



# 1 Implications of water management representations for watershed 2 hydrologic modeling in the Yakima River Basin 3

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10

11 **Abstract.** Water management substantially alters natural regimes of streamflow through modifying retention time and water exchanges  
12 among different components of the terrestrial water cycle. Accurate simulation of water cycling in intensively managed watersheds, such as the  
13 Yakima River Basin (YRB) in the Pacific Northwest of the U.S., faces challenges in reliably characterizing influences of management  
14 practices (e.g., reservoir operation and cropland irrigation) on the watershed hydrology. Using the Soil and Water Assessment Tool (SWAT)  
15 model, we evaluated streamflow simulations in the YRB based on different reservoir operation and irrigation schemes. Simulated streamflow  
16 with the reservoir operation scheme optimized by the RiverWare model better reproduced measured streamflow than the simulation using  
17 default SWAT reservoir operation scheme. Scenarios with irrigation practices demonstrated higher water losses through evapotranspiration  
18 (ET), and matched benchmark data better than the scenario that only considered reservoir operations. Results of this study highlight the  
19 importance of reliably representing reservoir operations and irrigation management for credible modeling of watershed hydrology. Both  
20 SWAT and RiverWare are community-based and have been widely tested and applied for reservoir operations and agricultural watershed  
21 modeling in regions across the globe. As such, the methods and findings presented here hold the promise to apply to other intensively managed  
22 watersheds to enhance water resources assessment.

23

24 **Keywords:** Reservoir operation; Irrigation; Managed watershed; RiverWare; SWAT

25

## 26 1. Introduction

27 Ever-intensifying human activities have profoundly affected terrestrial water cycling across the globe (Jackson et al., 2001),  
28 particularly at the watershed scale (Vörösmarty and Sahagian, 2000; Yang et al., 2015; Yang et al., 2014). Water management  
29 substantially alters natural regimes of streamflow through modifying retention time and water exchanges among different



30 components of the terrestrial water cycle (Haddeland et al., 2007). Hydrologic consequences of management activities should  
31 be explicitly investigated for effective water resource management (Siebert et al., 2010), especially for watersheds striving to  
32 maintain sustainable water supply for multiple users. Accurate simulation of water cycling in intensively managed watersheds  
33 faces challenges in reliably characterizing influences of management practices (e.g., reservoir operations and cropland  
34 irrigation) on the hydrologic cycling (Wada et al., 2017). Explicit analyses of how model representations of water  
35 impoundments and withdrawals would affect hydrologic modeling are needed to advance knowledge of water cycling in  
36 managed watersheds.

37 Construction of dams and reservoirs has substantial influences on the magnitude and variability of downstream runoff  
38 (Lu and Siew, 2006; Vicente-Serrano et al., 2017). For example, reservoir operations reduced 9% - 25% of summer runoff to  
39 the Pacific Ocean in western U.S. and Mexico (Haddeland et al., 2007). In heavily dammed regions, reduction of streamflow  
40 following dam construction even reached 100% (Graf, 1999). Reservoir operations affect the temporal variability of streamflow  
41 at multiple temporal scales in different regions across the globe (Huang et al., 2015; Zajac et al., 2017). Regulated streamflow  
42 from reservoirs to downstream areas contributes to attenuating flood peaks and volumes, but could increase baseflow in dry  
43 seasons (Batalla et al., 2004).

44 Reliable representation of reservoir operations in hydrological models is critical for credible simulation of water  
45 cycling (Coerver et al., 2018). To characterize impacts of reservoir operations on watershed hydrology, multiple methods have  
46 been developed to simulate reservoir releases. These models include mathematical tools which optimize water release for  
47 achieving management objectives, simulation models which consider physical processes of water cycling in reservoirs to allow  
48 users to evaluate impacts of different management alternatives on reservoir storages and releases, and a combination of these  
49 two types of models for reservoir planning and management (Branets et al., 2009; Dogrul et al., 2016; Yeh, 1985). Among  
50 these models, the RiverWare model and models developed based on RiverWare consider both management policies and  
51 physical processes (Zagona et al., 2001), and have proven capability of simulating reservoir storages and downstream flows.  
52 However, how reservoir operations affect watershed hydrology is still not explicitly examined.

53 In addition to reservoir operations, cropland irrigation also affects watershed hydrology. Water withdrawal for  
54 irrigation has been widely adopted to increase crop production in arid and semi-arid regions. Water redistribution through  
55 irrigation enhances water and energy fluxes between soils and the atmosphere (Rost et al., 2008; Sacks et al., 2009), and results  
56 in elevated water loss through evapotranspiration (Hao et al., 2015; Malek et al., 2017; Polo and Losada, 2016), and depletion  
57 of water resources (Aeschbach-Hertig and Gleeson, 2012) in different regions of the world. To better simulate impacts of  
58 irrigation, numerical models have been developed to quantify water fluxes among soils, vegetation, and water bodies induced  
59 by irrigation (Leng et al., 2013; Santhi et al., 2005). Impacts of irrigation on watershed hydrology should be further evaluated  
60 to application of this tool for effective management of water resources in basins with competing demands for water.



61           The Soil and Water Assessment Tool (SWAT) has been widely used to simulate water cycle dynamics in response to  
62 management practices across the watershed and regional scales (Arnold et al., 1998). Previous studies indicated that the default  
63 SWAT reservoir operation scheme which simulates water release based on target storages may either overestimate reservoir  
64 storages in no-flood seasons (Lv et al., 2016), or underestimate water releases when actual reservoir storages are lower than  
65 target storages (Wu and Chen, 2012). SWAT simulates water withdrawal for irrigation from different water sources (e.g.,  
66 reservoirs, streams, and groundwater aquifers). Multiple efforts have employed SWAT to evaluate impacts of different  
67 irrigation practices on watershed hydrology (Ahmadzadeh et al., 2016; Chen et al., 2017; Maier and Dietrich, 2016), and  
68 emphasized the importance of balancing water supply and irrigation demands in hydrologic simulations. However, applicability  
69 of SWAT in watersheds with interacting reservoir operations and irrigation has not been well studied, and thus deserves further  
70 investigation to inform effective water resource management.

71           The Yakima River Basin (YRB) in the Pacific Northwest of the U.S. has been regulated for regional hydropower,  
72 flood control, fishery, crop cultivation, and drinking water supply. Water supply for irrigation is one of the most important  
73 water resource management objectives in the YRB (USBR, 2012). The Yakima River Reservoir system supplies water to  
74 180,000 hectares of cropland through the operation of five reservoirs which store ca.30% of the mean annual runoff of the basin  
75 (Vano et al., 2010). Reservoir operations and cropland irrigation in the YRB altered historical streamflow regimes, resulted in  
76 severe low flow, and elevated flow events. Since the 1990s, increasing demands for irrigation, municipal water consumption,  
77 and critical environmental flow for conserving wildlife habitats in the context of climate change have challenged water resource  
78 management in the basin. Thus, there is an urgent need to reliably simulate water cycling in the basin to provide a solid basis  
79 for policy formulation and management actions which strive to achieve a balance among water demands for different purposes  
80 (Poff et al., 2003).

