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2 **Assessing the Impact of Groundwater Saturation Excess Runoff on Hydrologic**  
3 **Features and Processes in a Watershed Modeling Setting**  
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13 **Key Points:**

- 14 • We evaluate the effect of groundwater saturation excess flow on hydrologic features  
15 (wetlands) and temporal patterns of watershed water yield.
- 16 • We demonstrate relationships between precipitation, recharge, saturation excess flow,  
17 and streamflow before and during storm events.
- 18 • We compare locations of consistent groundwater saturation excess flow (simulated) with  
19 mapped wetlands.  
20  
21

## 22 **Abstract**

23 Groundwater saturation excess flow can be a major surface runoff mechanism in humid regions,  
24 characterized by shallow aquifers and soil profiles that become saturated during wet periods or  
25 intense storm events. This process often plays an important role in the creation and maintenance  
26 of groundwater-dependent ecosystems and the overall water yield of a watershed. In this paper,  
27 we examine the process of groundwater saturation excess flow and assess its influence on  
28 hydrologic features (wetlands) and temporal patterns of watershed water yield. We do this by  
29 applying a surface-subsurface hydrologic model (SWAT+ with the physically based spatially  
30 distributed *gwflo*w module for groundwater storage and flow) to the Little River Watershed,  
31 Georgia, USA, which has a high baseflow fraction and contains numerous wetlands, and for  
32 which groundwater saturation excess flow has been noted in past studies. The model is calibrated  
33 and tested against measured streamflow and groundwater head for the period 2000-2015, with  
34 and without groundwater saturation excess flow included in the *gwflo*w module. Model results  
35 indicate that including groundwater saturation excess flow improves hydrologic estimation, and  
36 demonstrate connections between precipitation, recharge, saturation excess flow, and streamflow  
37 before and during storm events. Finally, we compare locations of consistent groundwater  
38 saturation excess flow (simulated) with mapped wetlands, demonstrating that the model can be  
39 used to explore impacts of system changes (land use, climate, management) on wetland  
40 development and maintenance.

41

## 42 **Plain Language Summary**

43 In humid regions with high rainfall rates, groundwater can rise to the surface and discharge to the  
44 land surface and nearby wetlands and streams. This process is valuable for sustaining ecosystems  
45 that depend on groundwater and for general streamflow generation. In this study, we examine  
46 how this process of groundwater saturation excess flow influences hydrology and wetland  
47 development in the Little River Watershed in Georgia, USA. We employ a computer model to  
48 investigate this process and how it effects surface runoff, streamflow rates, and the presence of  
49 wetlands. The model is tested by comparing output to measured streamflow and groundwater  
50 levels during the 2000-2015 period. The results reveal that including groundwater saturation  
51 excess flow in the model helps us better understand water movement in the watershed,  
52 particularly during and after rain storms. We also look at how this groundwater process connects  
53 with water recharge, streamflow, and rainfall. Finally, we compare areas of consistent  
54 groundwater saturation excess flow to areas of mapped wetlands. This work enables us to  
55 understand how groundwater saturation excess flow effects the environment, and provides a  
56 modeling tool for studying relationships between climate, management, and groundwater-  
57 dependent ecosystems.

58

## 59 **1 Introduction**

60 Accurate prediction of hydrologic processes is essential for effective watershed  
61 management, given the complexity of watershed systems influenced by various factors including  
62 land use, climate, and anthropogenic impacts (e.g., Abbas and Xuan, 2019; Du et al., 2022;  
63 Gyamfi et al., 2016). Hydrologic models assist water resource managers in comprehending the  
64 impacts of natural and anthropogenic factors on, for example, hydrological features, optimizing

65 reservoir operations, and predicting changes in water resources (Wu and Xu, 2007; Mengistu et  
66 al., 2021). Proper representation of hydrologic processes, including surface runoff, groundwater  
67 recharge, and groundwater seepage, is vital for simulating streamflow and groundwater head in a  
68 distributed hydrologic model in an accurate fashion. Of these processes, groundwater seepage is  
69 often neglected when applying watershed models.

70 Groundwater seepage at the land surface, often referred to as “groundwater discharge” or  
71 “groundwater saturation excess runoff”, occurs when the water table intersects local topography  
72 (Bear, 1972; Deitchman and Loheide, 2009; Rath et al., 2023). This seepage water can play an  
73 important role in 1) the generation and maintenance of groundwater-dependent ecosystems  
74 (GDEs) such as wetlands and fens (i.e., wetlands reliant on groundwater discharge) (Shedlock et  
75 al., 1993; Hunt et al., 1996; Winter, 1999; Batelaan et al., 2003; Dekker et al., 2005; Feinstein et  
76 al., 2019; Lamber et al., 2022); and 2) the generation of surface water flow and overall water  
77 yield of a watershed, particularly during storm events that cause high groundwater levels (Beven,  
78 1989; Ruprecht and Schofield, 1989; Bari et al., 1996; Kazmierczak et al., 2016). For wetlands  
79 and fens, groundwater provides essential supplies of water, nutrients, and heat (Kløve et al.,  
80 2011). These internal (GDEs) and response (streamflow generation) features of a watershed  
81 system often are threatened by land use activities, climate change, and groundwater extraction  
82 (Dekker et al., 2005; Brown et al., 2010; Kløve et al., 2011; Aldous and Bach, 2014).

83 Providing informed water management in watersheds wherein groundwater seepage plays  
84 a significant role in either GDE development or streamflow generation requires the establishment  
85 of dynamics relationships between watershed features, weather patterns, groundwater storage  
86 and flow, wetland locations, and streams (Batelaan et al., 2003; Brown et al., 2010; Kløve et al.,  
87 2011). These relationships, however, have not yet been defined on the watershed scale. Many  
88 studies have investigated and quantified the generation of groundwater seepage along hillslopes  
89 (e.g., Cloke et al., 2003; Beaugendre et al., 2006; Scudeler et al., 2017; Bizhanimanzar et al.,  
90 2019; Rath et al., 2023), although at small spatial scales and under controlled numerical  
91 experiments. In connection with GDEs, Sampath et al. (2016) and Feinstein et al. (2019) used  
92 physically based groundwater models (MODFLOW) to simulate interactions between  
93 groundwater and fens. However, they used steady-state conditions and neglected other land  
94 surface features (e.g., evapotranspiration). For streamflow generation, semi-distributed models  
95 such as SWAT (Easton et al., 2008; White et al., 2011; Hoang et al., 2017; Steenhuis et al., 2019)  
96 have implemented saturation excess routines, but only in terms of soil water storage and not in  
97 relation to groundwater storage or a rising water table within a physically based framework.  
98 Other modeling studies in low-gradient watersheds (Bosch et al., 2010; Rathjens et al., 2015)  
99 have noted that correct simulation of daily or monthly streamflow requires the inclusion of  
100 interactions between groundwater and the soil surface.

101 In general, there is a lack of information on large-scale spatial and temporal relationships  
102 between groundwater saturation excess flow, GDEs, and streamflow generation, that include the  
103 influence of watershed inputs (weather, land use) and hydrologic processes. These relationships  
104 and influences should be quantified in a coupled surface-subsurface manner to inform planning  
105 and protection strategies for wetlands, fens, and general water supply. Also, to our knowledge,  
106 no studies have quantified the effect of groundwater saturation excess runoff on hydrological  
107 processes in a large-scale watershed system.

108 The objective of this paper is to quantify the impact of groundwater saturation excess  
109 flow on hydrologic features and hydrologic processes in a humid, regional watershed.

110 Specifically, we aim to quantify the impact of groundwater saturation excess flow on hydrologic  
111 fluxes, streamflow generation, and wetland development and location. For the latter, we employ  
112 a method similar to Feinstein et al. (2019), in which they compared locations and rates of model-  
113 simulated groundwater seepage to known locations of fens. In addition, we relate the frequency  
114 of groundwater saturation excess flow (i.e., the fraction of time that excess flow occurs) to  
115 known wetland locations.

116 To achieve these objectives, we apply the SWAT+ hydrologic model, amended with the  
117 *gflow* module (Bailey et al., 2020), to the Little River Watershed (2,309 km<sup>2</sup>), Upper Suwannee  
118 River Basin, south-central Georgia, USA. This watershed has been studied extensively as an  
119 experimental watershed (Bosch et al., 2007), with results suggesting significant contribution of  
120 groundwater saturation excess runoff to streamflow (Bosch et al., 1996; Inamdar et al., 1999;  
121 Bosch et al., 2017). The *gflow* module simulates groundwater storage and head in a process-  
122 based manner using a collection of grid cells connected to SWAT+ hydrologic response units  
123 and channels, providing a powerful tool for simulating water movement in a surface-subsurface  
124 system. The combined model is run on a daily time step from 2000 to 2015, and tested against  
125 streamflow at multiple gaging sites, groundwater head at multiple monitoring wells, and the  
126 location of established wetlands. The model is run with and without the groundwater saturation  
127 excess flow mechanism active in the modeling code, to determine its effect on watershed  
128 hydrology.

129 In addition to hydrologic insights gained from applying the modeling method, the model  
130 can, in general, be used as a tool to explore the impact of system changes (climate, land use,  
131 management) on wetland development, wetland maintenance, water supply, and conjunctive  
132 supply of surface water and groundwater. We note that the basic SWAT+*gflow* model set-up  
133 for the Little River Watershed has previously been outlined in Bailey et al. (2023), in which the  
134 SWAT+*gflow* set-up was demonstrated for several watersheds across the United States. This  
135 work builds on previous application by adding model calibration and testing, and a detailed  
136 exploration of the effect of saturation excess flow on watershed hydrology.

