

1 **Delayed Stormflow Generation in a Semi-humid Forested Watershed**

2 **Controlled by Soil Water Storage and Groundwater Dynamics**

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9 **Key Points:**

- 10 • Threshold dynamics between soil water content and groundwater levels govern delayed
11 stormflow generation.
- 12 • Groundwater fluctuations regulate the timing, magnitude, and merging of delayed and direct
13 stormflow peaks.
- 14 • Hydrological connectivity and hydraulic conductivity increase with rising groundwater
15 levels, driving delayed stormflow.

16

17 **Abstract**

18 Recent research by Cui et al. (2024) identified a distinct threshold governing bimodal rainfall-
19 runoff events in a semi-humid mountainous forested watershed in North China, where delayed
20 stormflow appeared to be influenced by shallow groundwater dynamics. Building on these findings,
21 this study delves deeper into the mechanisms driving these bimodal events, focusing on the
22 interactions between soil water content (SWC) and groundwater level (GWL) during storm events.
23 The results show that delayed stormflow is primarily governed by the interplay between SWC and
24 GWL. Delayed stormflow is initiated when SWC exceeds the soil's water storage capacity, while its
25 timing and magnitude further modulated by GWL fluctuations. During rainfall, SWC increases
26 rapidly but stabilizes after the rain ceases if the soil's water-holding capacity is not reached.
27 Conversely, when SWC surpasses the storage capacity, the excess rainwater infiltrates into the
28 subsurface, recharging groundwater and causing a gradual rise in GWL. As GWL rises, enhanced
29 hydraulic conductivity facilitates the lateral movement of shallow groundwater toward the stream
30 channel, generating delayed stormflow. Concurrently, the effective connectivity between the stream
31 channel and adjacent hillslopes increases in the vertical dimension. At higher GWL thresholds, GWL
32 responses across the watershed become synchronized, significantly boosting groundwater discharge
33 and reducing lag times. In extreme cases, the delayed stormflow peak converges with the direct
34 stormflow peak. These findings advance the understanding of delayed stormflow mechanisms in
35 semi-humid mountainous watersheds, contributing to refining runoff generation theories by providing
36 insights into the threshold-driven processes that govern the timing and volume of delayed stormflow.

37 **Keywords:** Delayed stormflow; Soil water storage; Groundwater outflow; Stormflow generation
38 mechanism; Hydraulic conductivity

39 **1. Introduction**

40 Stormflow processes in the Xitaizi Experimental Watershed (XEW), located in North China,
41 exhibit a frequent occurrence of bimodal stormflow hydrographs (Fig. A1), which often lead to

42 significant stormflow and associated localized inundation. Analysis of 15 such events over the past
43 decade revealed that the onset of these bimodal hydrographs is governed by threshold behavior.
44 Specifically, delayed streamflow peaks tend to emerge when the combined total of event rainfall and
45 antecedent soil moisture index exceeds 200 mm. The authors' findings suggest that shallow
46 groundwater contributions are primarily responsible for these delayed stormflow events (Cui et al.,
47 2024). However, the mechanisms behind the development of these bimodal hydrographs, which
48 represent complex emergent hydrological behaviors, remain poorly understood. Understanding the
49 formation of delayed stormflow is critical for advancing our comprehension of runoff generation
50 processes and improving flood forecasting.

51 Bimodal hydrographs, characterized by dual streamflow peaks, typically occur during the
52 wetting-up phases of catchments. Extensive research has identified several factors that influence dual
53 streamflow peaks, including antecedent soil moisture, antecedent precipitation, groundwater levels,
54 soil water storage, and rainfall amount (Haga et al., 2005; Graeff et al., 2009; Anderson and Burt,
55 1978; Padilla et al., 2015; Martínez-Carreras et al., 2016). Despite these advancements, the specific
56 mechanisms that lead to threshold behavior and how these mechanisms produce the diverse shapes
57 of stormflow hydrographs are still inadequately explained. For instance, Martínez-Carreras et al.
58 (2016) found that a delayed peak only occurred when watershed storage reached a critical threshold
59 of 113 mm. However, the precise reasons for this threshold and the underlying processes remain
60 unclear.

61 The occurrence of bimodal hydrograph reflects a nonlinear runoff response, which offers
62 valuable insights into the complex interactions between rainfall and runoff. The nonlinear pattern,
63 characterized by both the timing and magnitude of the response, plays a crucial role in understanding
64 stormflow processes. Recent decades have seen an increase in research on nonlinear and threshold
65 changes in rainfall-runoff responses, contributing to a deeper understanding of stormflow generation
66 mechanisms. Nonlinear patterns, often characterized by rapid runoff responses that may lead to
67 flooding, have been extensively documented in recent decades (Detty and McGuire, 2010; Farrick
68 and Branfireun, 2014; Graham et al., 2010; Tromp-van Meerveld and McDonnell, 2006a; Penna et

69 al., 2011; Scaife et al., 2020). However, many studies fail to explore the intricate post-threshold
70 mechanisms of these nonlinear shifts, leaving a gap in our understanding of stormflow generation
71 across various catchments. While threshold behaviors are widely recognized, the detailed processes
72 governing these shifts and their subsequent runoff dynamics remain underexplored.

73 Bimodal stormflow responses present an opportunity to investigate the relationship between
74 rainfall thresholds and runoff generation, offering new perspectives on the timing and variability of
75 stormflow. Despite this, many studies fail to distinguish between unimodal and bimodal streamflow
76 responses. For example, Detty and McGuire (2010) focused on hydrological threshold responses but
77 did not differentiate between unimodal and bimodal hydrographs, as their study primarily addressed
78 general nonlinear rainfall-runoff processes in general. Similarly, Martínez-Carreras et al. (2016)
79 observed delayed peaks and identified catchment storage as a key factor influencing streamflow
80 responses, however, they did not explicitly differentiate the underlying mechanisms between
81 unimodal and bimodal responses. Such limitations often arise because the second peak in bimodal
82 responses typically occurs after the rainfall event has ended, whereas many studies focus on
83 streamflow changes during the event itself. Additionally, bimodal responses are influenced by
84 catchment-specific topography and geology, making them less observable in certain regions. These
85 challenges highlight the need for more in-depth investigation into bimodal streamflow responses to
86 enhance our understanding of their mechanisms. Therefore, an in-depth investigation into the
87 mechanisms driving these responses is essential. Such research would enable the grouping of similar
88 hydrologic responses and facilitate comparisons of stormflow generation processes across different
89 watersheds (Graham and McDonnell, 2010; Tromp-van Meerveld and McDonnell, 2006a, b).

90 Extensive studies across diverse regions have explored the role of soil water content and
91 groundwater levels in generating delayed peaks in stormflow. Detty and McGuire (2010) emphasized
92 subsurface flow thresholds in a forested catchment in the USA, while Farrick and Branfireun (2014)
93 analyzed soil moisture and groundwater interactions in Canadian wetlands. Penna et al. (2011)
94 examined antecedent soil moisture and storage thresholds in alpine catchments in New Zealand.
95 These studies, along with others from regions such as Japan (Haga et al., 2005) and Europe (Graeff

et al., 2009), contribute to the growing body of knowledge on threshold behavior in stormflow responses. However, while these studies highlight the occurrence of thresholds, the complex interactions that drive post-threshold runoff processes remain insufficiently understood.

Investigating stormflow events in semi-humid regions, such as XEW, is challenging due to the relatively arid climate and low runoff coefficients. Over nearly a decade, 95 storm events were identified and analyzed in XEW, offering a rare and valuable dataset for examining bimodal stormflow responses in such regions. This study builds on prior findings to uncover the processes underlying delayed stormflow patterns. We hypothesize that the generation of delayed stormflow is governed by threshold-dependent interactions between soil water content (SWC) and groundwater level (GWL). The primary objectives of this study are: (1) to analyze the temporal dynamics of SWC and GWL during storm events, (2) to elucidate the mechanisms driving the threshold behavior observed in bimodal hydrographs, and (3) to reveal the underlying processes responsible for delayed stormflow in XEW.

2. Materials and methods

2.1 Study site

The study was conducted in the Xitaizi Experimental Watershed (XEW), a 4.22 km² catchment located in North China (40°32'N, 116°37'E), approximately 70 km northeast of Beijing at (Fig. 1). The watershed's elevation ranges from 676 to 1201 m above sea level, and the region experiences a monsoon-influenced semi-humid climate. The average annual precipitation is 625 mm, with 80% concentrated between June and September. The mean annual temperature is 11.5°C with an average relative humidity of 59.1%. Forests cover 98% of the watershed, with broad-leaved species and shrubberies accounting for 54.2% and 33.0%, respectively.