81           In recognition of the challenges in modeling hydrology in heavily managed watersheds, this study aimed to explore  
82 how different schemes of reservoir operations and irrigation affect streamflow modeling in the YRB. Using the YRB as a  
83 testbed, we evaluated streamflow simulations with different model representations of management activities. The knowledge  
84 discovered through our numerical experiments is expected to help understand uncertainties in water cycling simulations  
85 resulted from water management representations in hydrological models. Specifically, objectives of this study are to (1)  
86 examine how different representations of reservoir operations influence watershed streamflow simulations, and (2) assess  
87 impacts of cropland irrigation on watershed hydrology. Both SWAT and RiverWare are community models that have been  
88 widely tested and applied in diverse regions across the globe, as evidenced by the numerous peer-reviewed publications in the  
89 fields of reservoir operation (<https://www.colorado.edu/cadswes/publications/journal-articles>) and watershed modeling  
90 ([https://www.card.iastate.edu/swat\\_articles/](https://www.card.iastate.edu/swat_articles/)). Therefore, methods and findings derived from this study hold the promise to  
91 provide valuable information for improving hydrologic modeling in intensively managed basins across the globe.



## 92 2. Materials and methods

### 93 2.1. Study area

94

95 [Figure 1]

96 The Yakima River Basin (Figure 1) is located in central Washington, U.S. (N45.98~47.60, W121.53~119.20). The basin has a  
97 semi-arid climate with a Mediterranean precipitation pattern. Winters are cold, with a mean temperature of 28.5 °F. Annual  
98 average precipitation is ca. 675mm, with an average snowfall of 21.7 inches (55.1 cm), occurring mainly in December and  
99 January. Rangeland, forest, and cropland, are the primary land uses in the basin, and cover 36%, 33%, and 28% of the study  
100 area (Vaccaro and Olsen, 2007), respectively. Dams were built throughout the basin for the irrigated agriculture. There are five  
101 big reservoirs in the YRB, including Keechelus, Kachess, Cle Elum, Bumping, and Rimrock (Figure 1). (Malek et al., 2016)  
102 reported that the YRB experienced major droughts in 20% of the years between 1980 and 2010, and the frequency may double  
103 in the future. It is expected that the increasing competition for water from multiple users, especially for irrigation, fishery, and  
104 wildlife habitats, may escalate in the coming decades (Miles et al., 2000).

### 105 2.2. Management Schemes in SWAT and RiverWare model

#### 106 2.2.1. Reservoir operation schemes

107

108 [Table 1]

109 Settings of the five reservoirs, including locations, height, storage capacity, operating purpose, and surface area were compiled  
110 and added to SWAT input files (Table 1). We use three scenarios (R0, R1, and R2) to evaluate reservoir operation simulations  
111 in the YRB. Scenario R0 does not simulate reservoir operations and we use it as a baseline scenario. Scenario R1 uses the  
112 SWAT model's built-in reservoir management schemes which specifies monthly target volumes for managed reservoirs  
113 (Neitsch et al., 2011). Under the R2 scenario, the SWAT model uses reservoir releases calculated by the RiverWare model as  
114 the outflow from these reservoirs to downstream reaches.

115 The SWAT model calculates water balance for a reservoir on a daily scale as follows:

$$116 V_{net} = V_{stored} + V_{flowin} - V_{flowout} + V_{pcp} - V_{evap} - V_{seep} \quad (1)$$

117 where  $V_{net}$  is net volume changes of a reservoir on a given day ( $m^3$  water);  $V_{stored}$  is the water stored in a reservoir at the  
118 beginning of a day ( $m^3$  water);  $V_{flowin}$  is the water entering a reservoir in one day ( $m^3$  water);  $V_{flowout}$  is the amount of water  
119 release to downstream reaches of a reservoir ( $m^3$  water);  $V_{pcp}$  is the amount of water falling to a reservoir in one day ( $m^3$



120 water);  $V_{evap}$  is the water loss through evaporation from a reservoir ( $m^3$  water);  $V_{seep}$  is the amount of water loss through  
 121 seepage in a reservoir ( $m^3$  water).

122 Under the R1 scenario, the target release approach calculates reservoir storage using the following equations:

$$123 \quad V_{targ} = V_{em}, \text{ if } mon_{fld,beg} < mon < mon_{fld,end} \quad (2)$$

$$124 \quad V_{targ} = V_{pr} + \frac{\left(1 - \min\left[\frac{SW}{FC}, 1\right]\right)}{2} \cdot (V_{em} - V_{pr}), \text{ if } mon \leq mon_{fld,beg} \text{ or } mon \geq mon_{fld,end} \quad (3)$$

125 where  $V_{targ}$  is the target reservoir storage of a given day ( $m^3$  water);  $V_{em}$  is the volume of reservoir for filling to the  
 126 emergency spillway ( $m^3$  water);  $mon$  is the month of the year;  $mon_{fld,beg}$  is the beginning month of a flood season;  
 127  $mon_{fld,end}$  is the ending month of the flood season;  $V_{pr}$  is the reservoir volume when filled to the principal spillway ( $m^3$   
 128 water);  $SW$  is average soil water content (mm) on a given day, and  $FC$  is field capacity (mm).

129 With the target volume is determined, the reservoir outflow ( $V_{flowout}$ ,  $m^3/day$ ) for a given day is calculated as  
 130 follows:

$$131 \quad V_{flowout} = \frac{V_{stored} - V_{targ}}{ND_{targ}} \quad (4)$$

132 where  $V_{stored}$  is the volume of water stored in the reservoir on a given day; and  $ND_{targ}$  is the number of days required for  
 133 the reservoir to reach the target storage.

134 Under the R2 scenario, outflow from a reservoir is calculated based on the estimated daily release provided by the  
 135 RiverWare model as follows:

$$136 \quad V_{flowout} = 86400 \cdot q_{out} \quad (5)$$

137 where  $V_{flowout}$  is the volume of water flowing out of a reservoir in one day ( $m^3$ ) and  $q_{out}$  is the outflow rate estimated by  
 138 RiverWare ( $m^3/s$ ).



139 RiverWare simulates operations and scheduling of reservoir management objectives, including hydropower  
140 production, flood control, and irrigation (Zagona et al., 2001). RiverWare can model a variety of physical processes for  
141 reservoirs with computational time steps ranging from one hour to one year. In RiverWare simulations, the solver is based on  
142 operating rules or operating policies that provide instructions for operation decisions such as reservoir releases (Zagona et al.,  
143 2001). The rules are strictly prioritized, with high priority rules requiring that reservoir release should not be less than the  
144 minimum flow for downstream reaches; whereas a low priority rule requires that reservoir storage should fit a seasonal guide  
145 curve value. Conflicts are resolved by giving higher priority rules precedence. This model has been applied to the YRB to  
146 simulate outflow from the reservoirs (USBR, 2012).

#### 147 2.2.2. Irrigation representation in the SWAT model

148 SWAT irrigation schemes consider multiple water sources including reservoirs, streams, shallow aquifers, and sources outside  
149 the watershed. Irrigation can be triggered by a water stress threshold (a fraction of potential plant growth). In SWAT, water  
150 stress is simulated as a function of actual and potential plant transpiration:

$$151 \quad wstr = 1 - \frac{E_{t,act}}{E_t} = 1 - \frac{w_{actualup}}{E_t} \quad (6)$$

152 where  $wstr$  is the water stress;  $E_t$  is the potential plant transpiration (mm/day);  $E_{t,act}$  is the actual amount of transpiration  
153 (mm/day) and  $w_{actualup}$  is the total plant water uptake (mm/day). The plant water uptake is a function of the maximum plant  
154 transpiration, a water-use distribution parameter, the depth of the soil layer and the depth of plant root. In the SWAT auto  
155 irrigation algorithm, irrigation is applied when the water stress factor falls below a predefined threshold. Irrigation will increase  
156 soil moisture to field capacity, if irrigation water sources could provide enough water. We conducted two additional simulations  
157 by assuming that irrigation water was withdrawn from reservoirs and streams (R2S1), or groundwater (R2S2), based on the  
158 simulations with RiverWare reservoir schemes (R2).