137

## 138 **2 Materials and Methods**

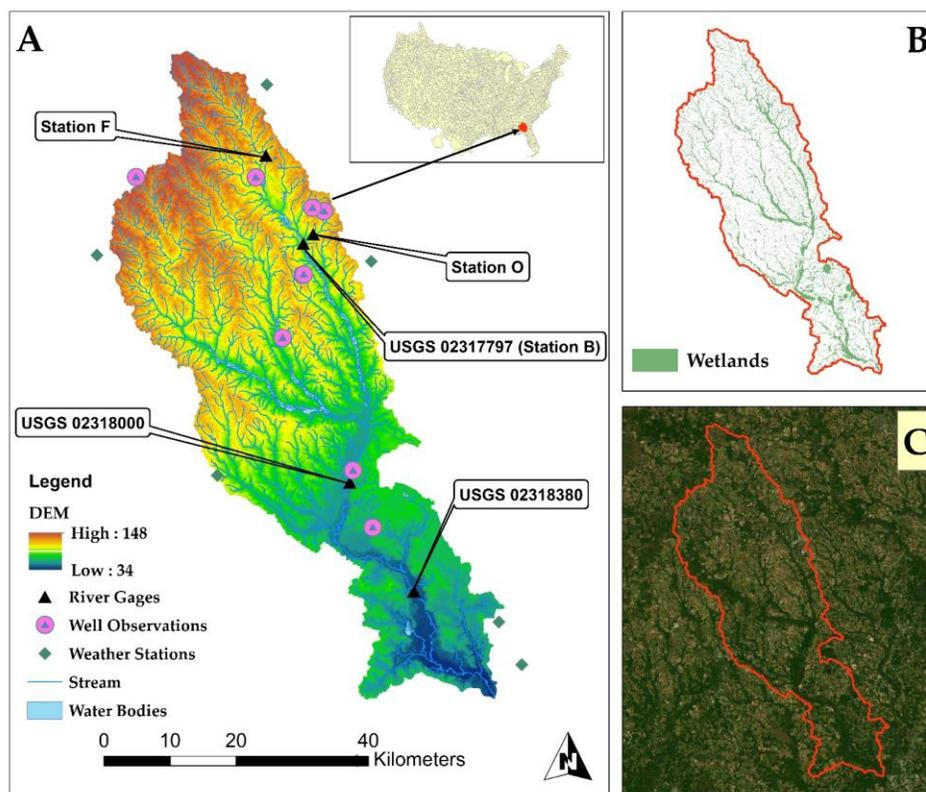
### 139 **2.1 Study Area**

140 We apply modeling methods to the Little River Watershed (2,309 km<sup>2</sup>), Upper Suwannee  
141 River Basin, south-central Georgia. Figure 1 shows a map of elevation, locations of USGS river  
142 gage stations used for model testing, locations of USGS monitoring wells used for testing,  
143 locations of weather stations, water bodies, and wetland delineation. Wetland locations are  
144 provided by the National Wetlands Inventory of the U.S. Fish and Wildlife Service (accessed  
145 October 2023). This watershed was selected due to reported hydraulic connection between  
146 streams and the shallow aquifer and the presence of groundwater saturation excess flow (Bosch  
147 et al., 2017).

148 The study area has a humid subtropical climate, with mild winters and hot humid  
149 summers. In the summer months, there is typically a significant increase of streamflow due to  
150 high-intensity thunderstorms. The average yearly precipitation is 1,287 mm, while the mean  
151 temperature for the year is 19 °C. Evapotranspiration accounts for approximately 70% of annual

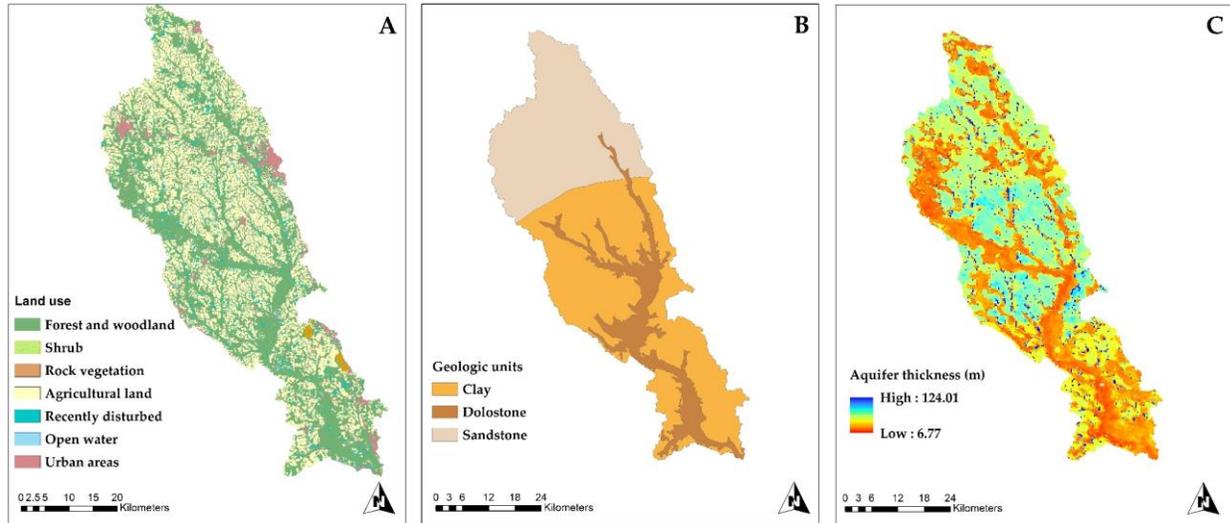
152 precipitation (Sheridan, 1997). Elevations in the watershed range from 34 to 148 meters (Figure  
 153 1A).

154 Land use (Figure 2A) comprises forest (45%), mostly evergreens and hardwoods in  
 155 riparian zones and pine trees in highland locations; agriculture (41%) major row crops of cotton  
 156 and peanuts; open water (1.5%); and urban areas (12.5%). Soils in the study area are clay,  
 157 dolostone, and sandstone (Figure 2B), which are underlain by the relatively watertight  
 158 Hawthorne formation, restricting the movement of groundwater to lower geological layers  
 159 (Stringfield, 1966), and indicated as the bedrock. The shallow nature of the aquifer and the  
 160 impermeable Hawthorne formation leads to reported groundwater saturation excess flow (Bosch  
 161 et al., 2010; Bosch et al., 2017), principally in the riparian areas. Estimated aquifer thickness (m)  
 162 (Shangguan et al., 2017) (vertical distance between ground surface and bedrock) is presented in  
 163 Figure 2C. These data illustrate distinct areas of shallow thickness along the riparian corridor.



164  
 165 **Figure 1.** Maps of the Little River Watershed, showing (A) elevation, stream gages, and  
 166 monitoring wells; (B) wetland delineations (USFWS, 2018); and (C) satellite image, showing  
 167 riparian areas along the complex channel network.

168



169

170 **Figure 2.** Land use, geologic, and aquifer thickness (m) of the study watershed.

171

172 

## 2.2 Hydrologic Model of the Little River Watershed

173 In this section we outline the hydrologic modeling tool used to address the study  
 174 objectives. The base model is SWAT+, amended to include the *gwf* module which simulates  
 175 groundwater saturated excess flow in a physically based spatially distributed manner.

176 

### 2.2.1 SWAT+ model

177 The Soil and Water Assessment Tool (SWAT; Arnold et al., 1998) is a process-based,  
 178 semi-distributed, continuous-time, basin-scale hydrologic model that has been widely used over  
 179 the last 25 years (Bieger et al., 2016). The model is often used to predict the enduring effects of  
 180 land use practices, climate change impact, and non-point source pollution on water resources,  
 181 sediment, and agricultural chemical yields within river basins. The model segments the  
 182 watershed into hydrological response units (HRUs) which are computational entities defined by  
 183 specific values of slope, land use, and soil properties (Neitsch et al. in 2011). Water balances are  
 184 calculated on a daily time step for the soil profile (HRU), aquifer (HRU), channel (subbasin), and  
 185 reservoirs. Recently, a restructured version of the SWAT model called SWAT+ was introduced  
 186 by Bieger et al. (2016). The new version offers enhanced flexibility in linking hydrologic  
 187 components within a watershed system. Unlike the original SWAT model, which is restricted to  
 188 a single stream channel per subbasin, SWAT+ allows the simulation of sediment, water, and  
 189 nutrient movement across numerous channels throughout catchments of varying spatial scale.  
 190 Within this framework, hydrologic objects comprise reservoirs, stream channels, ponds,  
 191 wetlands, HRUs, and aquifers.

192 For this study, the SWAT+ model is constructed for the 2000-2015 period using the  
 193 datasets listed in Table 1. This model is part of the National Agroecosystem Model (NAM)  
 194 initiative (Arnold et al., 2020; White et al., 2022), a nationwide endeavor for evaluating  
 195 conservation policies. Hydrologic features of the SWAT+ model is presented in Table 2,  
 196 including 1,816 channels (NHD+ stream segments) and 4,844 HRUs.

197

198 **Table 1.** Datasets used to construct the standalone SWAT+ models and the gwflow inputs  
 199 (Bailey et al., 2023).

	Dataset	Resolution (m)	Source
<b>SWAT+ model</b>	Land use, Land cover	30	U.S. Geological Survey, National Land Cover Data
	Field boundaries		Yan and Roy (2016)
	Topographic slope map	10	USGS National Elevation Dataset (Gesch et al., 2018)
	Weather		Global historical climatology network; PRISM
	Soil boundaries and properties	10	Soil Survey Staff (2014)
	Stream segments		NHD+ Moore and Dewald (2016)
	Crop rotation		USDA–NASS, CDL
	Lakes and reservoirs		Moore and Dewald (2016)
	Water use		Dieter et al. (2018)
	Discharge from facilities		Skinner and Maupin (2019)
<b>Gwflow</b>	Groundwater head	Vector Points	Bailey and Alderfer (2022)
	Aquifer thickness	250	Shangguan et al. (2017)
	Tile drainage	30	Valayamkunnath et al. (2020)
	Geologic units	Vector Polygons	Horton et al. (2017)

200

201 **Table 2.** Characteristics for the study watershed.