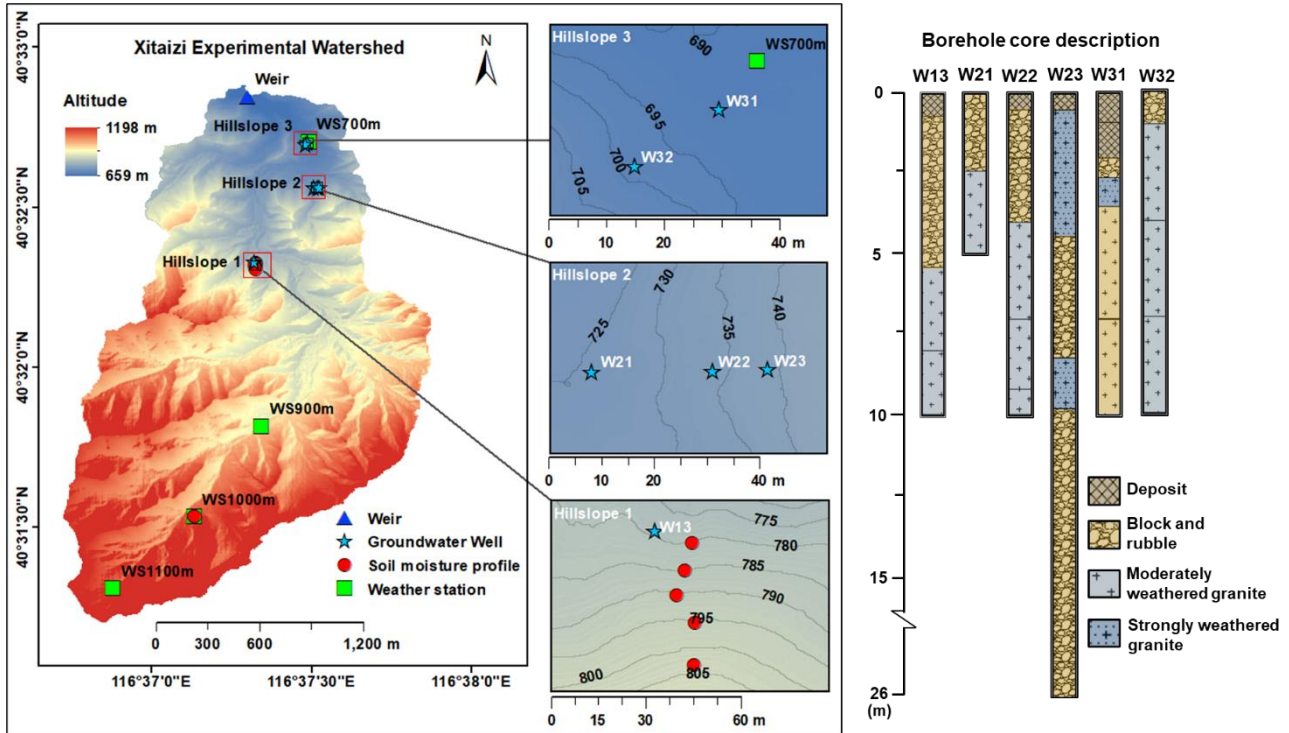


Figure 1. Location of Xitaizi Experimental Watershed (XEW) and a simple description of the borehole cores. This figure shows the distribution of monitoring instruments, including four weather stations (WS700, WS900, WS1000, and WS1100), an outlet weir, six groundwater observation wells, and eight soil moisture observation profiles. Of the eight soil moisture profiles, five are located on Hillslope 1, while the remaining three are positioned on the slope near WS1000. Research hillslopes (Hillslope 1, Hillslope 2, and Hillslope 3) are delineated as key zones for hydrological and geological investigations.

The soils in XEW are primarily brown earth and cinnamon soils, with depths up to 1.5 m and an average saturated hydraulic conductivity of 45 mm/h. The surface soil is rich in organic matter, enhancing infiltration and reducing surface runoff potential. Underlying geology is predominantly compacted, deeply weathered granite (80% of the area), with smaller portions of gneiss and dolomite. Fractured granite facilitates vertical and lateral subsurface flow, contributing to delayed groundwater responses. Slug tests estimated the saturated hydraulic conductivity of weathered granite to range from 5.2×10^{-3} m/day to 1.16 m/day.

2.2 Research hillslopes and instrumentation

Three research hillslopes (Hillslope 1, Hillslope 2, and Hillslope 3) were selected to investigate hydrological processes under varying geological and topographical conditions. Hillslope 1 (HS1)

136 features thick soils overlying fractured granite, Hillslope 2 (HS2) has a highly permeable fractured
137 block layer, and Hillslope 3 (HS3) consists of shallow soils over weakly weathered bedrock.

138 To capture spatial variability, SWC probes and boreholes were installed along hilltops, mid-
139 slopes, and foot slopes. Groundwater boreholes, ranging from 5 to 26 m deep, were equipped with
140 HOBO capacitance water level loggers to record GWLs (Fig. 1).
141 data collection

142 Meteorological data spanning 2013–2023 were collected from four GRWS100 automatic
143 weather stations (WS700, WS900, WS1000, and WS1100), positioned at elevations of 700, 900, 1000,
144 and 1100 m, respectively. Rainfall was recorded at 10-minute intervals using six tipping-bucket rain
145 gauges near the weather stations, and the data were averaged for analysis.

146 Streamflow was measured at the catchment outlet using a Parshall flume, with water levels
147 logged every 5 minutes since 2014. Data from some events were excluded due to sensor malfunctions
148 or poor data quality, including key rainfall events in 2018 and 2019. Despite these exclusions, 95
149 rainfall-runoff events were analyzed, offering robust data for investigating bimodal stormflow
150 characteristics.

151 **2.4 Soil water content and groundwater level monitoring**

152 Volumetric SWC was monitored at eight sites using CS616 time-domain reflectometry (TDR)
153 probes installed at 10 cm intervals from the surface to 80 cm depth. Five profiles were located along
154 HS1, and three were near WS1000. Measurements were recorded every 10 minutes, and the arithmetic
155 mean of SWC values was used for analysis.

156 GWLs (below the ground surface, hereinafter referred to as bgs) were observed in six boreholes
157 distributed across the hillslopes. HOBO capacitance water level loggers recorded hourly data. To
158 facilitate comparisons, GWLs were normalized using the method described by Detty and McGuire
159 (2010). This normalization, expressed as the GWL index (I_G), standardizes GWLs across wells with
160 varying ranges.

Groundwater levels were normalized following the method described by Detty and McGuire (2010). For each well and event, the median height of the water table above the lowest recordable depth of the instrument was calculated and normalized to the total observed range, where 0 represents the minimum height and 1 represents the maximum height. This normalized value was referred to as the groundwater index (I_G). We used I_G to facilitate comparisons across wells with different absolute GWL ranges and to represent the overall GWL dynamics in the watershed.

Streamflow and SWC data were aggregated to hourly intervals for alignment with GWL data. Preliminary analysis confirmed that the delayed second streamflow peak had response times exceeding the hourly scale, rendering this aggregation sufficient for the study's purposes.

2.5 Rainfall-runoff event identification and hydrograph analysis

Rainfall events were identified using an intensity-based automatic algorithm described by Tian et al. (2012) that defines event with rainfall intensity >0.1 mm/h and a minimum separation of six hours between events. Events with cumulative rainfall exceeding 5 mm were analyzed.

Bimodal rainfall-runoff events were manually identified based on two criteria: (1) the presence of a secondary, arch-shaped runoff peak occurring after rainfall cessation or during minimal intermittent rainfall, and (2) A distinct separation between the direct (sharp) and delayed (broad) peaks. More details of the classification are described in Cui et al. (2024, HESS).

The combination of automatic event delineation and manual identification ensured the accurate selection of 14 rainfall-runoff events with well-defined delayed peaks for subsequent analysis. Streamflow was separated into storm runoff and baseflow using the HYSEP program with the constant slope method (Hewlett and Hibbert, 1967; Sloto & Crouse, 1996), supplemented by manual adjustments for complex hydrographs. Event stormflow volumes were calculated as total discharge minus baseflow.