#### 159 2.3. Model setup, sensitivity analyses, and simulations

160

161 [Table 2]

162

163 We used a plethora of geospatial datasets to parameterize and drive hydrological simulations in the YRB (Table 2). Topography  
164 information was derived from U.S. Geological Survey (USGS) National Elevation Dataset (NED) (<https://data.cr.usgs.gov/NED>)  
165 with a spatial resolution of 30 meters. The U.S. Department of Agriculture (USDA) Crop Data Layer (CDL)



166 (<https://nassgeodata.gmu.edu/CropScape/>) with a spatial resolution of 30 meters was used to obtain land covers including  
167 shrubland, forestland, grassland, developed land and barren land, cultivated land and orchard in the YRB (Figure 1). We  
168 derived daily climate data for the period of 1980-2012 from North America Land Data Assimilation System (NLDAS)  
169 (<https://ldas.gsfc.nasa.gov/nldas/NLDAS2forcing.php>). In addition, we obtained nitrogen and phosphorus fertilizer application  
170 rates (USDA-ERS, 2018), tillage intensity (CTIC, 2008), and planting and harvesting (USDA, 2010) for crop management.  
171 The SWAT model divides the YRB into 181 subbasins and 1950 hydrologic response units (HRUs). Streamflow simulations  
172 in four subbasins (Figure 1) with long-term observations were explicitly examined to evaluate how different schemes affected  
173 model performances. We also collected the Moderate Resolution Imaging Spectroradiometer (MODIS) evapotranspiration (ET)  
174 data to evaluate SWAT ET simulations (Mu et al., 2011).

175 We quantified parameter sensitivities with a global sensitivity method described by (Abbaspour et al., 2017), which  
176 employs model runs driven by randomly sampled parameter sets, a multi-regression approach, and a T-test to identify and rank  
177 sensitive parameters. Sensitivity analysis for SWAT simulations in the YRB is computationally expensive. For each scenario,  
178 we spent about three weeks to run SWAT 10000 times (Zhang et al., 2009a; Zhang et al., 2009b) to understand parameter  
179 sensitivity and minimize discrepancy between simulations and observations under different scenarios. We used Nash–Sutcliffe  
180 efficiency coefficient ( $Ens$ ) (Nash and Sutcliffe, 1970) and correlation coefficient ( $r$ ) (Legates and McCabe, 1999) as the  
181 metrics to evaluate model performance.

### 182 3. Results

#### 183 3.1. Parameter sensitivity under different scenarios

184

185

[Table 3]

186

187 Table 3 shows the ranking of parameter sensitivity under different scenarios. In general, selected parameters demonstrated  
188 similar sensitivities among all scenarios, particularly for the ten most sensitive parameters, indicating that the five scenarios  
189 captured critical processes regulating water cycling in the basin. For all scenarios, the most sensitive parameters are CN2 and  
190 the snow factors, including SFTMP, SMTMP, SMFMX, SMFMN, and TIMP, indicating that snowmelt is the key hydrological  
191 process in the YRB. SWAT uses the Soil Conservation Service curve number method (SCS-CN) to predict runoff. As a result,  
192 parameter CN2 affects the partition of water between surface runoff and infiltration, and has significant impacts on streamflow  
193 estimates. We also observed that sensitivities of several parameters were different among the five scenarios. Specifically,  
194 parameters relevant to reservoir operations or irrigation management, including the RES\_K and NDTARGR, played key roles  
195 in simulations with reservoir operations. The differences could be attributed to the inclusion of reservoir operation and irrigation



196 schemes, and further suggest that significant impacts of the management activities on water cycling should be considered in  
197 hydrologic modeling.

### 198 3.2. Streamflow simulations under different reservoir operation scenarios (R0, R1, and R2)

199

200 [Figure 2]

201 [Figure 3]

202 [Figure 4]

203 Without considering impacts of reservoir operations and water withdrawals on water cycling, the R0 scenario demonstrated  
204 poor performance in streamflow simulations (Figure 2). Streamflow simulations in R1 and R2 were significantly improved  
205 when reservoir operation schemes were added to SWAT, which further confirmed the importance of considering reservoir  
206 operations in hydrologic modeling in the YRB. Note that reservoirs either increase or reduce streamflow, as reservoirs could  
207 increase water release water in dry seasons, or retain upstream water for flood control in wet seasons. In addition, streamflow  
208 simulated in the R2 scenario showed a better agreement with measured flow than that of the R1 scenario. R2 exhibits better  
209 *Ens* and *r* in three of the four subbasins than R1, indicating that reservoir outflow estimated by RiverWare more accurately  
210 simulated water releases than the default reservoir operation scheme in SWAT. The streamflow simulations in subbasins 67  
211 and 99 were more sensitive to the different reservoir schemes, as evidenced by greater improvements in the *Ens* and *r* values  
212 than those of the other two downstream subbasins (Figures 3 and 4).

213 [Figure 5]

214 We also compared ET simulations of the YRB under the three scenarios (R0, R1, and R2). Specifically, monthly ET  
215 of the R0 scenario was lower than the other two scenarios (Figure 5). ET estimates increased in May and June, but decreased  
216 in winter in R1 and R2 simulations. In addition, annual ET increased by 7.83% and 8.05% for R1 and R2 simulations relative  
217 to the R0 simulation, respectively. The changes could be attributed to increased evaporation from reservoirs.

### 218 3.3. Streamflow and ET simulations under the two irrigation operation scenarios (R2S1 and R2S2)

#### 219 3.3.1. Streamflow and ET

220

221 [Figure 6]

222 [Figure 7]



223 Settings of scenario R2S1, which used reservoirs and streams as water sources for irrigation, are consistent with the actual  
224 irrigation practices in the YRB where surface water is the primary irrigation water source (Figure 6). For the R2S2 scenario,  
225 shallow groundwater was assumed to be the water source for irrigation (Figure 7). Consequently, streamflow simulations under  
226 the scenario R2S1 matched observations better than that in R2S2. Compared with the R2 scenario, the simulated flow decreased  
227 by 24.87% and 31.29% in R2S1 and R2S2, respectively.

228 [Figure 8]

229 ET is an important component of terrestrial water cycling and this variable is used in the calculation of irrigation  
230 demand in SWAT simulations. Figure 8 compares simulated monthly ET of the irrigation scenarios (R2S1 and R2S2) with the  
231 RiverWare reservoir operation scenario (R2) which did not consider irrigation. The mean monthly ET rates of the irrigation  
232 scenarios (R2S1 and R2S2) were significantly higher than simulations without irrigation, particularly during March-July, when  
233 irrigation was applied to support crop growth.

234 [Figure 9]

235 We further compared simulated annual ET in the R2S1 and R2 scenarios, and evaluated model simulations against  
236 the MODIS ET data (Figure 9). We observed substantially underestimated cropland ET in the R2 scenario relative to the  
237 MODIS ET. When irrigation was included in our simulation, biases in ET estimates were reduced from 38% to 14%, compared  
238 with the MODIS ET data. The comparison demonstrated that inclusion of irrigation schemes achieved better estimates of water  
239 losses during irrigation, and contributed to enhancing streamflow simulations (Figure 6). In addition to magnitude, the irrigation  
240 scenario (R2S1) also simulated well interannual variability of ET, as evidenced by the high coefficient of determination in the  
241 scatter plot against MODIS ET (Figure 9).