Watershed	State	HUC2 Region	HUC8	# Channels	# HRU	<i>mm</i>	<i>km<sup>2</sup></i>	<i>gwflow</i> grid		
						Annual Precip.	Area	Rows	Cols	Cell size (m)
Little River	GA	South Atlantic-Gulf	03110204	1816	4844	1287	2309	197	120	500

202

### 203 2.2.2 *gwflow* module

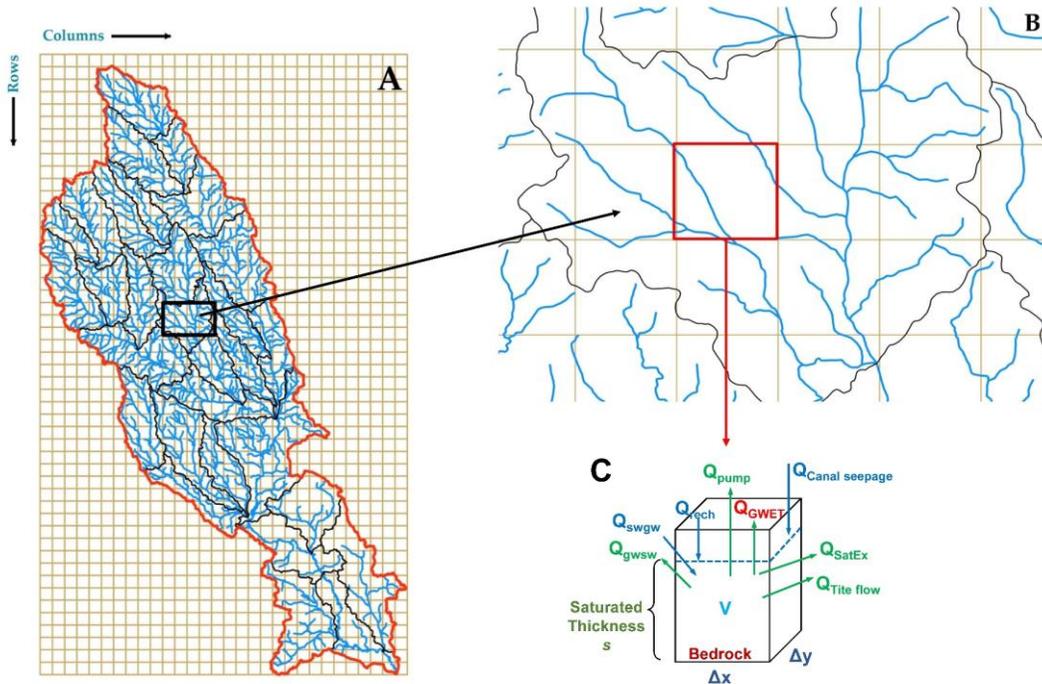
204 We modified the base SWAT+ model to include the *gwflow* module for groundwater  
 205 modeling and groundwater interaction with hydrologic objects within the watershed. The *gwflow*  
 206 module, developed by Bailey et al. (2020), has been included into SWAT+ as an optional  
 207 subroutine and serves as alternative to the original groundwater module. This addition enables  
 208 physically based spatially distributed modeling of groundwater storage and flow in unconfined  
 209 aquifer systems. The *gwflow* module employs a collection of grid cells, also known as aquifer  
 210 control volumes, to model the storage and movement of groundwater (Figure 3). The thickness  
 211 of each cell is equivalent to the thickness of the aquifer, extending from the ground surface to the  
 212 bedrock. The user specifies the cell size.

213 Groundwater storage volume  $V(\text{m}^3)$  is updated during each daily time step (time  $n$  to time  
 214  $n + 1$ ) for each grid cell ( $i, j$ ) using the following groundwater balance equation:

215

$$V_{i,j}^{n+1} = V_{i,j}^n + (\text{sources}_{i,j}^n - \text{sinks}_{i,j}^n \mp \text{lateral flow}_{i,j}^n)(t^{n+1} - t^n) \quad (1)$$

216 where sources include stream seepage, lake seepage, and recharge, and sinks include  
 217 groundwater discharge to streams, saturation excess flow to streams, tile drainage outflow to  
 218 streams, pumping, groundwater discharge to lakes, and groundwater ET. Recharge is provided  
 219 by soil percolation from HRUs, using a geographic intersection between grid cells and HRUs.  
 220 Groundwater exchange with channels and lakes/reservoirs is performed for grid cells that  
 221 intersect channel and reservoir objects. The calculation of tile drainage outflow, groundwater-  
 222 stream exchange, and groundwater-lake exchange involves the use of Darcy's Law, which  
 223 utilizes many object qualities such as streambed conductivity, stream width, and stream length.  
 224 Lateral flow is Darcy flow between neighboring cells, which uses hydraulic conductivity ( $K$ ) and  
 225 saturated thickness specific to each cell and cell-to-cell gradients in hydraulic head  $h$ .



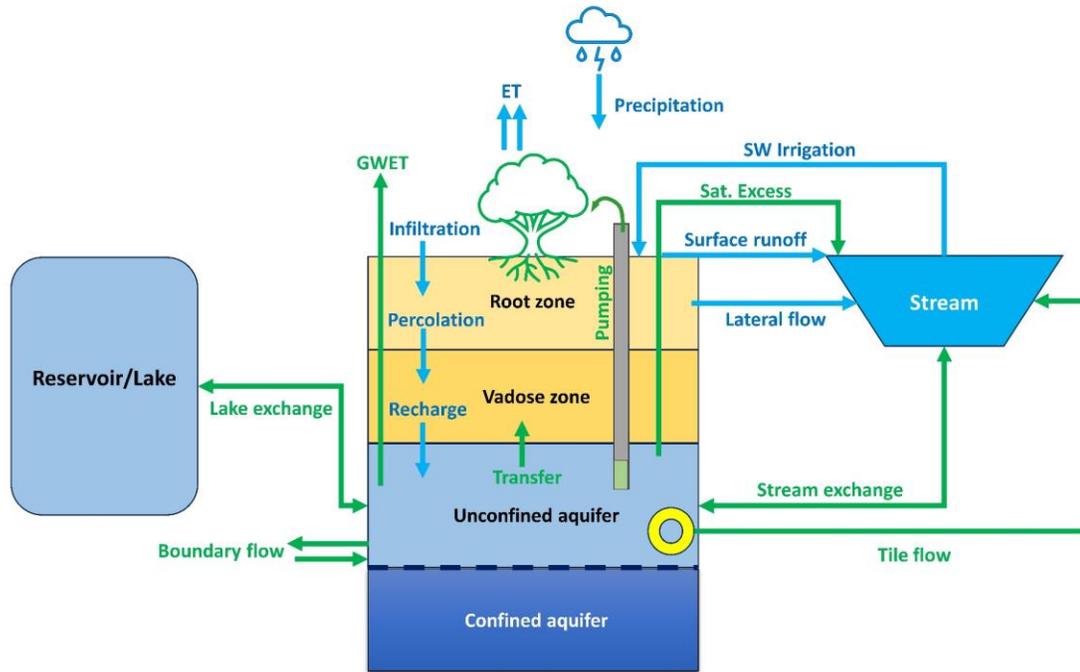
226  
 227 **Figure 3.** Spatial layout and calculation method of the *gwflow* module within the Little River  
 228 Watershed, showing (A) grid cells, watershed boundary (red line), stream channels (blue lines),  
 229 and subbasins (black lines) for the study watershed; (B) close-up of grid and channels; and (C)  
 230 representing the hydrologic fluxes for each individual cell.

231  
 232 After determining the new volume  $V_{i,j}^n$ , groundwater head is calculated using the specific  
 233 yield ( $S_y$ ) of the cell. By incorporating the *gwflow* module, SWAT+ simulates the dynamics of  
 234 soil, land surface, and channel processes, while the *gwflow* module specifically models  
 235 subsurface processes (Figure 4) and aquifer-object interactions. Groundwater-induced saturation  
 236 excess runoff is simulated when the groundwater head  $h$ , rises above the elevation of the ground  
 237 surface. This phenomenon often happens after precipitation events that induce a fast rise in the  
 238 water table. The volumetric flux ( $\text{m}^3/\text{day}$ ) of groundwater excess flow is computed as:

239

$$Q_{satex} = (h_{i,j} - Z_{surf_{i,j}})(\Delta x \Delta y) S_{y_{i,j}} \quad (2)$$

240 where  $h$  is groundwater head for cell  $(i,j)$ ,  $Z_{surf}$  is the ground surface elevation for cell  $(i,j)$ , and  
 241  $S_y$  is specific yield ( $\text{m}^3$  water per  $\text{m}^3$  of bulk material). The volumetric flow rate  $Q_{satex}$  is extracted  
 242 from the cell and then transferred to the nearest stream channel, within the same day.



243  
 244 **Figure 4.** Schematics representation of the hydrologic processes in a typical watershed stream-  
 245 aquifer system with Saturation Excess Runoff; showing main hydrologic elements and  
 246 hydrologic processes for SWAT+ and *gwflow*. Green arrows demonstrate fluxes that are  
 247 simulated by *gwflow*, blue arrows represent fluxes that are simulated by SWAT+.

248  
 249 We note that, under real-world conditions, shallow groundwater enters the soil profile  
 250 before discharging to the land surface. Within the modeling framework used in this study,  
 251 however, groundwater in *gwflow* cells is not connected hydrologically to soil water in the SWAT  
 252 HRUs. While this linkage between the aquifer and the soil profile is included in the *gwflow*  
 253 module (Yimer et al., 2023), we neglect it here to provide detailed analysis of where  
 254 groundwater discharge occurs. At a regional scale, we assume that groundwater discharge fluxes  
 255 will be the same, whether simulated by the *gwflow* subroutine, on a cell basis, or by the soil  
 256 saturation excess routine of SWAT+, on an HRU basis. We selected a cell size of 500 m (Table  
 257 2). The datasets used for populating the cell values in *gwflow* consist of  $S_y$ ,  $K$ , aquifer thickness,  
 258 and initial groundwater head (see Table 1). For this study, initial head for each cell was estimated  
 259 by spatial interpolating between USGS monitoring wells (see Figure 1) using values from the  
 260 year 2000. The identification of cells involved in groundwater-lake exchange and groundwater-  
 261 channel exchange is achieved by intersecting cells with NHD+ channels and water bodies (Table  
 262 1).