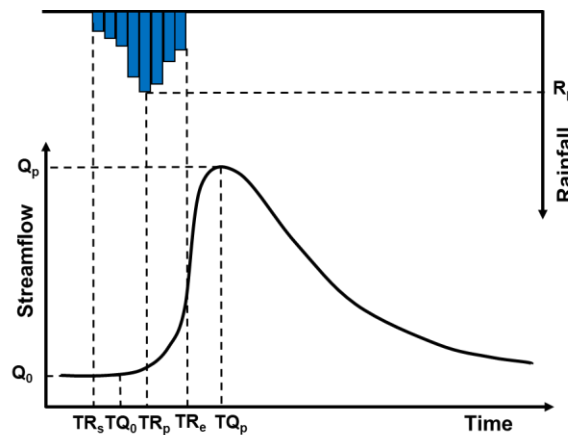
Streamflow was separated into storm runoff and baseflow using the HYSEP computer program with the constant slope method, supplemented by manual adjustments for complex hydrographs.

186 Throughout the manuscript, stormflow refers to the total discharge, and event stormflow volumes
 187 were calculated as total discharge minus baseflow, which are expressed in q_s .

188 2.6 Hydrological connectivity analysis

189 Hydrological connectivity among streamflow, SWC, and GWL was analyzed to examine the
 190 interplay of subsurface flow pathways. Rainfall-runoff events were analyzed based on their total
 191 rainfall (>5 mm) and corresponding streamflow peaks. As illustrated in Fig. 2, The peak rainfall
 192 intensity (R_p) was determined based on the maximum 1-hour rainfall intensity, with the time of
 193 occurrence recorded as TP_p . Metrics such as initial streamflow (Q_0) and peak streamflow (Q_p) were
 194 determined alongside their respective timings (TQ_0 and TQ_p).

195 Similar metrics were calculated for SWC and GWL, including initial values (SWC_0 and I_{G0}) and
 196 peak values (SWC_p and I_{Gp}), with corresponding times of occurrence (TS_0 , TI_{G0} , TS_p and TI_{Gp}). These
 197 metrics allowed for a comprehensive evaluation of the soil water-groundwater-streamflow response
 198 relationship across 95 distinct rainfall-runoff events.



200 **Figure 2.** Conceptual framework of rainfall event analysis.

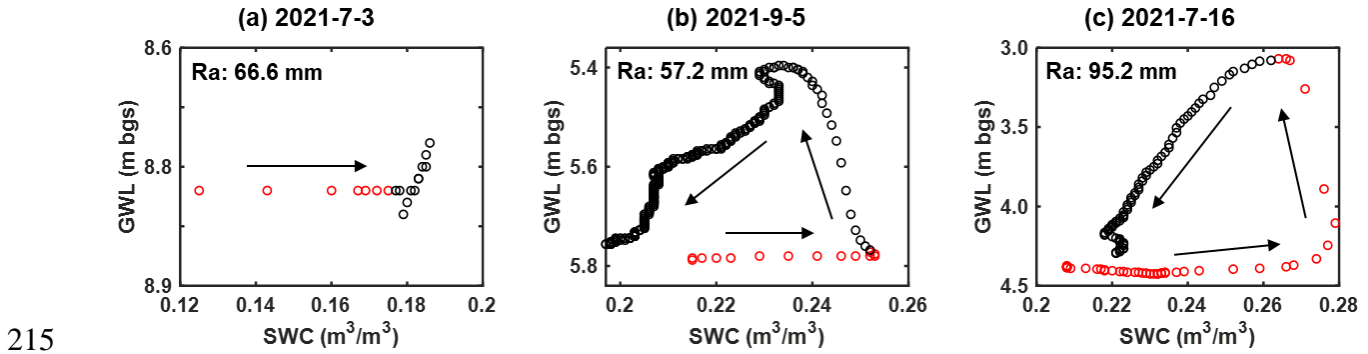
201 3. Results

202 3.1 Hillslope-scale dynamics of SWC and GWL during rainfall-runoff events

203 The temporal evolution of SWC and GWL was analyzed across 95 rainfall-runoff events to

204 understand their dynamic interaction. Our analysis revealed a clear relationship between SWC and
 205 GWL dynamics, with SWC initially increasing rapidly during rainfall, followed by a stabilization or
 206 decline once a threshold was reached. In contrast, GWL showed a more delayed response (Fig. 3).
 207 Three distinct patterns of SWC and GWL interaction were identified.

208 Figure 3 illustrates the dynamics of SWC and GWL during three representative events. These
 209 events were selected to demonstrate the variability in SWC and GWL patterns identified across the
 210 95 rainfall-runoff events. The selected events all occurred within the same year to minimize inter-
 211 annual variability and ensure comparability. Red circles indicate rainfall periods, while black circles
 212 represent post-rainfall periods. In dry conditions, despite 66.6 mm of rainfall, SWC remained
 213 relatively low (<0.20), with a gradual increase during rainfall followed by stabilization after rainfall
 214 ceased. GWL showed minimal response. (Fig. 3a).



215 **Figure 3.** Three typical SWC-GWL dynamics patterns during rainfall-runoff events. Ra is rainfall
 216 amount. Arrows indicate the temporal evolution of the events. Red circles indicate periods of
 217 rainfall, while black circles denote post-rainfall periods.
 218

219 In events with wet conditions, both SWC and GWL showed significant increases. However, the
 220 timing of GWL rise varied: in some cases, GWL rose after the cessation of rainfall, while in other
 221 cases, it began rising before the rainfall ended. The primary distinction between these patterns lies in
 222 the timing of the GWL rise: in Fig. 3b (57.2 mm rainfall), GWL began to rise after the rainfall ended,
 223 whereas in Fig.3c (95.2 mm rainfall), GWL started to rise noticeably before the end of the rainfall.

224 In the case represented by Fig. 3b, SWC increased significantly, surpassing 0.20, while GWL
 225 showed a delayed rise after the rainfall ceased. The counterclockwise hysteresis was observed as

SWC continued to increase while GWL remained largely unchanged during rainfall. Fig. 3c, which typically represents intense storm events, showed a sharp increase in both SWC and GWL, with SWC exceeding 0.25. GWL began to rise before the rainfall ended, reaching a peak as rainfall continued, and both variables showed a substantial decline after rainfall ceased. These representative events highlight the variability in the SWC-GWL relationship, with timing differences in the rise of GWL and distinct hysteresis patterns during moderate and extreme events.

We further quantified the frequency distribution of SWC and GWL increases or decreases across the 95 rainfall-runoff events (Fig. 4). Notably, in 49 events, SWC increased, while GWL increased in 43 events. In contrast, SWC declined in 26 events and GWL declined in 15 events. Importantly, 15 events showed a simultaneous decline in both SWC and GWL, which were associated with delayed stormflow and larger stormflow volumes. One such event, on August 15, 2021, exhibited fluctuating SWC and GWL values throughout the rainfall event due to the more dispersed rainfall distribution. As a result, our subsequent analysis primarily focused on the remaining 14 events with well-defined response characteristics.

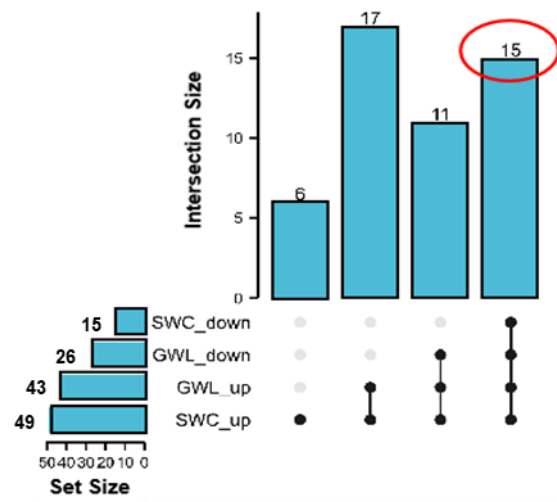


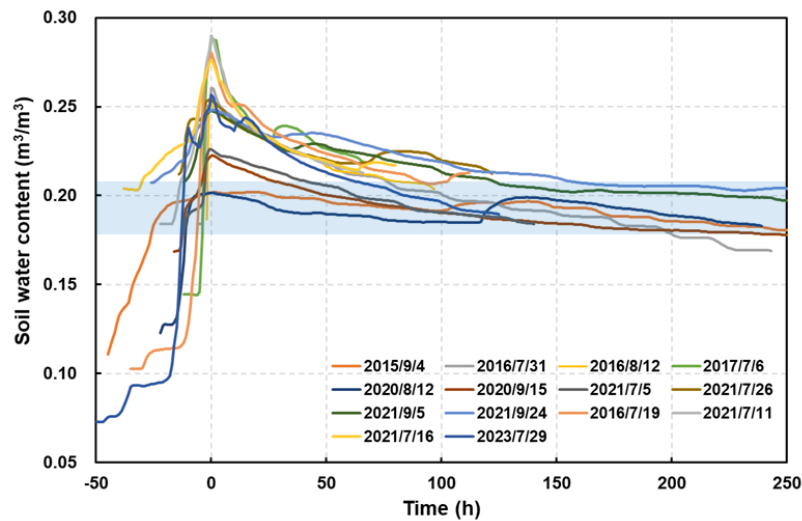
Figure 4. UpSet plot of the response characteristics of SWC and GWL during rainfall-runoff events.