### 242 3.3.2. Irrigation water consumption

243 The mean annual irrigation depth for the irrigation scenarios of R2S1 and R2S2 was 480.66 mm/year and 228.46 mm/year,  
244 respectively. Under the R2S1 scenario, water for irrigation was provided by the five reservoirs in the corresponding subbasins;  
245 in subbasins without reservoirs, irrigation water was withdrawn from local streams. Average irrigation water was higher in the  
246 R2S1 scenario than that of R2S2. There are notable differences in irrigation depths for different crop species between the two  
247 irrigation scenarios. In general, the irrigation water consumption for all crops was higher in the R2S1 scenario than that of the  
248 R2S2 scenario.

### 249 3.4. Management impacts on watershed hydrology

250 As indicated by the improved *Ens* and *r* values, streamflow simulations under scenarios simulating both reservoir operations  
251 and irrigation schemes (R2S1 and R2S2) are more comparable with observations than those of the baseline scenario (R0) which  
252 does not consider water management activities in the simulation. Reservoirs have contributed to streamflow increases in dry



253 periods and streamflow reduction in wet seasons by regulating water storage and release. Compared with the baseline scenario  
254 (R0), we found reductions in simulated streamflow in the scenarios that consider reservoir and irrigation operations, indicating  
255 that water withdrawal for irrigation tends to reduce streamflow as a result of enhanced water loss through ET.

256 ET in the composite scenarios (R2S1 and R2S2) was higher than the R0 scenario, which can be attributed to the  
257 elevated evaporation from reservoirs and irrigated cropland. Direct evaporation from reservoirs increased by 7% - 8% over the  
258 study period (1980s to 2010) due to improved simulation of reservoir surface areas in the R1 and R2 simulations relative to the  
259 R0 simulation. Irrigation practices led to more pronounced increases in ET in R2S1 and R2S2 simulations as compared with  
260 that of R2 (Figure 8). These results indicate that irrigation may have more pronounced impacts on water cycling through  
261 stimulating ET across the basin than reservoir operations.

## 262 4. Discussion

### 263 4.1. SWAT simulation of water cycling in response to management activities

264 In recent decades, water users of the YRB passed the Yakima River Integrated Water Management Plan, which is a  
265 comprehensive agreement that advances water infrastructures and management (USBR, 2012). Enhanced hydrologic modeling  
266 provided by this study will provide valuable information for goals of the Integrated Plan, which requires accurate streamflow  
267 information to manage water resources to meet ecological objectives as well as to secure water supply for domestic uses.

268 Although previous investigations highlighted the importance of irrigation and reservoir management to water balance  
269 and availability (Hillman et al., 2012; Malek et al., 2014), joint impacts of these two water management practices on watershed  
270 hydrology have not been fully understood. In recognition of this challenge, we enhanced SWAT representations of the two  
271 critical water management activities, including reservoir operations and irrigation, to constrain uncertainties in hydrologic  
272 simulations. We achieved improved model performances through including the two activities in the SWAT modeling  
273 framework. The simulated streamflow was generally lower in simulations with management activities than the baseline  
274 simulation (R0). Without including reservoir management and irrigation, SWAT may overestimate streamflow due to the  
275 unreasonably estimated water loss through ET.

276 Water management activities have altered natural hydrological cycling and posed challenges to reliable simulation of  
277 watershed hydrology. The YRB is a typical watershed that is regulated to support agricultural production. Maintaining  
278 sustainable water supply in basins like the YRB calls for sound understanding of hydrological impacts of management activities.  
279 Management schemes developed and evaluated in this study will be transferable and applicable to future SWAT and other  
280 watershed models applications for investigating water cycling that is influenced by reservoir operations and water withdrawal  
281 for irrigation across broader spatial scales.



## 282 4.2. Water cycling under reservoir operation scenarios

283 Reservoir operations have both direct and indirect impacts on streamflow. Water release from reservoirs directly affects the  
284 magnitude and variability of streamflow in downstream reaches. Dam and water diversion operations determine the amount  
285 and timing of water discharge to downstream river channels. As a result, reservoir operations may either attenuate flood peaks  
286 in wet seasons, or increase streamflow in dry years in compliance with minimum instream flow policies (Yoder et al., 2017).  
287 In addition, multiple hydrological processes, such as vertical flow in surface or subsurface waters, water routing, evaporation,  
288 precipitation and microclimate, are also responsive to reservoir operations (Lv et al., 2016). Our simulations suggested that  
289 reservoir operations altered both streamflow and ET in the YRB.

290 Most precipitation in the YRB occurs in winter as snowfall. Snowpack serves as a water reservoir for spring and  
291 summer streamflow. Consequently, streamflow is high in spring but low in summer. As shown in Table 1, most of the reservoirs  
292 were built to support cropland irrigation. Presence of reservoirs positively contributed to water availability in dry periods.  
293 Water storage management in reservoirs is one adaptation strategy particularly applicable to snowmelt-dominant watersheds  
294 like the YRB which experiences water scarcity during the summer irrigation season (Yoder et al., 2017), and thus alters natural  
295 flow regimes. Without representing reservoir regulations, SWAT simulations failed to reasonably reconstruct temporal  
296 variability in streamflow (R0 scenario). Results of this study indicated that reservoir algorithms based on RiverWare (R2) were  
297 relatively more realistic compared with the default reservoir operation algorithms in SWAT (R1), as evidenced by the improved  
298 model performances. Enhanced model performances in the R1 and R2 scenarios further corroborated the significant impacts  
299 of reservoir operations on seasonal patterns of streamflow (Adam et al., 2007).

300 Compared with the baseline scenario (R0), R1 and R2 simulations showed that the ET rates increased considerably  
301 from April to September due to reservoir operation. Direct evaporation from reservoirs increased under the R1 and R2 scenarios  
302 because of improved estimates of reservoir surface areas. Consideration of such an impact on ET in the R1 and R2 scenarios  
303 also contributed to enhanced model performances relative to the baseline scenario (R0).

## 304 4.3. Impacts of irrigation on water cycling

305 Water withdrawal for irrigation has increased pressures on maintaining sustainable water resources in the YRB (Malek et al.,  
306 2017). Insufficient water supply for agricultural production, drinking water supply, and environmental flows has raised  
307 concerns on the local economy and ecosystem integrity (Hillman et al., 2012). Due to the significant impacts on soil moisture  
308 and plant growth, the amount and timing of irrigation have influences on ET losses and watershed hydrology (Maier and  
309 Dietrich, 2016). As a result, the irrigation impacts on streamflow should be evaluated to provide reliable estimates of streamflow  
310 in basins like the YRB to help balance the water supplies and demands for effective water resource management.

311 As reported in previous studies, most of the water for agricultural irrigation was provided by surface water and one-  
312 third was from groundwater in the YRB (USBR, 2012). Under the R2S1 scenario, our assumption that irrigation water was



313 from the reservoirs and streams generally agreed with the actual water uses for irrigation in the basin. The less satisfactory  
314 model performances in the R2S2 scenario may stem from the unrealistic assumption of water sources, irrigation efficiencies,  
315 and return flow of irrigation. In addition, SWAT simulates streamflow based on water balance among multiple water pools,  
316 including shallow groundwater which is recharged by subsurface runoff (Shadkam et al., 2016). Under the R2S2 scenario,  
317 water withdrawal from the shallow renewable groundwater was used in our simulation. This simplification did not consider  
318 water withdrawal from deep nonrenewable aquifers. As a result, water availability based on shallow groundwater for irrigation  
319 and ground water recharge, may have been unreasonably estimated, and partially contributed to unsatisfactory model  
320 performances under this scenario (R2S2).