## 263 2.2.3 Model Calibration and Testing

264 The SWAT+*gwflow* model has a warm-up period of 2000–2001, a calibration period of  
 265 2002–2008, and a testing period of 2009–2015. The SWAT+*gwflow* model is calibrated and  
 266 tested using the Parameter Estimation Software (PEST; Doherty, 2020). In this work, the  
 267 objective function (OF) includes monthly streamflow ( $\text{m}^3/\text{sec}$ ) obtained from USGS stream gage  
 268 stations at one river gage location for calibration, and four other gages sites for testing, and  
 269 average annual groundwater head (m) collected from USGS monitoring wells at ten different  
 270 locations. The impact of each of these sites on the composite OF was modified by adjusting the  
 271 weights assigned to the residuals to ensure that each site has a comparable level of importance  
 272 and relevance in selecting the best parameter values. Monthly simulated streamflow is evaluated  
 273 using Nash–Sutcliffe Efficiency Index (NSE), Kling–Gupta Efficiency Index (KGE), percent  
 274 bias (PBIAS), and coefficient of determination ( $R^2$ ). Annual simulated groundwater head is  
 275 evaluated using mean absolute error (MAE).

276 The parameters to be modified by PEST (Table 3) were selected based on the SWAT  
 277 model documentation and literature (e.g., Koo et al., 2020; Arnold et al., 2013), and focus on  
 278 surface runoff, evaporation, soil properties, groundwater processes, lateral flow, time of  
 279 concentration, and channel flow processes.

## 280 2.2.4 Addressing Study Objectives

281 We use the calibrated and tested model to quantify the impact of groundwater saturation  
 282 excess flow on hydrologic fluxes, streamflow generation, and wetland development and location.  
 283 To quantify influence on hydrologic fluxes, we compare hydrologic results (runoff, recharge, soil  
 284 lateral flow, groundwater-channel interactions) from calibrated simulations with and without  
 285 groundwater saturation excess flow enabled. For streamflow generation, we compute the fraction  
 286 of streamflow that originates from groundwater saturation excess flow, aggregated and  
 287 temporally. For wetland development and location, we compare volumetric fluxes of  
 288 groundwater saturation excess flow spatially to wetland areas, and the frequency of saturation for  
 289 each cell in the *gwflow* grid as an indicator of wetland persistence.

290

291 **Table 3.** Description and hydrological processes of the selected parameters for the  
 292 SWAT+*gwflow* model for Little Watershed.

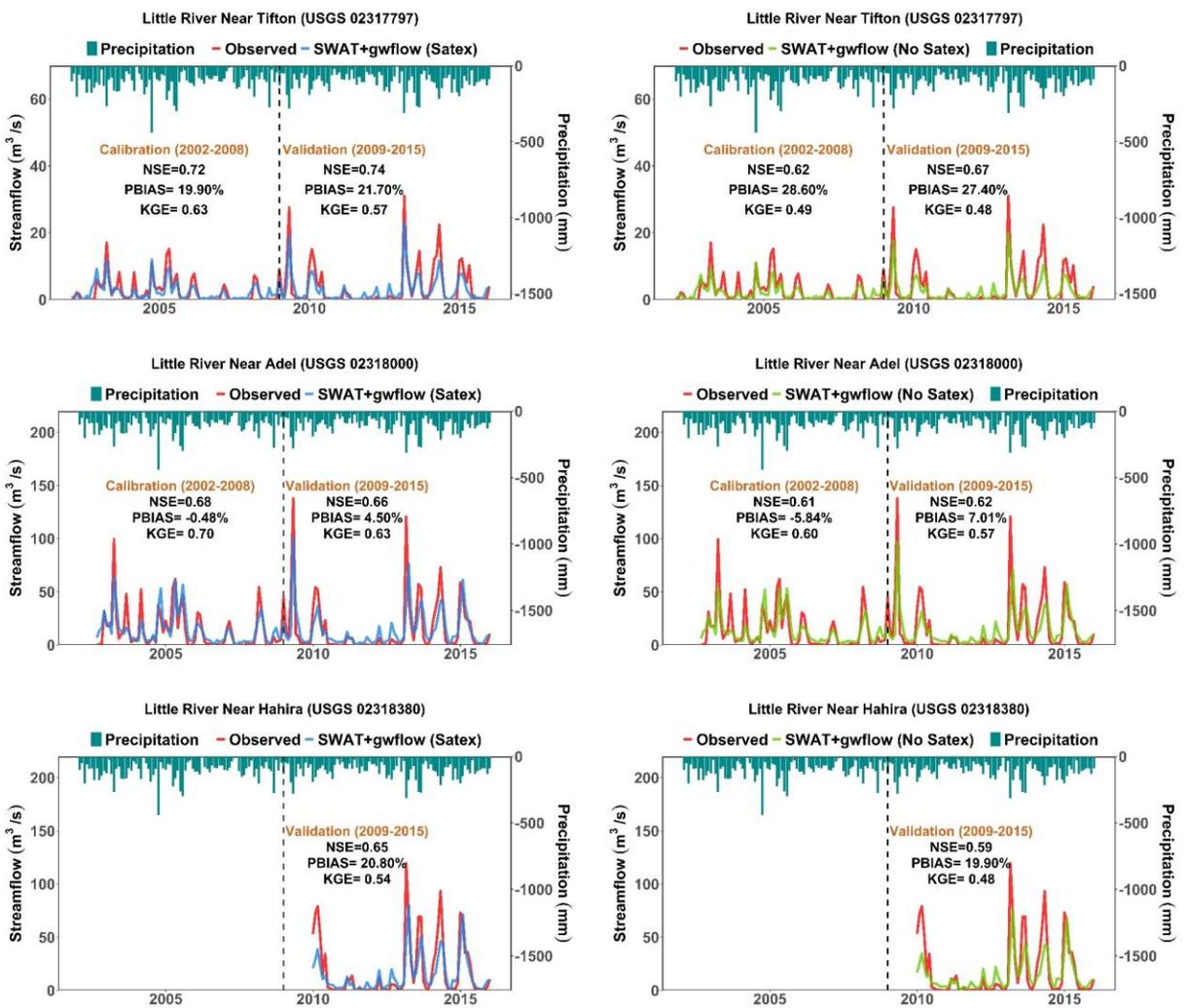
Parameters	Description	Hydrologic Processes
<b>CN2 #</b>	SCS runoff curve number for moisture condition II	Surface runoff processes (cn)
<b>surq_lag</b>	Surface runoff lag time (days)	Time of concentration processes (par)
<b>esco #</b>	Soil evaporation compensation factor	Potential and actual evapotranspiration processes, and percolation (hydro)
<b>epco #</b>	Plant uptake compensation factor	
<b>perco #</b>	Percolation coefficient	
<b>rech_del</b>	Recharge delay (days)	Groundwater flow processes ( <i>gwflow</i> )
<b>Kaqu #</b>	Aquifer hydraulic conductivity for a specific zone ( $\text{m}/\text{day}$ ) for $i^{\text{th}}$ zone	
<b>Syaqu #</b>	Aquifer specific yield for a specific zone for $i^{\text{th}}$ zone	
<b>bed_k</b>	Streambed hydraulic conductivity ( $\text{m}/\text{day}$ )	
<b>bed_thick</b>	Streambed thickness (m)	
<b>bed_depth</b>	River depth (m)	
<b>tile_depth</b>	Depth of tiles below ground surface (m)	
<b>tile_area</b>	Area of groundwater inflow ( $\text{m}^2$ ) to tile	
<b>tile_k</b>	Hydraulic conductivity of the drain perimeter ( $\text{m}/\text{day}$ )	

<b>ch_n #</b>	Manning's n for the main channels	Channel flow processes (cha)
<b>ch_k #</b>	Effective hydraulic conductivity of the main channels (mm/h)	Channel flow processes (cha)
<b>awc #</b>	Available water capacity of the soil layer (mm H <sub>2</sub> O/mm soil) for i <sup>th</sup> layer	Soil water processes (sol)

293 **3 Results and Discussion**

294 **3.1 Effect of Groundwater Saturation Excess Flow on Hydrologic Fluxes**

295 The comparison between observed and simulated monthly streamflow at five locations  
 296 (Figure 5, Table S2) shows that simulated monthly discharge considering saturation excess  
 297 runoff in the hydrologic simulation a significant improvement in term of statistical performance  
 298 of NSE, R<sup>2</sup>, KGE, and PBIAS. For the most downstream gauge (USGS 02318000; Figure 1A),  
 299 the NSE improves from 0.61 to 0.68 when including saturation excess flow, and the PBIAS  
 300 improves from 5.8 to -0.5. For Station F, the upstream most gauge in a local area with high  
 301 wetland density (Figure 1B), the NSE improves from 0.42 to 0.58 (38% increase), and KGE  
 302 from 0.37 to 0.56.



303  
 304 **Figure 5.** Observed and simulated monthly streamflow for SWAT+gwflow model for selected  
 305 USGS stream gaging sites within the study watershed for two scenarios: minimizing streamflow

306 considering saturation excess runoff [left column] and minimizing streamflow without saturation  
 307 excess runoff [right column]. Performance statistics (NSE, PBIAS, KGE) are shown for each  
 308 gage site.

309

310 The improvement in streamflow estimation is due to the higher overall water yield  
 311 generated when including saturation excess flow. As seen in Table 4, water yield when including  
 312 vs. excluding saturation excess flow is 324 mm/year (for a yield fraction of 0.25) and 296  
 313 mm/year respectively. Of the 324 mm/year, approximately half (163 mm/year) is from  
 314 groundwater saturation excess flow, resulting in a baseflow fraction of 41%. When groundwater  
 315 saturation excess flow is not considered in the simulation, the PEST seeks to compensate (i.e.,  
 316 minimize the difference between observed and simulated streamflow) by increasing tile drainage  
 317 flow and surface runoff. Nevertheless, the resulting water yield (295 mm) does not achieve the  
 318 correct magnitude of water yield (i.e., 324 mm). It is interesting to note that, rather than  
 319 increasing groundwater discharge via the channel bed by increasing channel bed hydraulic  
 320 conductivity, tile drainage outflow was increased.