3.2 SWC dynamics across rainfall-runoff events

Figure 5 presents the SWC dynamics observed during 14 distinct rainfall-runoff events, each

245 characterized by minimal or no intermittent rainfall during the recession period. To facilitate a clear
 246 comparison of SWC changes across events, the peak of each event was aligned with a horizontal axis
 247 value of 0.

248 During the initial rainfall phase, SWC increased rapidly, reaching a peak value. As the rainfall
 249 ceased, SWC began to decline, though at a slower rate, eventually stabilizing at a specific value. To
 250 quantify the threshold at which SWC stabilizes, we conducted a statistical analysis of the stable SWC
 251 during these events. The stable phase was defined as the period following the recession phase when
 252 SWC exhibited minimal variation before subsequent rainfall. The statistical analysis of the stable
 253 SWC revealed a mean value of 0.1974, with a standard deviation of 0.0158 and a 95% confidence
 254 interval of [0.1945, 0.2003]. These results validate the visually observed threshold of 0.20 for SWC
 255 stabilization. The general pattern of SWC variation is schematically illustrated in Fig. 6.



256
 257 **Figure 5.** SWC dynamics during different storm events.

258 The SWC response to rainfall is rapid. Upon rainfall onset, SWC increased sharply. Once the
 259 rainfall ceased, the subsequent behavior of SWC depends on whether the peak value exceeds the 0.20
 260 threshold. If SWC remains below or at 0.20, it either stabilizes or declines slowly. However, when
 261 SWC exceeds 0.20, it decreased rapidly before stabilizing around the 0.20 threshold. The magnitude
 262 of the peak above 0.20 influences the speed of the subsequent decline in SWC: the greater the peak,
 263 the faster the decline.

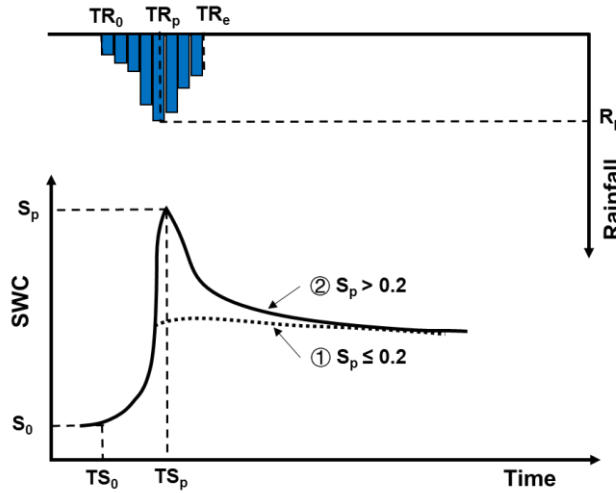


Figure 6. Conceptual diagram of SWC response during storm events. S_p is the maximum SWC value.

3.3 GWL dynamics and response types

This section examines GWL dynamics during 14 selected rainfall-runoff events, chosen for their clear and consistent GWL and SWC patterns. These events facilitate a detailed investigation into groundwater response to storm events. Two distinct GWL response types—quick and slow—were identified and are conceptually illustrated in Fig. 7. It is important to note that Fig. 7 is a conceptual representation, not based on specific rainfall-runoff events, and does not include rainfall depth data.

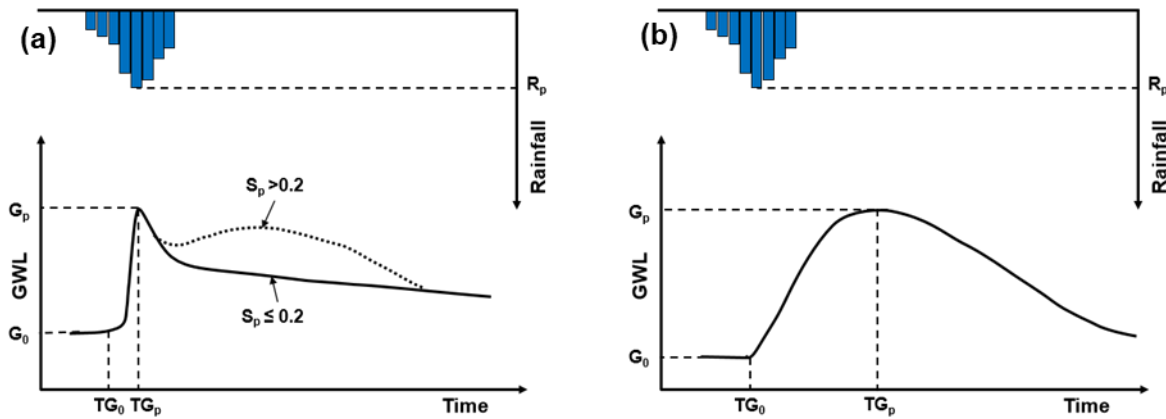


Figure 7. Conceptual diagram of GWL response during storm events. G_0 and G_p represent the initial and maximum values of GWL, respectively. S_p denotes the maximum SWC value.

In events exhibiting a quick response, the GWL rises rapidly, closely aligning with the SWC

277 peak. The GWL response typically lags behind the SWC peak by 0 to 6 hours (Fig. 7a). For events
 278 where SWC exceeds 0.20 (and particularly when it surpasses 0.24), the GWL often shows a secondary
 279 rise following the initial peak, as indicated by the dotted line in Fig. 7a. Conversely, the slow response
 280 occurs when SWC declines sharply after reaching its peak, resulting in a delayed rise in GWL (Fig.
 281 7b).

282 An analysis of GWL responses across various hillslopes revealed spatial variability. For instance,
 283 the GWL at HS2 (well W21-23) exhibited a quick response (Fig. 7a), whereas GWLs at HS1 (W13)
 284 and HS3 (W31 and W32) displayed slow response characteristics (Fig. 7b). These findings suggest
 285 that the GWL dynamics are influenced not only by SWC but also by the underlying geological
 286 structure of each hillslope.

287 At the watershed scale, GWL response to storm events demonstrated considerable spatial
 288 variability. I_G , represents the average normalized GWL across multiple wells, was used to capture an
 289 integrated view. Analysis revealed that I_G often exhibited two distinct peaks during storm events.
 290 Among the 14 events analyzed, 9 events displayed dual I_G peaks, coinciding with the two peaks in
 291 streamflow. However, at the individual well level, only W13 (HS1) and W23 (HS2) exhibited dual
 292 GWL peaks. Specifically, W13 showed two peaks during one event, while W23 exhibited two peaks
 293 during five events. The remaining wells displayed only a single peak across all events analyzed (see
 294 Table 1).