321 Note that model performances of the R2S1 scenario were not substantially improved relative to the R2 scenario. The  
322 irrigation operation scheme that used surface water as the single source may have introduced uncertainties to streamflow  
323 simulations, since ground water is also an important water source for irrigation, particularly in dry years in the YRB. Future  
324 simulations need to incorporate more management information about the source, amount, and timing of groundwater  
325 withdrawals into hydrologic modeling to better simulate agricultural hydrology.

326 As most reservoirs were built for irrigation in the YRB, impacts of reservoirs should be assessed jointly with the  
327 accelerating development of irrigated agriculture in the basin. Presence of reservoirs positively contributed to water availability  
328 for irrigation, particularly for dry seasons. In general, the combination of reservoir operations and irrigation have reduced  
329 streamflow in the YRB when compared with the baseline scenario (R0). This is attributable to the large amounts of water loss  
330 through ET in irrigation and additional water storage in reservoirs.

## 331 5. Conclusions

332 Reservoir operations and irrigation have substantial impacts on water cycling globally. Hydrologic simulation in the managed  
333 basins faces challenges in reliably characterizing water management activities. This study assessed the hydrological impacts of  
334 reservoir systems and irrigation practices through numerical model experiments with SWAT. Reservoir operation  
335 representations by coupling the RiverWare model and SWAT significantly improved streamflow simulations. We achieved  
336 reasonable model performances in the scenario using reservoirs and streams as the water sources for irrigation, since these  
337 assumptions are consistent with the actual irrigation practices in the basin. Model simulations suggested that reservoir  
338 operations and irrigation water withdrawal generally reduced streamflow by enhancing water loss through ET in the study area.  
339 Results of this study demonstrated importance of incorporating water management activities into hydrologic modeling.  
340 Methods and findings derived from this study are expected to help enhance future hydrologic modeling in managed watersheds  
341 with intensive reservoir and irrigation activities.



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## Tables and Figures



Table 1 Reservoir information of the YRB's five reservoirs (Locations are marked in Figure 1).

Reservoir name	River	Completion year	Dam height (m)	Active Capacity (10 <sup>6</sup> m <sup>3</sup> )	Surface area (km <sup>2</sup> )
Bumping	Bumping River	1909	19	42	5.3
Keechelus	Yakima River	1916	39	195	12.8
Kachess	Kachess River	1911	35	295	18.6
Cle Elum	Cle Elum River	1932	50	539	19.5
Rimrock	Tieton River	1924	97	244	10.2



Table 2 Dataset used in the SWAT simulations.

Data type	Spatial/Temporal Resolution/scale	Data description
Topography	30 m	Elevation
Land use	30 m	Land use classifications
Soils	1:250,000	Soil physical and chemical properties
Weather	Daily data in a one-eighth grid resolution	Precipitation, maximum and minimum air temperature, relative humidity, wind speed, and solar radiation.
Hydrological data	Daily	Streamflow
Dam	N/A	Locations, completion year, height, normal and maximal storage capacity, operating purpose, and surface area



Table 3 Parameter sensitivity analysis under various scenarios.

Variable	Description	Lower limit	Upper limit	Sensitivity rank <sup>1</sup> of five scenarios <sup>2</sup>				
				R0	R1	R2	R2S1	R2S2
SFTMP	Snowfall temperature (°C)	-20	20	2	2	2	14	2
CN2	SCS moisture condition II curve number for pervious areas	-1.2	1.2	1	1	1	1	1
SMFMX	Maximum melt rate for snow during year (occurs on summer solstice) (°C)	0	20	4	5	7	24	6
SMTMP	Snow melt base temperature (°C)	-20	20	5	3	3	18	4
CH_N2	Manning's "n" value for the main channel	-0.01	0.30	7	16	5	19	11
SMFMN	Minimum melt rate for snow during the year (occurs on winter solstice) (°C)	0	20	15	13	28	17	15
SLSUBBSN	Average slope length (m)	10	150	3	6	4	2	3
CH_N1	Manning's "n" value for the tributary channels	0.01	30	23	23	17	22	25
SOL_K	Saturated hydraulic conductivity of the first layer (mm/hr)	-0.8	0.8	8	12	8	3	7
GW_REVAP	Groundwater "revap" coefficient	0.02	0.20	14	18	12	13	14
CANMX	Maximum canopy storage (mm H <sub>2</sub> O)	0	100	26	25	19	27	28
HRU_SLP	Average slope steepness (m/m)	0	1	16	10	23	6	19
RES_K	Hydraulic conductivity of the reservoir bottom (mm/hr)	0	1	11	11	26	4	22
GW_DELAY	Groundwater delay (days)	0	500	12	19	18	25	9
EVRSV	Lake evaporation coefficient	0	1	17	8	20	12	18
TIMP	Snow pack temperature lag factor	0	1	27	27	16	28	24
ESCO	Soil evaporation compensation coefficient	0	1	24	15	24	15	23
GWQMN	Threshold water level in the shallow aquifer for the base flow (mm)	0	5000	22	20	15	16	27



Table 4 (continued)

PLAPS	Precipitation lapse rate (mm H <sub>2</sub> O/km)	-10	10	21	7	6	8	13
OV_N	Manning's "n" value for overland flow	0.01	30	9	24	22	11	8
REVAPMN	Threshold depth of water in the shallow aquifer for "revap" to occur (mm)	0	500	25	26	21	21	26
SOL_AWC	Available water capacity of the soil layer (mm H <sub>2</sub> O/mm soil)	0	1	28	14	27	23	16
NDTARGR	Number of days to reach target storage from current reservoir storage	1	200	13	22	11	9	20
ALPHA_BF	Baseflow alpha factor (days)	0	1	20	21	14	10	17
SOL_Z	Depth from soil surface to the bottom of the layer (mm)	-1	1	6	9	9	5	5
TLAPS	Temperature lapse rate (°C/km)	-10	10	19	4	13	7	21
SURLAG	Surface runoff lag time	0.05	24	18	28	25	26	10
EPCO	Plant uptake compensation factor	0	1	10	17	10	20	12

<sup>1</sup> The sensitive parameters were identified using the Global sensitivity analysis method (Abbaspour, 2007).

<sup>2</sup> R0 represents the scenario without any reservoir operations; R1 represents the scenario that used the target release approach for the simulation of reservoir outflow in the SWAT model; R2 represents the scenario that used the output of RiverWare model as the daily outflow of the five reservoirs in the SWAT model; R2S1 represents the scenario with irrigation operation that withdraws water from the reservoirs and streams based on the R2 scenario; R2S2 represents the scenario using groundwater as the water source for irrigation based on the R2 scenario.