321

322 **Table 4.** Average annual hydrologic fluxes (mm) for the study watershed and key hydrologic  
 323 fractions for the two calibration scenarios (with/without groundwater saturation excess flow).  
 324 Groundwater discharge = groundwater flowing into channel across channel bed.

Flux (mm)		Minimizing Streamflow (SatEx)	Minimizing Streamflow (No SatEX)
Input	Precipitation	1310.2	1310.2
	Boundary Outflow	1.1	1.5
Watershed Output	ET	953	957
	Surface Runoff	171	184
	Soil Lateral Flow	20	17
	Stream seepage	35	4.8
	Groundwater discharge	0	7.2
	Saturation Excess Flow	163	0
	Tile flow	6	92
Internal Flows	Recharge	138	126
	Pumping Irrigation	0.00	0.00
	GW-Lake Outflow	1.2	0.06
	Surface Water Irrigation	0.00	0.00
Fractions	Water Yield <sup>a</sup>	<b>324</b>	<b>296</b>
	ET Fraction <sup>b</sup>	0.72	0.73
	Baseflow Fraction <sup>c</sup>	<b>0.41</b>	<b>0.32</b>
	Yield Fraction <sup>d</sup>	0.25	0.23
	Recharge Fraction <sup>e</sup>	0.11	0.10

325 a: Water Yield = Surface Runoff + Lateral Flow + Groundwater discharge - Stream seepage + Saturation Excess Flow + Tile flow

326 b: ET / Precipitation

327 c: Net groundwater inflow to streams (Stream Seepage + Sat Excess Flow+ Tile flow) / Water Yield

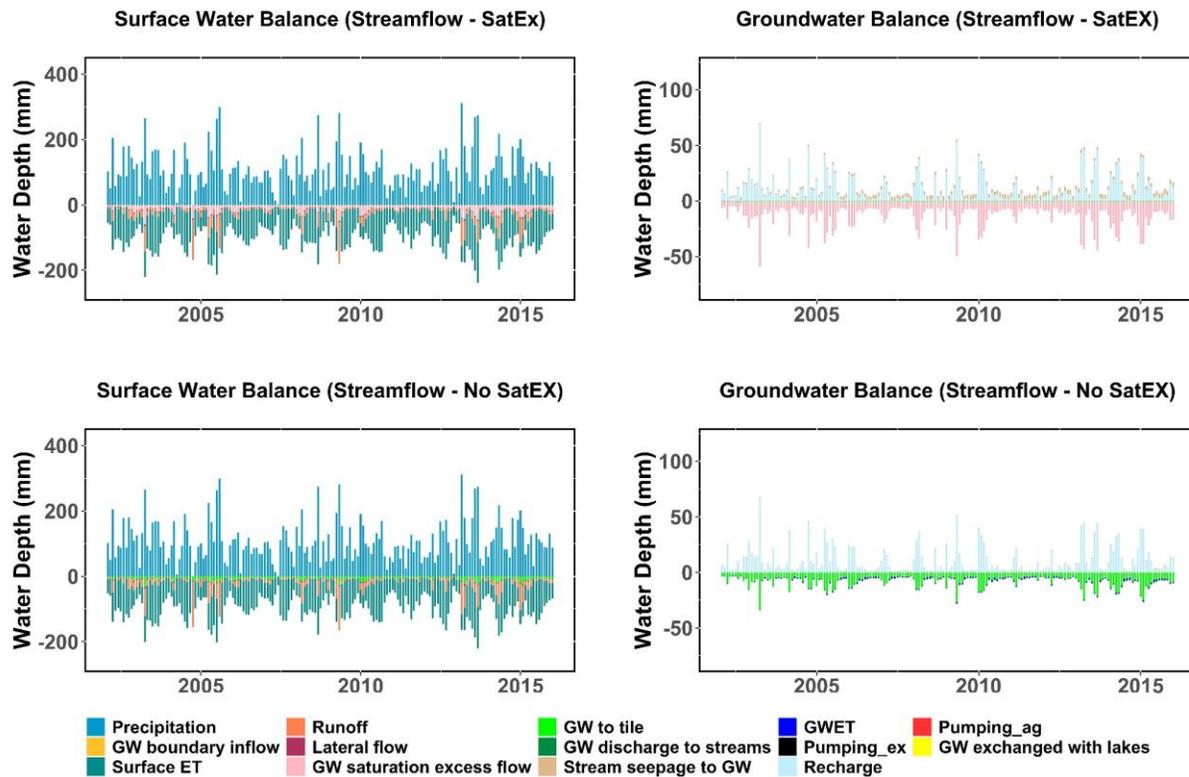
328 d: Water Yield / Precipitation

329 e: Recharge / Precipitation

330

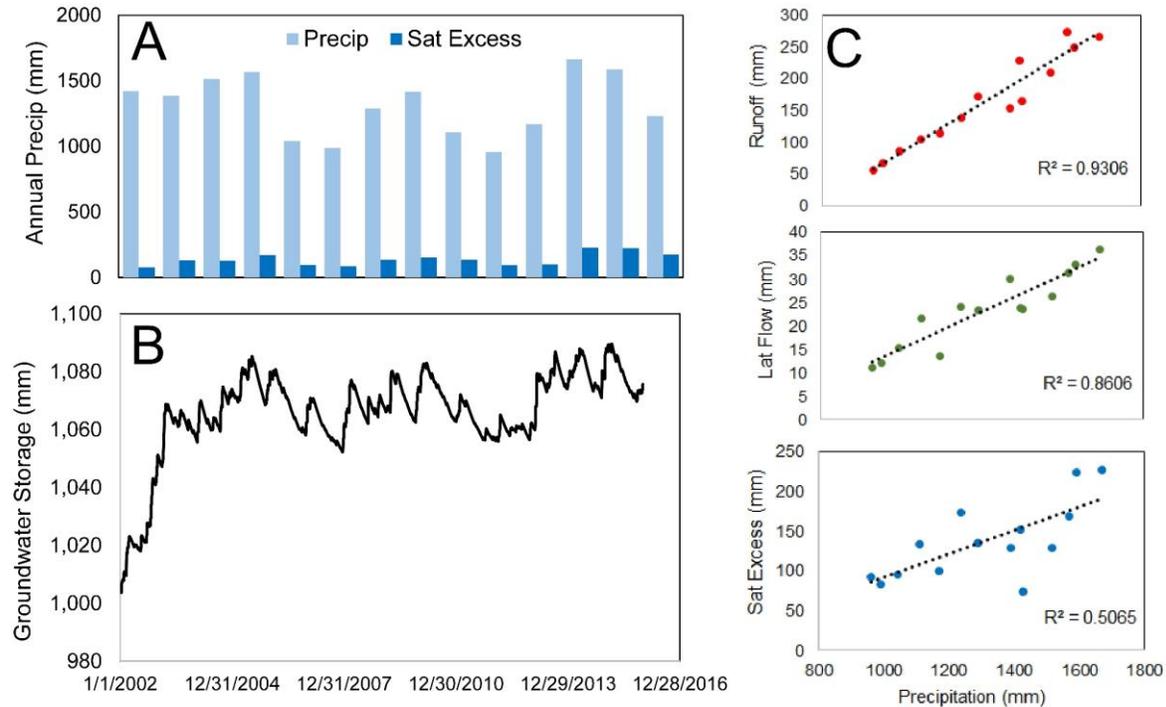
331 Monthly groundwater saturation excess flow is shown in Figure 6 for both the watershed  
 332 and groundwater systems. System inputs (precipitation, boundary inflow) are displayed as

333 positive values, whereas system outputs (tile drainage, runoff, groundwater saturation excess  
 334 flow, surface ET, and lateral flow) are displayed as negative values. Groundwater saturation  
 335 excess flow fluctuates seasonally, based on incoming recharge from rainfall events. On an annual  
 336 basis, the fraction of water yield that is groundwater saturation excess flow ranges from 0.29  
 337 (2002) to 0.58 (2011) (Figure 7A).



338  
 339 **Figure 6.** Monthly surface water fluxes (mm) [left column], and groundwater fluxes (mm) [right  
 340 column] for the simulation period of (2002–2015) for four scenarios of Little Watershed.

341  
 342 Typically, years of high groundwater saturation excess fractions correspond to years of  
 343 low rainfall. For example, rainfall for 2007 (saturation excess fraction = 0.51) was 987 mm  
 344 (compared to 1,310 mm/year average), and rainfall for 2011 (saturation excess fraction = 0.58)  
 345 was 959 mm/year. This is due to antecedent groundwater conditions, as high groundwater levels  
 346 generated during previous years intersect the ground surface during the following years,  
 347 producing high flows as compared to runoff and soil lateral flow. This is also demonstrated by  
 348 comparing annual fluxes (runoff, soil lateral flow, groundwater saturation excess flow) to annual  
 349 rainfall (Figure 7C), with groundwater saturation excess flow exhibiting a much weaker  
 350 relationship to rainfall, due to the impact of antecedent conditions.