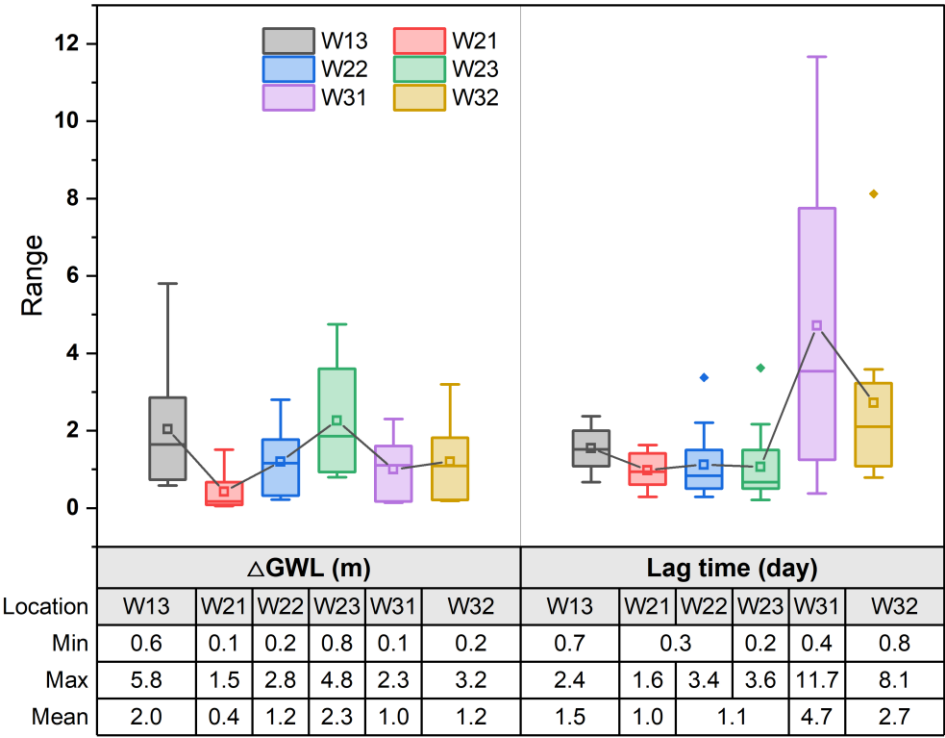
295 **Table 1.** Statistical results of response characterization of streamflow, I_G and groundwater levels.

	Streamflow	I_G	HS1		HS2		HS3	
			W13	W21	W22	W23	W31	W32
Total number of events	14	14	14	8	14	14	9	9
Number of events with two peaks	9	9	1	0	0	5	0	0

296 3.4 GWL responses across hillslope positions

297 Further examination of GWL responses across various locations is presented in Fig. 8, which

298 shows the magnitude of GWL increases and their lag times relative to rainfall onset. While variations
 299 in GWL were observed among the monitoring wells, the differences in GWL increments were
 300 generally modest, with mean increases ranging from 1 to 2 meters. Notably, smaller GWL changes
 301 were recorded at the foot of the hillslope (e.g., W21 and W31). Within the same hillslope, GWL
 302 increments tended to increase progressively from the foot to the top, as seen in HS2 (W21-W23) and
 303 HS3 (W31 and W32).



304
 305 **Figure 8.** GWL increments (ΔGWL) and lag time of peak GWL relative to rainfall onset at different
 306 locations.

307 In contrast, the lag times for maximum GWL exhibited greater variation across locations. For
 308 instance, at HS3, lag times ranged from 0.4 to 11.7 days at W31 and from 0.8 to 8.1 days at W32,
 309 significantly longer than those at HS1 (0.7 to 2.4 days) and HS2 (0.2 to 3.6 days). Interestingly, within
 310 a single hillslope, no consistent relationship was found between the lag time of maximum GWL and
 311 its distance from the foot of the hillslope.

312 To further investigate these dynamics, the relationship between GWL increments and SWC was
 313 analyzed across 14 storm events (Fig. 9). The analysis focused on six observation wells (W13, W21–

W23, W31, and W32) located on three hillslopes (see Fig. 1 for well locations). The variability in
 GWL response types—quick versus slow—was attributed to spatial differences in SWC thresholds
 and hillslope geological structures.

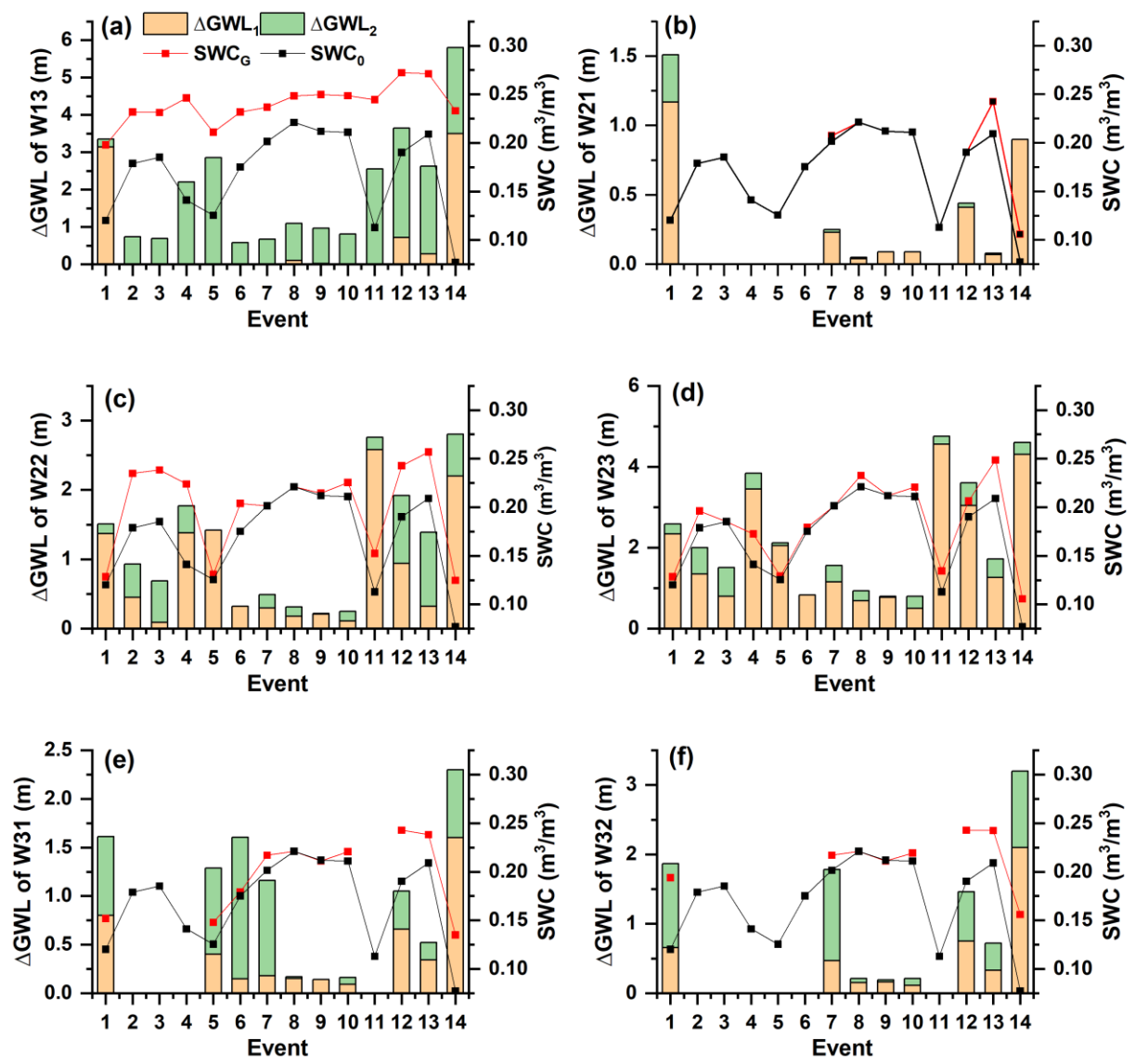


Figure 9. GWL increments (ΔGWL) across various locations during 14 storm events, along with initial SWC (SWC_0) and SWC at the onset of GWL rise (SWC_G). The orange bars represent ΔGWL during the SWC increase phase, while the green bars represent ΔGWL during the SWC decline phase. The red and black lines denote SWC_G and SWC_0 , respectively.

In Fig. 9, the orange bars represent GWL increments during the SWC increase phase (up to its peak), while the green bars indicate GWL increments during the SWC decline phase (from its peak

324 to when GWL reached its maximum. The black and red dotted lines mark the initial SWC (SWC_0)
325 and the SWC at the onset of GWL rise (SWC_G), respectively. Missing data for some locations are
326 indicated by the absence of bars in Figs. 9b, 9e, and 9f.

327 The analysis revealed that the magnitude of the SWC increase following rainfall onset is a key
328 determinant of delayed GWL responses. Specifically, a greater difference between SWC_G and SWC_0
329 corresponded to an onset of GWL rise begins. Conversely, when SWC_G and SWC_0 are similar, GWL
330 rose almost simultaneously with the SWC increase.

331 At HS1 (W13), GWL began to rise only after SWC exceeded 0.20. Most of the GWL increase
332 occurred during the SWC decline phase, suggesting that soil wetness exerts a threshold effect on
333 GWL dynamics. This delayed response aligns with the slow response type. At HS2 (W21-23), GWL
334 responses were more immediate, with GWL increases closely following SWC rises. SWC_G values at
335 these locations ranged widely (0.13-0.26) but were generally close to SWC_0 , indicating that GWL
336 responses at HS2 are less dependent on SWC thresholds and exhibit a quick response type.