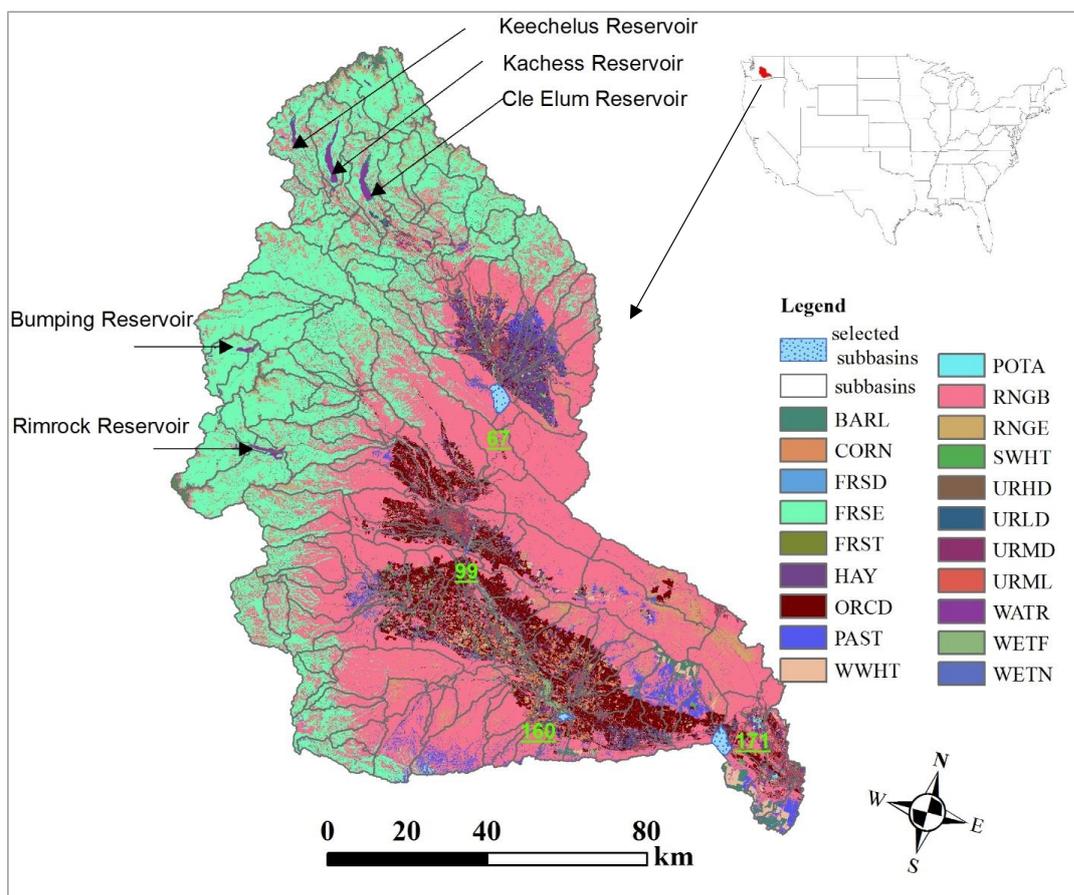


Figure 1 Location and land use of the Yakima River Basin (67, 99, 160, and 171 are subbasin which were used for streamflow calibration and validation. BARL: Spring Barley; CORN: Corn; FRSD: Deciduous forest; FRSE: Evergreen forest; FRST: Mixed forest; HAY: Hay; ORCD: Orchard; PAST: Pasture; POTA: Potato; RNGB: Range-bush; RNGE: Range-grasses; SWHT: Spring wheat; URHD: Residential-high Density; URLD: Residential-Low Density; URMD: Residential-Medium Density; WATER: Water; WETF: Wetland-forested; WETN: Wetland-non-forested; WWHT: Winter wheat).

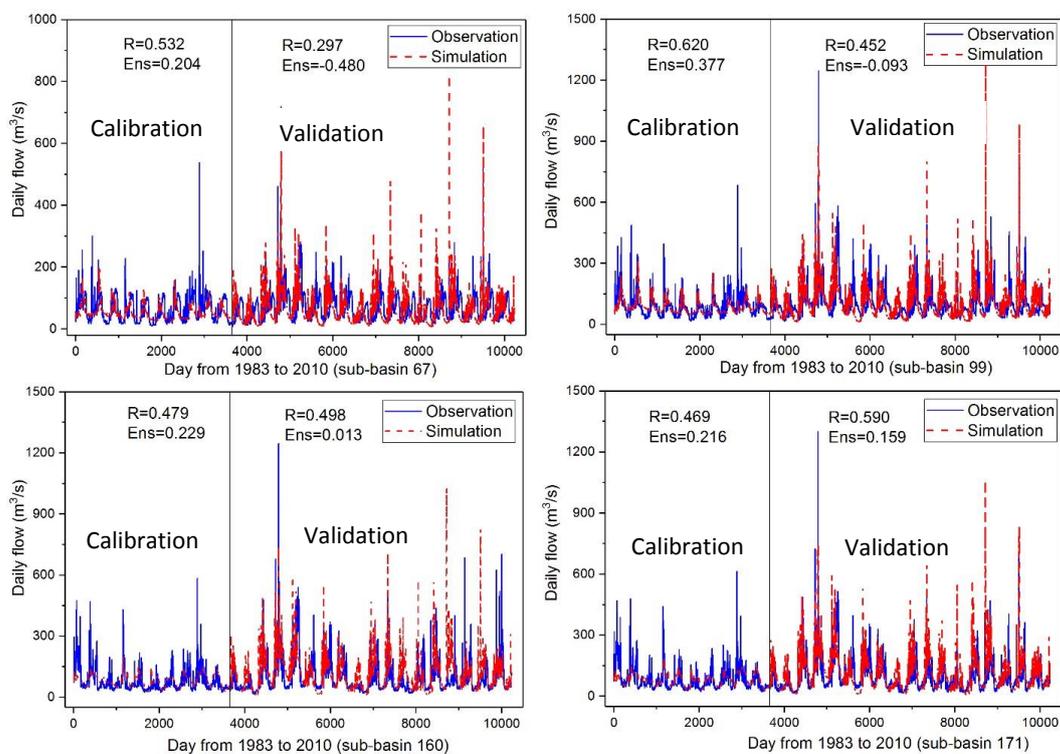


Figure 2 Calibration and validation results in four subbasins under the R0 scenario (baseline simulation does consider management activities).