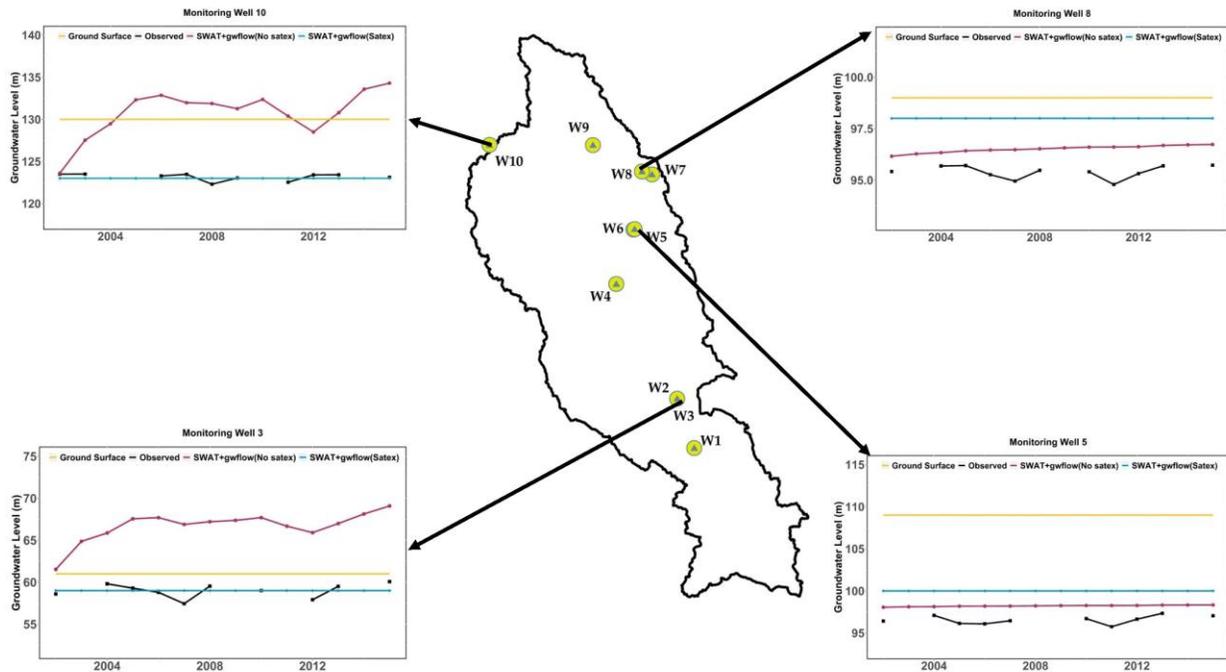


351  
 352 **Figure 7.** (A) annual precipitation (mm) and groundwater saturation excess flow (mm); (B) daily  
 353 total groundwater storage (mm), normalized to watershed area; (C) 1:1 relationships between  
 354 annual basin precipitation and hydrologic fluxes (runoff, soil lateral flow, groundwater saturation  
 355 excess flow).

356

### 357 3.2 Effect of Groundwater Saturation Excess Flow on Aquifer Features and Fluxes

358 Table S3 lists the MAE of groundwater monitoring wells (10 locations) for the two  
 359 calibration scenarios. MAE results demonstrate a significant improvement (average MAE = 1.84  
 360 m) compared to excluding groundwater saturation excess flow (MAE = 3.48 m). Nevertheless, a  
 361 few locations have greater error, although these residuals are small compared to the saturated  
 362 thickness of the aquifer. Figure 8 compares annual observed and simulated groundwater heads  
 363 for both calibration scenarios. Including groundwater saturation excess flow improves the  
 364 simulated groundwater head fluctuation at most locations. Without groundwater saturation  
 365 excess flow, groundwater head rises above the ground surface, but with no mechanisms to  
 366 release groundwater to runoff. Groundwater responds to near-surface hydrology of rainfall, ET,  
 367 soil percolation and recharge, although this response can be delayed due to antecedent  
 368 groundwater storage conditions (see Figure 7C, relationship between precipitation and  
 369 groundwater saturation excess flow).

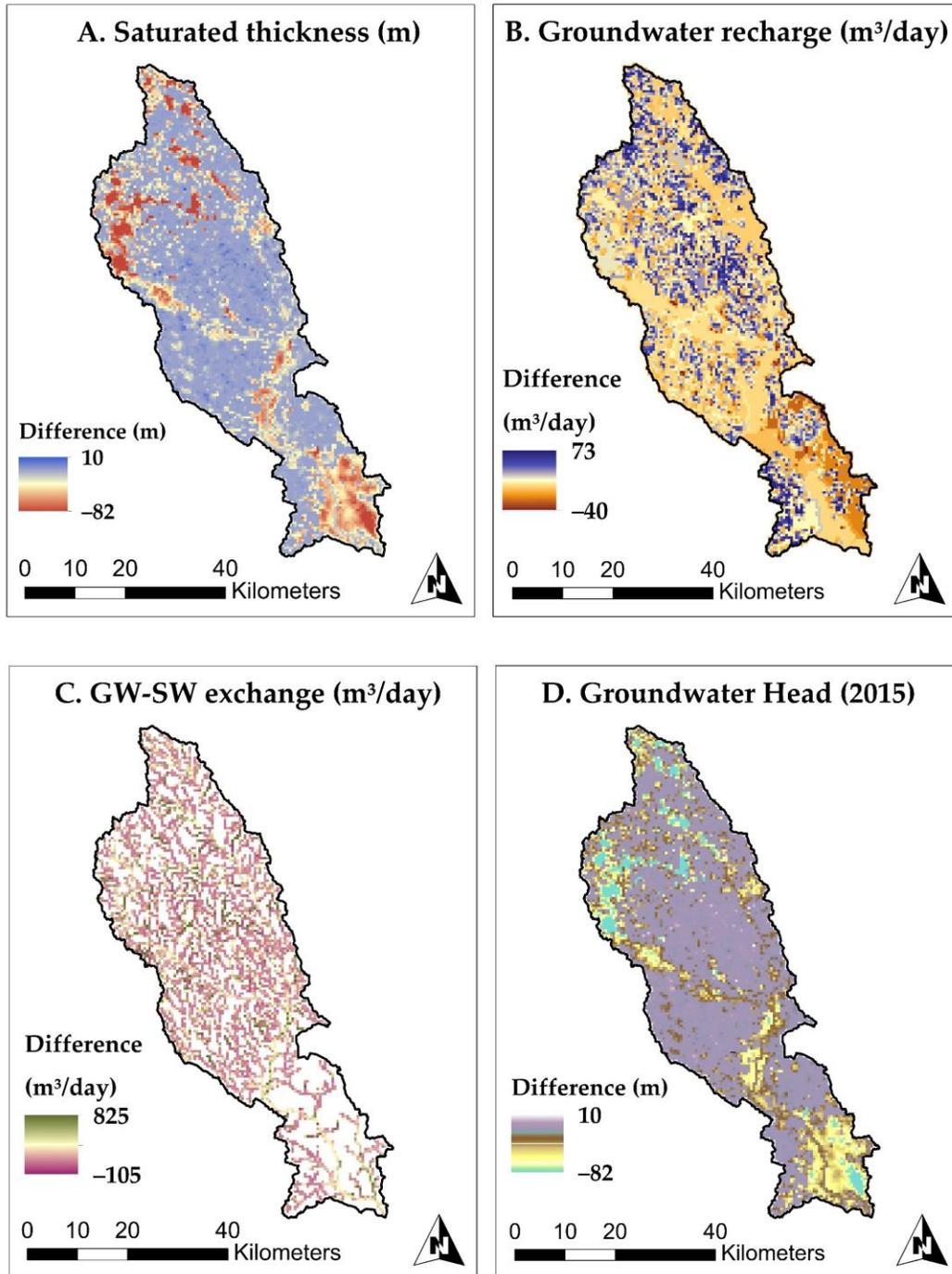


370

371 **Figure 8.** Maps of mean absolute error (MAE) (m) for a selected USGS groundwater monitoring  
 372 well locations for the simulation period of (2002–2015) for scenarios minimizing streamflow.

373

374 The influence of groundwater saturation excess flow on groundwater fluxes can be shown  
 375 spatially (Figure 9). Using cell-by-cell differences for the year 2015, saturated thickness (m) and  
 376 groundwater head (m) are much lower when including groundwater saturation excess flow, due  
 377 to the release of high groundwater to streams; groundwater recharge is much higher (purple  
 378 color) in areas between streams; and groundwater discharge via channel bed is lower. Over the  
 379 simulation period, recharge is higher when including groundwater saturation excess flow (138  
 380 mm) than excluding (126 mm). In the exclusion simulation, surface runoff is increased in an  
 381 attempt to increase streamflow to match measured values; increasing surface runoff decreases  
 382 infiltration and soil percolation, which in turns decreases recharge to the water table. This results  
 383 in a local lowering of groundwater head, in relation to the inclusion simulation.