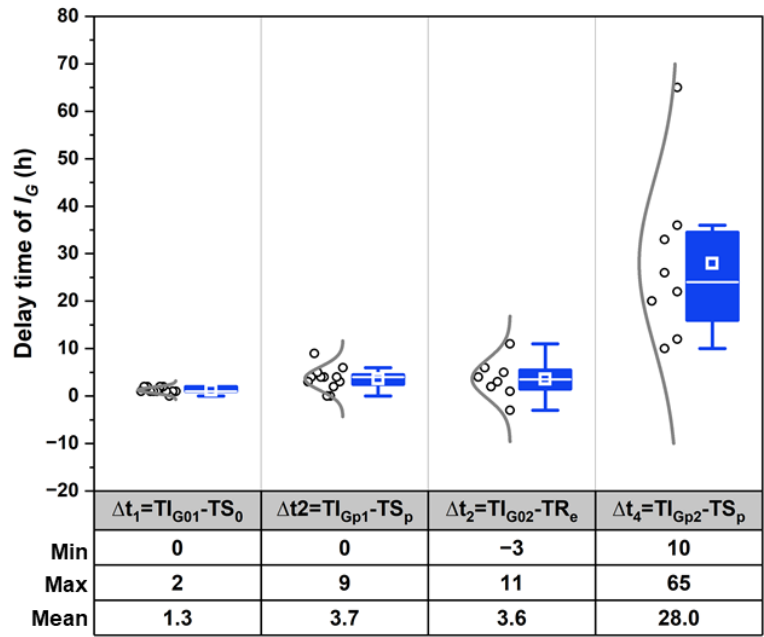
337 HS3 demonstrated both quick and slow GWL responses. Initial rises occurred soon after the
338 SWC increase, but the majority of GWL increments took place during the prolonged SWC decline
339 phase following its peak. This pattern suggests a more complex interaction of immediate and delayed
340 factors influencing GWL dynamics at HS3.

341 These findings highlight a strong relationship between the emergence of quick and slow GWL
342 response types and SWC dynamics. In quick response types, GWL increments occur primarily during
343 the SWC increase phase, resulting in a steep response curve. In slow responses, GWL increments
344 predominantly occur during the SWC decline phase, producing an arch-shaped response curve. These
345 distinctions underscore the pivotal role of SWC dynamics in regulating the timing and magnitude of
346 GWL responses across different hillslopes.

347 **3.5 Characterization of groundwater response at the watershed scale**

348 Figure 10 illustrates the timing of I_G peaks relative to SWC response. The first I_G peak occurred

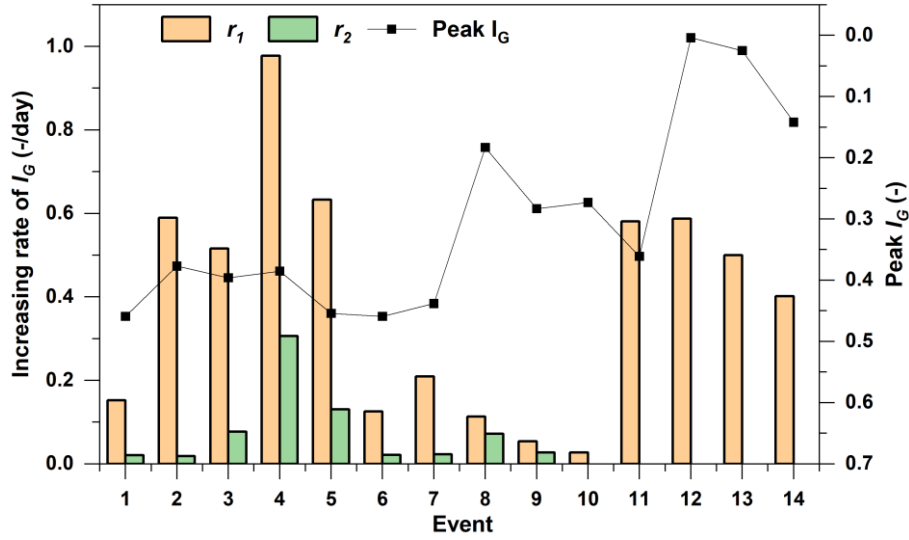
349 rapidly following rainfall, initiating 0-2 h after the SWC began to rise and reaching its peak occurring
 350 0-9 h later (mean: 3.7 h) after SWC reached its maximum. In contrast, the second I_G peak typically
 351 occurred post-rainfall, lagging behind the SWC peak by 10-65 h (mean: 28 h). These patterns align
 352 with the quick and slow GWL response types identified in section 3.2. The occurrence of dual I_G
 353 peaks can be attributed to the superimposition of groundwater contributions from different hillslopes
 354 with differing response rates. The first (quick) GWL response is tightly coupled to rainfall onset and
 355 SWC increases, while the second (slow) GWL response reflects gradual infiltration and groundwater
 356 recharge occurring over a broader timescale.



358 **Figure 10.** Delay time of I_G peaks relative to peak SWC. TI_{G01} and TI_{G02} represent the onset times
 359 of the first and second peaks of I_G , respectively. TS_0 and TS_p indicate the time when SWC started to
 360 increase and peaked, respectively. TI_{Gp1} and TI_{Gp2} represent the time when I_G started to increase and
 361 peaked, respectively. TR_e indicates the end of rainfall.

362 The growth rates of I_G towards the two peaks were also quantified (Fig. 11). A notable disparity
 363 was observed between the growth rates of the first (r_1) and second (r_2) I_G peaks. The first I_G peak
 364 exhibited a markedly faster rates (0.03 to 0.98/day, mean: 0.38/day) compared to the second peak
 365 (0.01 to 0.31/day, mean: 0.07/day). These differences reflect the contrasting dynamics of quick and
 366 slow GWL responses across hillslopes. In events featuring dual I_G peaks, the maximum I_G was

367 typically observed at the second peak. However, in events with higher GWLs (indicating lower I_G),
 368 the disparity between the growth rates diminished, making the two peaks harder to distinguish (e.g.,
 369 Events 9 and 10). In Events 11-14, where GWLs were significantly higher, only a single I_G peak was
 370 observed.



371
 372 **Figure 11.** Growth rates of I_G and the maximum I_G value across storm events. r_1 and r_2 denote the
 373 ascent rates during the first and second peaks, respectively.

374 The contrasting dynamics of the two I_G peaks highlight their distinct formation mechanisms.
 375 The first I_G peak, occurring during rainfall, is closely associated with the rapid rise in SWC.
 376 Conversely, the second I_G peak emerges post-rainfall, coinciding with soil draining and groundwater
 377 recharge processes. As reported by Dang et al. (2023), rainfall induces pressure waves that rapidly
 378 expel soil water from the lower soil column, while infiltrated rainwater migrates downwards at slower
 379 pace. pressure-driven displacement generates a near-instantaneous GWL response during the initial
 380 phase of rainfall.

381 We conjecture that the rapid I_G peak is linked to kinematic wave triggered by increased SWC,
 382 which displaces pre-existing "old" soil water and groundwater, leading to a synchronized GWL rise
 383 (e.g., Anderson and Burt, 1978). Despite the slow percolation of water through soil and bedrock, the
 384 theoretical celerity of this kinematic response is near-instantaneous, accounting for the rapid GWL
 385 rise. Furthermore, drilling data suggest that the presence of faults in the bedrock of HS2, potentially

386 facilitating faster groundwater response on this hillslope compared to others.

387 The second, slow I_G peak is likely driven by the gradual infiltration of rainwater into deeper soil
388 and bedrock layers, ultimately recharging the groundwater. This process is regulated by the soil's
389 water storage threshold. Before reaching this threshold, the soil retains all incoming rainfall. Once
390 exceeded, excess water drains rapidly into deeper layers, leading to a decline in SWC and a concurrent
391 rise in GWL due to groundwater recharge.