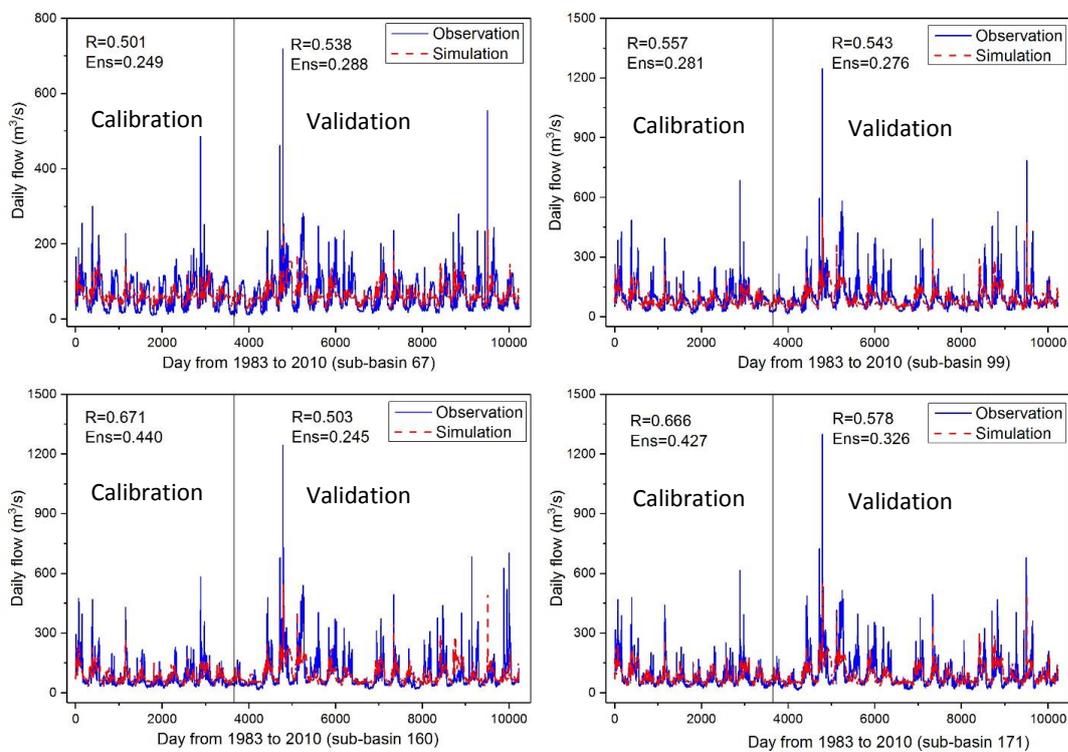


Figure 3 Calibration and validation results under the R1 scenario (Default SWAT schemes for reservoir operations)

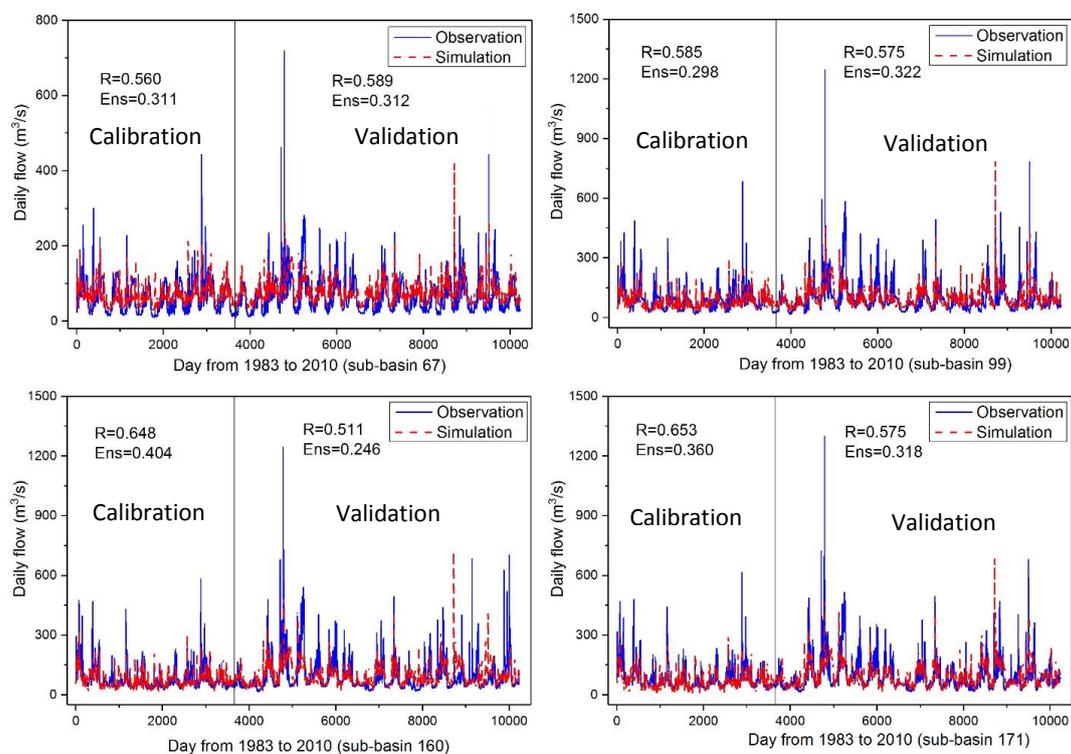


Figure 4 Calibration and validation results under the R2 scenario (RiverWare for reservoir operations).

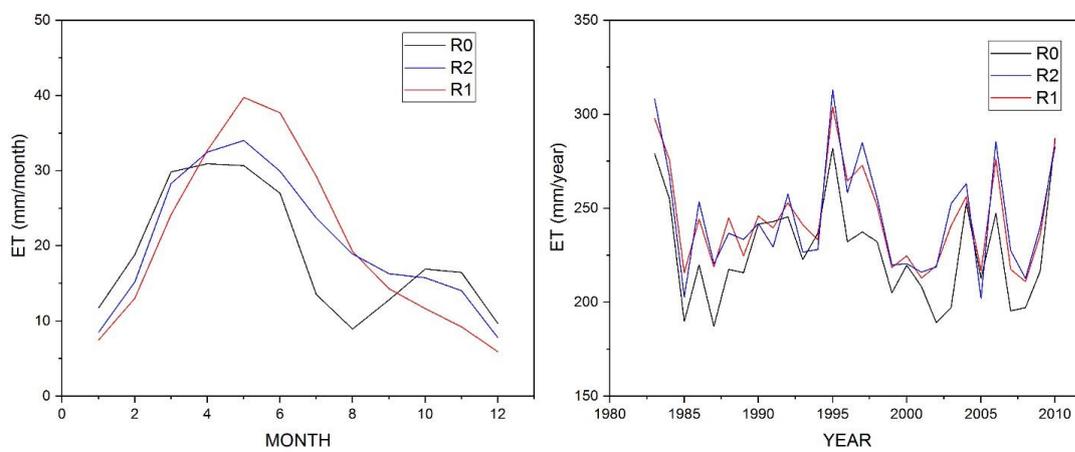


Figure 5 Monthly and annual ET simulated under three scenarios only consider reservoir operations (R0, R1, and R2).

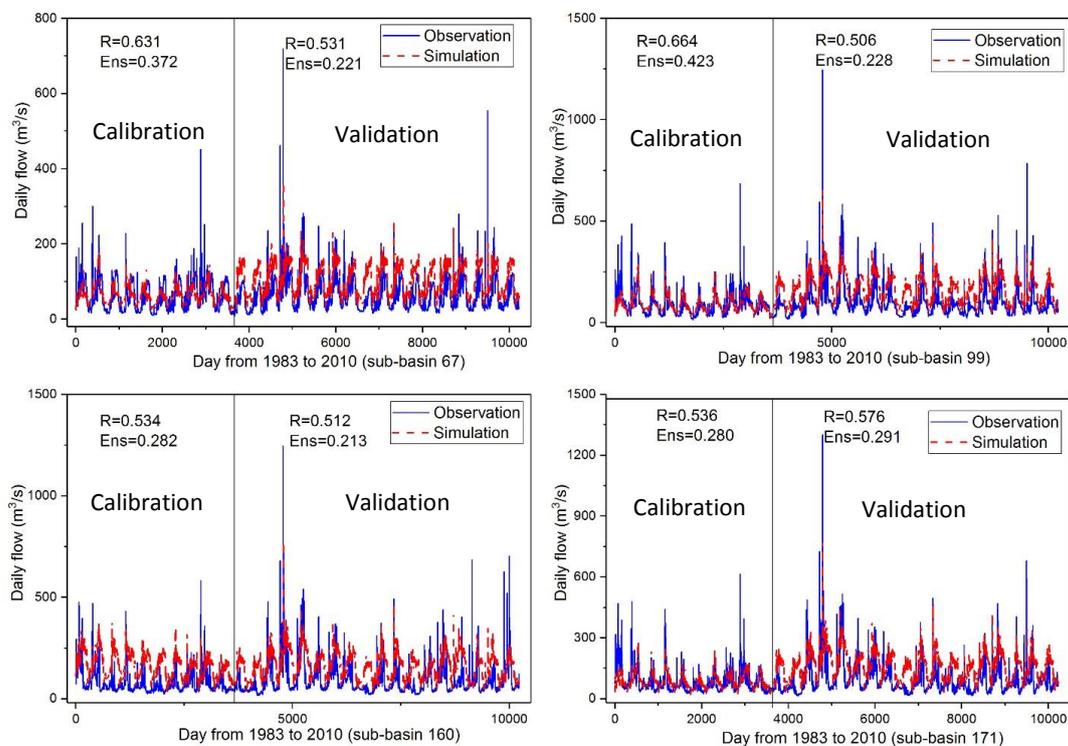


Figure 6 Calibration and validation results under the R2S1 scenario (Default SWAT algorithms for reservoir operation and surface water as the water source for irrigation)