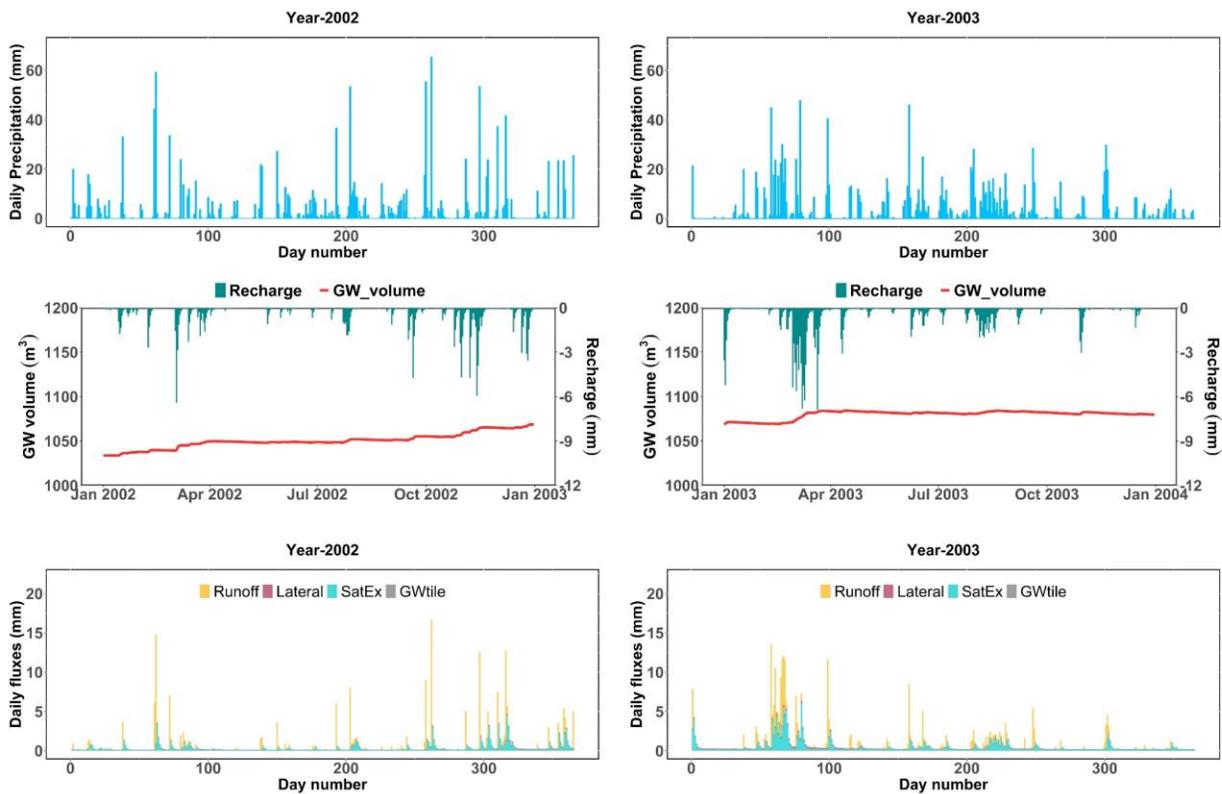


384

385 **Figure 9.** Difference maps between hydrologic simulation considering saturation excess flow  
 386 and hydrologic simulation without saturation excess flow for (a) saturated thickness (m) for year  
 387 2015; (b) average annual recharge flow (m<sup>3</sup>/day) for year 2015; (c) average annual groundwater-  
 388 stream exchange rate (m<sup>3</sup>/day) for year 2015; and (d) final (end of year 2015) groundwater head  
 389 (m).

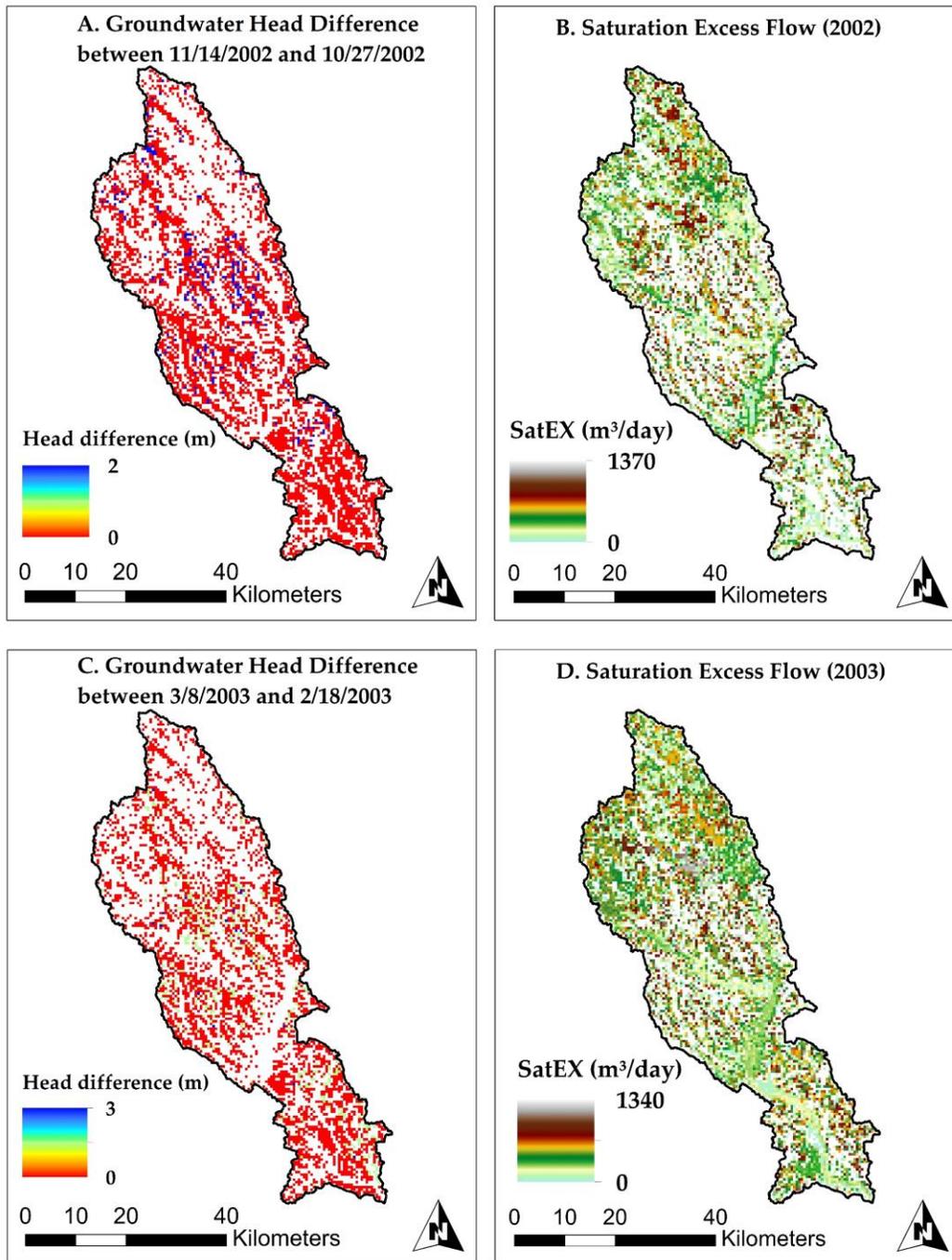
390

391 To elucidate temporal dynamics of groundwater saturation excess flow, we analyze daily  
 392 time series rainfall, recharge, groundwater volume, and groundwater saturation excess flow for  
 393 two years (2002-2003) (Figure 10). These time series demonstrate the response of groundwater  
 394 storage to rainfall-induced recharge, which in turn leads to groundwater saturation excess flow in  
 395 many local areas of the watershed. We note that groundwater saturation excess flow responds  
 396 temporally in a like manner to runoff, indicating its strong dependence on short-term rainfall  
 397 events that raise the water table to the ground surface. Figure 11 shows spatially the impact of  
 398 two rainfall periods (10/27/2002 to 11/14/2002; 2/18/2003 to 3/8/2003) on groundwater head and  
 399 groundwater saturation excess flow. Local areas of high increases in groundwater head, due to  
 400 rainfall-induced recharge, produce local areas of groundwater saturation excess flow. These  
 401 results emphasize the need to include groundwater saturation excess flow in hydrological  
 402 modeling for watersheds with shallow groundwater.



403 **Figure 10.** Average daily basin precipitation, groundwater volume, recharge, and streamflow  
 404 components for the years 2002 [left column] and 2003 [right column].  
 405

406



407

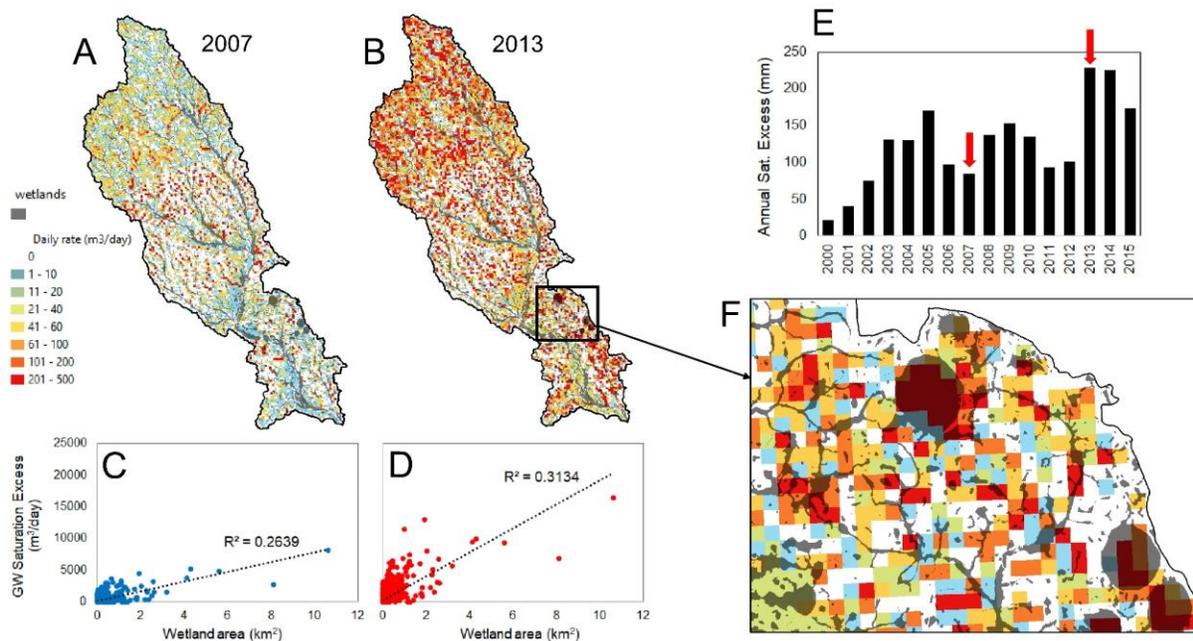
408 **Figure 11.** Maps represent (a) groundwater head difference (11/14/2002; during storm) and  
 409 (10/27/2002; before storm); (b) average daily saturation excess flow ( $\text{m}^3/\text{day}$ ) for the year 2002;  
 410 (c) groundwater head difference (3/8/2003; during storm) and (2/18/2003; before storm); and (d)  
 411 average daily saturation excess flow ( $\text{m}^3/\text{day}$ ) for the year 2003.