392 **4. Discussion**

393 **4.1 Inter-hillslope GWL dynamics**

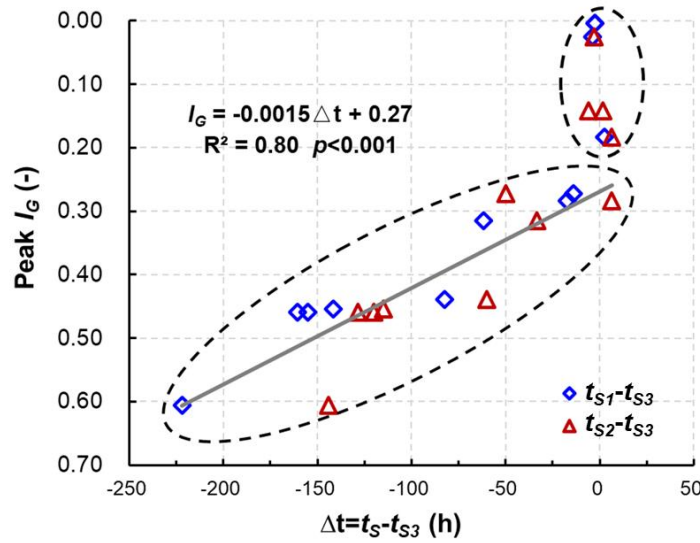
394 GWL variations in lag times and response magnitudes across hillslopes can be attributed to
395 differences in geological conditions. HS1 and HS3 are primarily underlain by fully to strongly
396 weathered granite, with upper layers comprising significant soil-rock mixtures. These features lead
397 to relatively slower GWL responses, likely due to the limited permeability of the regolith and
398 underlying materials. In contrast, HS2 is characterized by a fractured rock layer at depths of 10-30
399 meters (see Fig. 1), which enhances subsurface flow and facilitates faster GWL responses. These
400 geological contrasts explain the observed differences in GWL response times among the hillslopes.

401 Among the three hillslopes, HS3 exhibited the slowest GWL responses, characterized by the
402 longest lag times. This distinct behavior makes HS3 a crucial reference for understanding inter-
403 hillslope variations in GWL dynamics. Previous study by Cui et al. (2024) highlighted that GWL
404 response times are closely linked to delayed stormflow timing, emphasizing the importance of
405 examining GWL dynamics. Comparing the GWL response times of HS1 and HS2 with those of HS3
406 provides insights into how geological structures and SWC thresholds influence delayed stormflow
407 generation.

408 Furthermore, the deeply weathered regolith and extensive fracturing in HS2 promote more rapid
409 stormflow generation, as water stored in the regolith layer contributes to streamflow over extended
410 periods. This finding aligns with previous studies (Kosugi et al., 2011; Padilla et al., 2015), which

411 demonstrated that geological features such as fracture density and weathering depth influence
 412 subsurface flow paths and, ultimately, groundwater dynamics.

413 To deepen understanding of the inter-hillslope differences in GWL responses, we calculated the
 414 lag times between rainfall onset and peak GWL responses for all observation wells on each hillslope,
 415 incorporating spatial variability. Average lag times, denoted as t_{S1} , t_{S2} and t_{S3} for HS1, HS2, and HS3,
 416 respectively, were used to calculate the time differences $\Delta t = t_{S1} - t_{S3}$ and $\Delta t = t_{S2} - t_{S3}$. These time
 417 differences were then analyzed for their correlation with I_G , as illustrated in Fig. 12.



418
 419 **Figure 12.** Correlation between peak I_G and the time differences from peak GWL responses on
 420 HS1, and HS2 to HS3 ($\Delta t = t_{S1} - t_{S3}$), where t_{S1} , t_{S2} and t_{S3} are the average lag times of peak GWLs on
 421 HS1, HS2 and HS3, respectively.

422 In Fig. 12, blue diamonds represent $\Delta t = t_{S1} - t_{S3}$, while red triangles represent $\Delta t = t_{S2} - t_{S3}$. Both
 423 pairs exhibit a significant negative correlation with peak I_G , described by the equation: $I_G = -$
 424 $0.0015 \times \Delta t + 0.27$ ($R^2 = 0.80$, $p < 0.001$). These results indicate that higher I_G values correspond to
 425 shorter inter-hillslope lag times, suggesting enhanced hydrological connectivity and transmissivity
 426 feedback mechanisms, as described in previous studies (Kendall et al., 1999; Bishop et al., 2011).

427 As peak I_G approaches 0.30, Δt converges to near-zero with minimal fluctuations, particularly
 428 during extreme storm events. This finding supports the results presented in Fig. 11, which demonstrate

429 that elevated GWLs synchronize GWL responses across the watershed. This synchronization reflects
430 a critical hydrological mechanism driven by transmissivity feedback, which amplifies groundwater
431 movement, reduces lag times, and enhances watershed-scale connectivity. This dynamic is consistent
432 with the work of Padilla et al. (2015), who reported that shorter lag times in bedrock aquifers with
433 high-transmissivity conduits, and Scaife et al. (2020), who noted that increased connectivity during
434 high GWL conditions reduces lag times and enhances watershed-scale hydrological responses.

435 Furthermore, although Fig. 12 labels the vertical axis as I_G to represent watershed-wide GWL
436 status, a similar pattern emerges when replacing I_G with site specific GWL values, though the GWL
437 thresholds may vary among observation sites. These observations reinforce the idea that watershed-
438 scale groundwater dynamics are influenced by the interplay between spatially variable geological
439 conditions and temporal variations in GWL.

440 **4.2 Delayed stormflow processes linked to GWL dynamics**

441 Previous studies have shown that streamflow in XEW frequently exhibits a bimodal hydrograph
442 during heavy rainfall, with delayed stormflow likely originating from shallow groundwater outflow
443 (Cui et al., 2024). Understanding the timing and lag between groundwater and streamflow responses
444 is crucial for identifying dominant runoff generation mechanisms (Beiter et al., 2020). Discrepancies
445 in these timings can indicate contributions from different water sources to the stream channel. Fig.
446 13 illustrates the relative timing of maximum I_G (I_{Gp}) and maximum SWC (SWC_p) for eight storm
447 events, alongside rainfall duration.

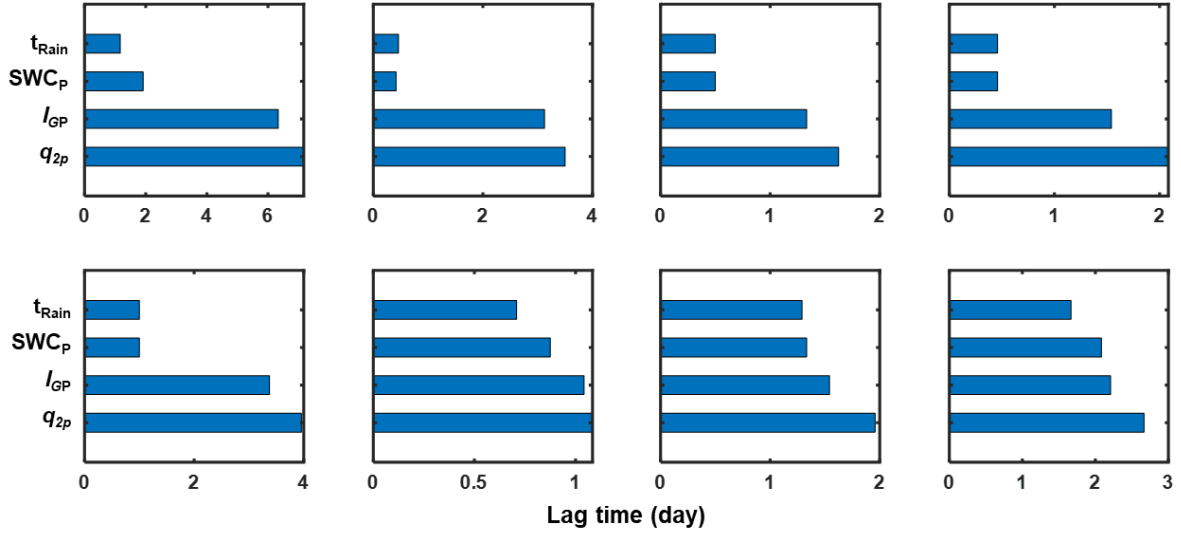


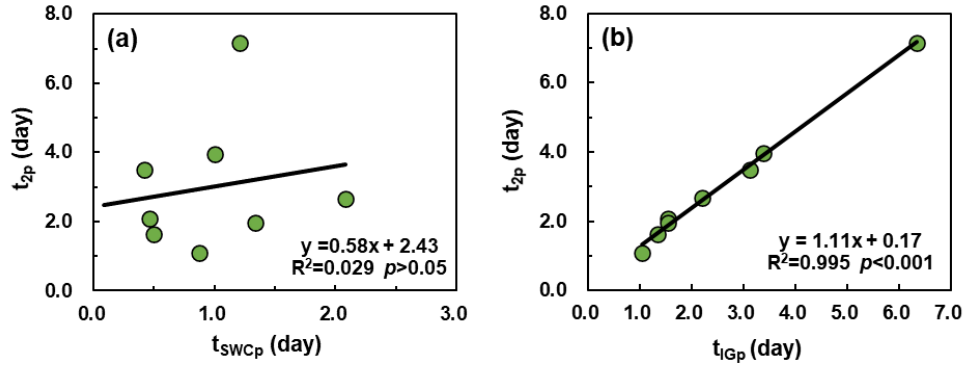
Figure 13. Lag times of maximum SWC and GWL relative to rainfall onset. Each bar indicates the rise and peak times of the corresponding variable, with t_{Rain} indicating rainfall duration. SWC_{SP} and I_{GP} represent the maximum SWC and I_{G} , respectively, while q_{2p} denotes the delayed streamflow peak.