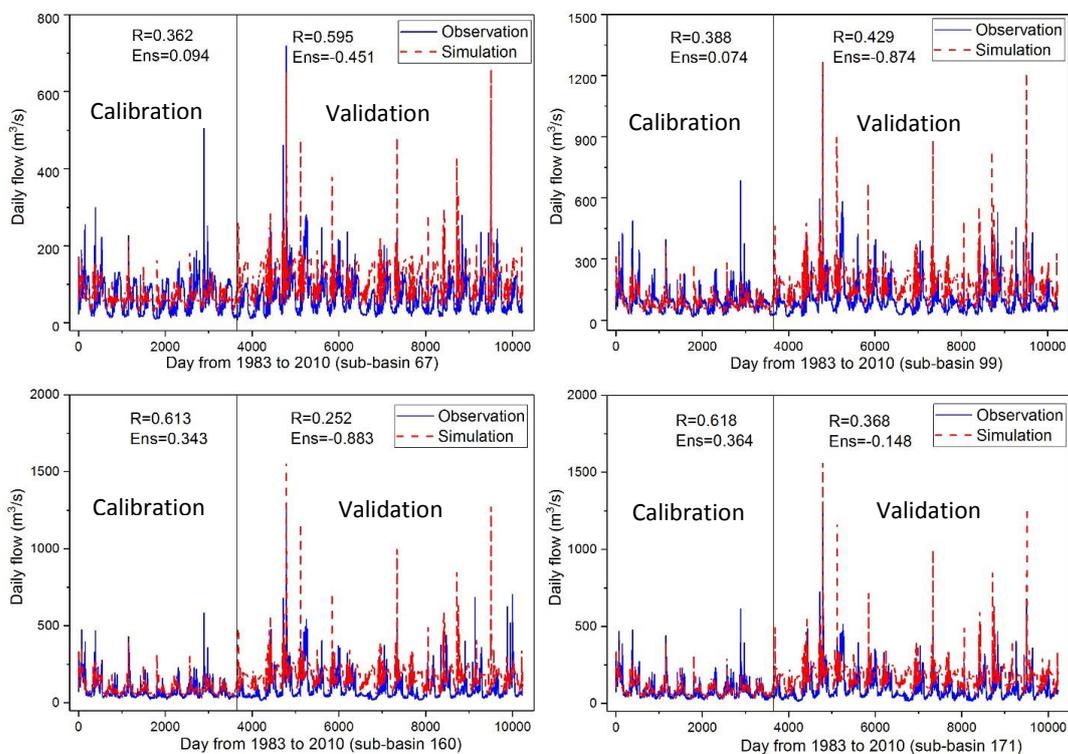


Figure 7 Calibration and validation results under the R2S2 scenario (RiverWare for reservoir operation and ground water as the water source for irrigation)

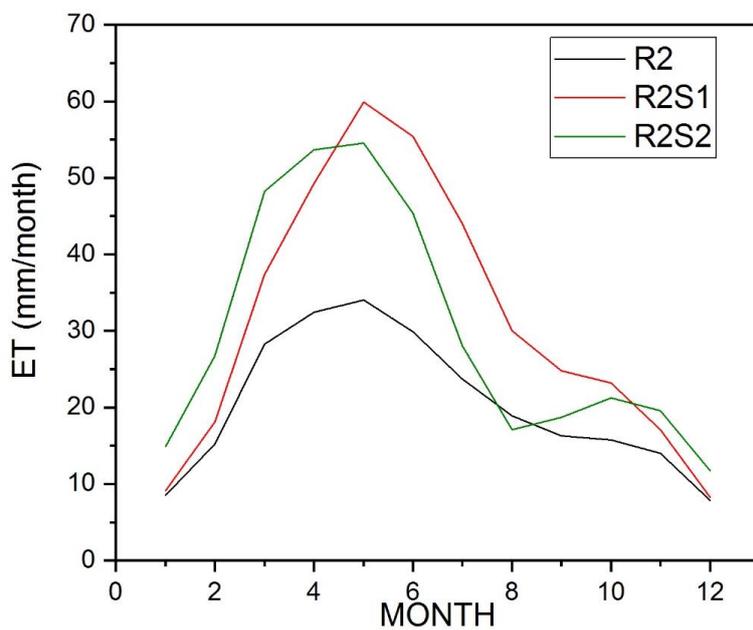


Figure 8 Monthly ET simulated under the irrigation operation scenarios (R2S1 and R2S2) relative to the reservoir operation-only scenario (R2).

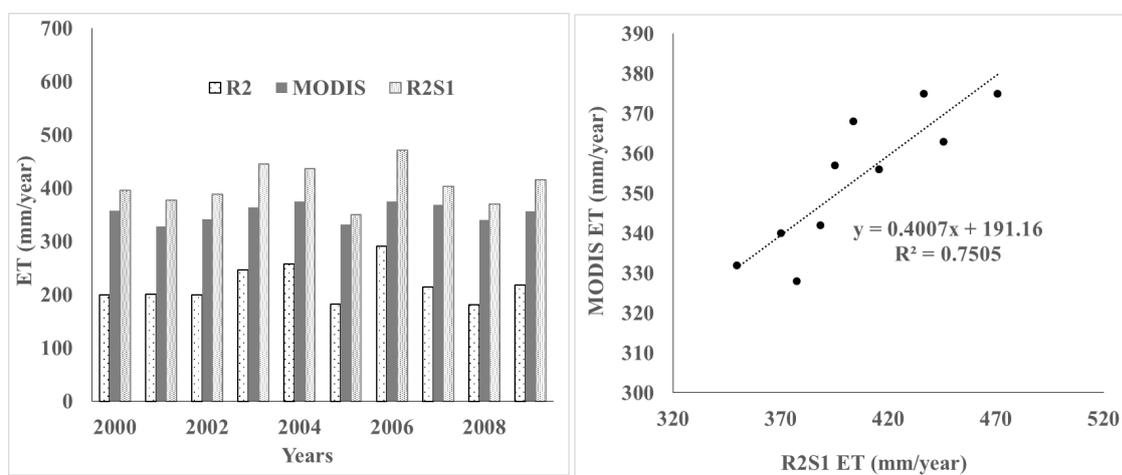


Figure 9 Comparison of ET simulations for cropland during 2000-2009 under the R2 and R2S1scenarios with the Moderate Resolution Imaging Spectroradiometer (MODIS) ET.