glendenning et al. 2012).

480 Compared with model applicability, model evaluation accuracy is also very
481 important. The hydrological simulation of the assessment model is based on the SWAT,
482 and its calculation effect has been widely verified. The selection of time scale may
483 affect the evaluation results because the rainfall runoff and evaporation process in the
484 basin have a certain lag. There may be errors in the evaluation on a short time scale.
485 However, on the longer scale (monthly or annual scale), the error will be significantly
486 reduced, and the model can accurately evaluate the utilization of rainwater resources.

487 While verifying the performance of the model, the utilization status of rainwater
488 resources after the management measures in the study area was evaluated in detail.
489 However, it should be noted that multiple drivers often interact in complex and non-
490 additive ways (Wen et al. 2018). For example, on a wider spatial and ecological scale,
491 although the increase of forest land coverage positively related to rainwater use and soil
492 water storage, it may have a negative impact on other ecological services, especially
493 food production (Iacob et al. 2014). And it may affect the interaction between riparian

494 forest land, terrestrial and aquatic ecosystems, leading to nutrient migration,
495 microclimates, and ecological changes in floodplains (Swanson et al. 1991). In addition,
496 rainfall variability, runoff quality and quantity, local skills and investment capacity,
497 labour availability and institutional support also influence sustainability of rainwater
498 harvesting and utilization (Pachpute et al. 2009). Although, the possible wide-ranging
499 impact is not considered in the performance verification process of this assessment
500 model, it can consider these by refining the simulation settings to calculate more
501 elements. Therefore, the model provides as detailed evaluation performance as possible
502 for rainwater utilization and hydrological elements, and other functions such as
503 ecosystem services assessment can be added in the later stage, considering the
504 provisioning, regulating, cultural and other aspects to support the watershed
505 management.

506 **5 Conclusions**

507 Improving the utilization rate of rainwater resources is one of the keys to alleviate
508 the ecological and environmental problems in arid and semi-arid areas. The rainwater
509 utilization or management in the watershed should be based on adequate evaluation
510 studies. We put forward the concept of RRU and RRUR, and construct a spatio-
511 temporal assessment model for the rainwater resources utilization and driving forces.
512 Taking the Zuli River Basin of the Loess Plateau as the research area, and extensive
513 performance verification was carried out to analyze the applicability and accuracy of
514 the model.

515 The main conclusions and findings of this study can be summarized as follows:

516 (1) Overall, the performance of the assessment model is very good, which can
517 accurately evaluate the utilization of rainwater resources and driving factors in the
518 watershed on the spatial and temporal scale. Accurate calibration of the hydrological

519 process and appropriate selection of influencing factors can ensure that the evaluation
520 results of rainwater resources utilization and driving force are credible.

521 (2) The RRU and RRUR in the Zuli River Basin have a positive correlation with
522 precipitation on the spatial and temporal scale. The annual average value of the RRUR
523 is 0.44, and the average value of the RRU is 164 mm. The RRUR in the upper and
524 middle reaches of the basin are relatively high and it shows a decreasing trend overall.
525 The LAI is the greatest driving factor of rainwater utilization.

526 (3) The four self-designed scenarios of watershed management are evaluated by
527 the model in the basin. The annual average RRUR of forest rehabilitation and cropping
528 rotations increased by 0.1 compared with the original state. The RRUR of grassland
529 rehabilitation has been reduced by 15%, and the increase of RRUR is not significant
530 after taking the parallel terrace measurements. The assessment model based on SWAT
531 and GTWR provides a strong spatial and temporal evaluation function of rainwater
532 utilization, and it also inherits their advantages and disadvantages.

533 In this paper, the spatial scale of assessment model verification is relatively large,
534 and smaller-scale studies can be conducted in the future to explore the performance of
535 the model (for example, on the hydrological response unit). At the same time, based on
536 the planned river basin management measures, considering potential negative impacts,
537 further discussions the changes and impacts of regional flood control, agriculture, and
538 social-economy can be carried out through the assessment model to comprehensively
539 study rainwater utilization in arid and semi-arid regions.

540

541 **Declarations**

542 **Ethical Approval**

543 The authors declare that the submitted manuscript is original and unpublished

544 elsewhere, and that this manuscript complies with the Ethical Rules applicable for this
545 journal.

546 **Consent to participate**

547 All the authors consent to participate in this research work.

548 **Consent for publication**

549 All the authors consent to publish this work.

550 **Authors Contributions**

551 F. Li and X. Yuan contributed to the study conception and design. Analysis was
552 performed by F. Li and X. Zhang. The first draft of the manuscript was written by X.
553 Zhang and all authors commented on previous versions of the manuscript. All authors
554 read and approved the final manuscript.

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558 **Competing Interests**

559 The authors declare no conflict of interest.

560 **Availability of data and materials**

561 Data and materials are available upon request to the corresponding author.

562 **Code availability**

563 Code is available upon request to the corresponding author.

564

565 **References:**

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