412

## 413 3.3 Effect of Groundwater Saturation Excess on Wetland Development

414 Groundwater saturation excess runoff is key to wetland development and vegetation as it  
 415 offers a continuous or near-continuous source of water and nutrients. Although our approach  
 416 does not explicitly simulate wetland objects within the hydrologic model, we can use the model  
 417 results to indicate locations of likely wetland development, based on timing and magnitude of  
 418 groundwater saturation excess runoff. A similar approach was used by Feinstein et al. (2019) in  
 419 delineating fen locations and comparing these locations with mapped fens for a 223 km<sup>2</sup> basin in  
 420 southeastern Wisconsin, USA, although under steady state groundwater conditions.

421 Raster maps of simulated daily average rates (m<sup>3</sup>/day) of groundwater saturation excess  
 422 flow are overlain by watershed wetlands (see Figure 1B) for years of low flux (2007) and high  
 423 flux (2013) (Figure 12A, B) based on annual groundwater saturation excess flow (Figure 12E),  
 424 showing the influence of timing on flux rates. The close comparison of wetland locations and  
 425 saturation excess flow (Figure 12F; year 2013) reveals that wetland development is often more  
 426 naturally pronounced in regions where the water table periodically rises to the ground surface.  
 427 By spatial joining saturation excess flow rates to wetland areas, we show that there are  
 428 relationships between the size of the wetland and the amount of simulated groundwater  
 429 saturation excess flow (Figure 12C, D).

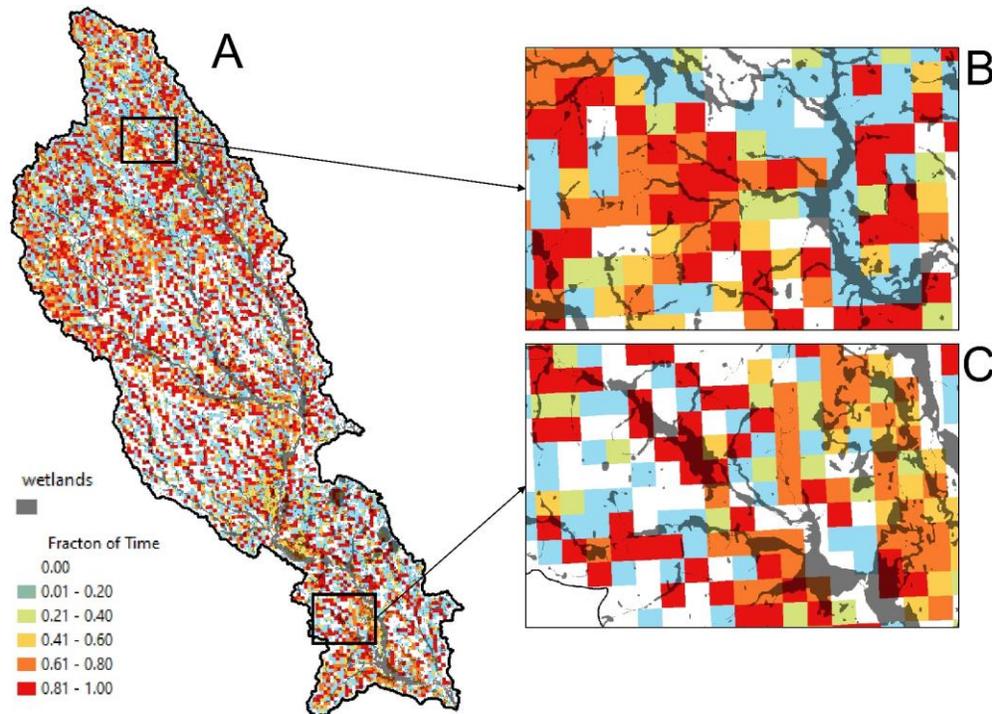


430

431 **Figure 12.** Maps represent (A) average daily saturation excess flow (m<sup>3</sup>/day) for the year 2007  
 432 (dry year) with wetland areas; (B) average daily saturation excess flow (m<sup>3</sup>/day) for the year  
 433 2013 (wet year) with wetlands areas; (C) relationship between groundwater saturation excess  
 434 runoff and wetland area for the year 2007; (D) relationship between groundwater saturation  
 435 excess runoff and wetland area for the year 2013; (E) basin average annual saturation excess  
 436 flow (mm); and (F) zoomed region of wetland and cell-by-cell saturation excess runoff flux for  
 437 the year 2013.

438

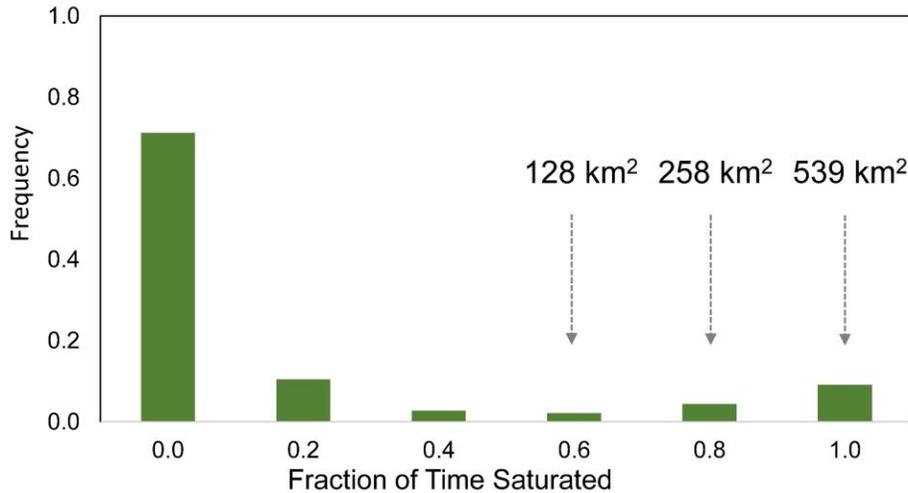
439 Finally, we show the fraction of time (over the 16-year simulation period, 2000-2015)  
 440 that each cell experiences groundwater saturation excess flow (Figure 13) and relate this fraction  
 441 to wetland locations. Although 71% of the watershed area experiences groundwater saturation  
 442 excess flow less than 20% of the time, many local areas (9% of watershed area) experience flow  
 443 more than 80% of the time (Figure 14). The locations that are saturated more than 80% of the  
 444 time cover a spatial area of 539 km<sup>2</sup> (2,155 grid cells, each with a spatial area of 500 m x 500 m  
 445 = 250,000 m<sup>2</sup>). This area corresponds well to the 399 km<sup>2</sup> covered by the delineated wetlands.  
 446 From these results we conclude that this modeling approach can simulate, in a physically based  
 447 manner, the locations and quantities of a necessary groundwater source for wetlands and other  
 448 groundwater-dependent ecosystems.



450 **Figure 13.** (A) Map of fraction of time saturated during the 2000-2015 simulation period  
 451 overlain by mapped wetland areas, showing two local areas (B, C).

452

453 The model could therefore be used as a tool to quantify the impact of system changes on  
 454 these locations and quantities, providing insight into the sustainability of wetlands in the face of  
 455 changes in climate, land use, population, and management practices. To provide a more  
 456 physically realistic modeling approach and model wetland development and evolution explicitly,  
 457 the modeling approach presented here could be further developed by implementing wetland  
 458 objects, a current option in SWAT+, and setting groundwater saturation excess flow as inflow to  
 459 these objects. Outflow from wetlands would then be added to streams, thereby providing  
 460 important timing of streamflow generation.



461  
462 **Figure 14.** Histogram of fraction of time that land surface is saturated in the study area.  
463

#### 464 **4 Summary and Conclusions**

465 In this article, we use a surface–groundwater hydrologic modeling approach with  
466 physically based spatially distributed groundwater storage and flow modeling (SWAT+*gwflow*)  
467 to represent the mechanism of groundwater saturation excess flow. We apply the approach to a  
468 humid, low-gradient watershed, the Little River Watershed in Georgia, USA and investigate its  
469 impact on hydrologic features and fluxes. Upon model calibration and testing, we analyze model  
470 results to quantify the impact of groundwater saturation excess flow on surface hydrologic  
471 fluxes, groundwater fluxes, and the generation of source water for wetlands. From the results, we  
472 conclude the following:

- 473 1. Including groundwater saturation excess flow in a watershed model can improve the  
474 estimation of streamflow generation, for the right reasons. Without this process  
475 included, automated calibration procedures (e.g., PEST, as used in this study) attempt  
476 to compensate by altering other hydrologic fluxes to unrealistic magnitudes, to match  
477 measured streamflow. In this study, tile drainage outflow was increased dramatically,  
478 even though tile drainage is not a prevalent cultivation practice in the watershed.  
479 Also, surface runoff was increased to match streamflow, thereby decreasing recharge  
480 to the water table.
- 481 2. Groundwater saturation excess flow plays a key role in governing hydrologic  
482 behavior of the watershed. Rainfall-induced recharge raises groundwater levels on a  
483 short temporal scale, leading to flashy streamflow on the same time scale as surface  
484 runoff.
- 485 3. The fraction of streamflow that originates from groundwater saturation excess flow  
486 ranges from 0.29 to 0.58, during the 2002-2015 period. Years with high fractions  
487 typically have lower annual rainfall, indicating the influence of antecedent  
488 groundwater storage conditions. Due to this influence, groundwater saturation excess  
489 flow has a weak correlation to annual rainfall as compared to surface runoff and soil  
490 lateral flow.

491 4. Locations of persistent predicted groundwater saturation excess flow correspond to  
492 observed locations of wetlands. The hydrologic model can be used to quantify the  
493 impact of system changes (climate, land use, management practices) on groundwater  
494 saturation excess flow and, through association, the presence and persistence of  
495 wetlands.

496

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501

#### 502 **Author contributions**

503 R.B., S.A., and J.A. designed the research; R.B., and J.A. performed model coding; R.B., S.A.,  
504 J.A., and M.W. performed the research; S.A., and R.B. analyzed the data; R.B., and S.A. wrote  
505 the paper.

506

#### 507 **Data Availability Statement**

508 All hydrologic simulations for SWAT+*gflow* for the Little River Watershed, GIS arrays of  
509 models outputs, R codes for plots are available at (<https://doi.org/10.5281/zenodo.10079906>).

510

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