Rainfall durations for the analyzed events ranged from 0.46 to 1.67 days. SWC, I_{G} , and delayed stormflow (q_{2p}) followed a clear sequence in their peak timings relative to rainfall onset. SWC responded rapidly, with its peak occurring 0.4 to 2.1 days after rainfall began, usually coinciding with or slightly after rainfall cessation. In contrast, I_{G} continued to increase after the SWC peak and reached its maximum before the delayed stormflow peak (q_{2p}). While the lag times between SWC_{p} , I_{Gp} , and q_{2p} varied among events, the lag between I_{Gp} and q_{2p} remained relatively consistent.

This pattern aligns with findings from Haught and Meerveld (2011) and Rinderer et al. (2016), who suggest that when groundwater response precedes or synchronizes with streamflow, it indicates strong hillslope-stream connectivity, with groundwater serving as the primary driver of streamflow. Our results corroborate this view, showing that q_{2p} timing is predominantly governed by groundwater dynamics. This relationship is further validated by the strong linear correlation between the lag times of q_{2p} (t_{2p}) and I_{Gp} (t_{IGp}), as indicated by the regression equation $t_{2p} = 1.11 \times t_{\text{IGp}} + 0.17$, with a slope of 1.11, showing a high determination coefficient ($R^2 = 0.995$, $p < 0.01$). (Fig. 14).

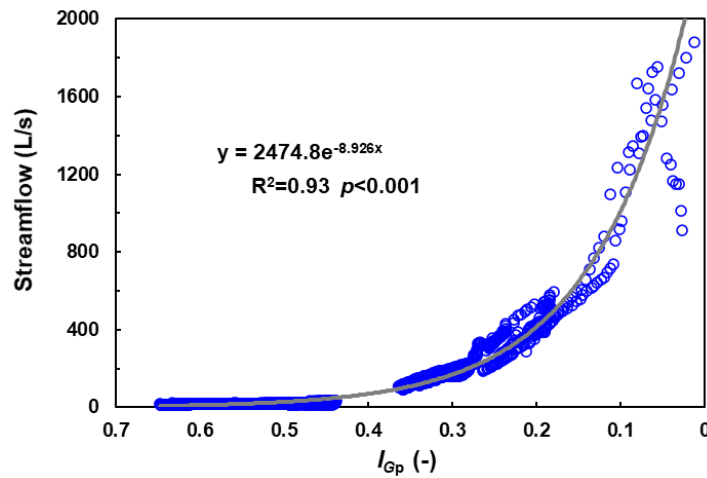
Conversely, the correlation between t_{2p} and the lag time of SWC_{p} (t_{SWCp}) was weak ($R^2 = 0.029$,

467 $p = 0.688$), indicating that the timing of SWC_p has minimal influence on the delayed streamflow peak.
 468 Additionally, the I_G pattern during the delayed stormflow period closely mirrored the shape of the
 469 streamflow hydrograph (Fig. A1), reinforcing the dominant role of I_G plays in controlling delayed
 470 stormflow.



471
 472 **Figure 14.** Lag times of maximum (a) SWC and (b) I_G relative to delayed streamflow peaks (t_{2p}).
 473 t_{SWCp} and t_{IGp} denote the peak times of the SWC and I_G , respectively.

474 Further quantitative analysis revealed a strong exponential relationship between streamflow and
 475 I_G during the delayed stormflow period (Fig. 15). In the non-rainfall phase of the eight bimodal events,
 476 streamflow increased exponentially with GWL (IG_p), exhibiting a highly significant correlation ($p <$
 477 0.001) and a determination coefficient of $R^2 = 0.90$. This exponential increase in streamflow is
 478 attributed to the increase in lateral hydraulic conductivity as the water table approaches the surface.
 479 Similar findings have been reported by Detty and McGuire (2010) and Kendall et al. (1999), where
 480 groundwater outflow dominates during storm events.



482 **Figure 15.** Correlation between I_G and streamflow during delayed stormflow periods.

483 At higher GWLs, the curve of GWL vs. streamflow begins to flatten, suggesting a feedback
484 mechanism. As the rising water table mobilizes shallow groundwater outflow, water is rapidly
485 transported to the stream via shallow flow paths. This process, often referred to as transmissivity
486 feedback, is consistent with Lundin's (1982) description of groundwater dynamics during delayed
487 stormflow periods.

488 **4.3 Delayed stormflow processes linked to GWL dynamics**

489 Understanding the critical thresholds that govern the movement of water within landscapes is
490 essential for accurately predicting delayed stormflow, as emphasized by McDonnell et al. (2021).
491 This study identified a strong Relationship between delayed stormflow and the gradual response of
492 GWL, primarily influenced by a sharp decline in SWC when it exceeds a critical threshold of 0.20.

493 To identify the threshold for delayed stormflow initiation in XEW, we analyzed 63 out of 95
494 rainfall-runoff events with complete streamflow data. The relationship between SWCp and q_s for
495 these events is illustrated in Fig. 16. A clear threshold behavior emerged: when SWC was below 0.20,
496 q_s remained minimal consistent with unimodal events. However, as SWC exceeded 0.20, a noticeable
497 increase in q_s was observed, signaling the onset of delayed stormflow in some events. Specifically,
498 when SWC surpassed 0.23, a pronounced surge in stormflow volume occurred, accompanied by the
499 emergence of a secondary stormflow peak in all events. These findings suggest that the critical
500 threshold for delayed stormflow initiation lies within the SWC range of 0.20 to 0.23.

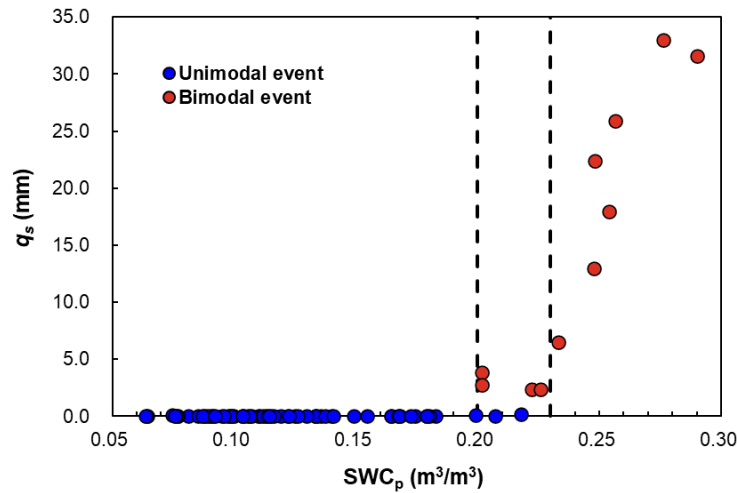


Figure 16. Relationship between maximum SWC (SWC_p) and event stormflow volume (q_s).

These results underscore the pivotal role of the surface soil layer's water deficit or water-holding capacity in triggering delayed stormflow. During rainfall events, the soil retains water until its water-holding capacity is exceeded. Once SWC surpasses the threshold of 0.20, the soil begins to release water more rapidly, initiating delayed stormflow. Additionally, during unimodal events, stormflow (q_s) remained consistently below 1 mm despite variations in SWC, indicating that stormflow in these cases arises mainly from direct rainfall interception by the channel rather than delayed soil water release.

While the depth and distribution of soil layers likely influence the watershed's overall water storage capacity, observed SWC data showed minimal spatial variability across locations within the watershed. This suggests that SWC can reliably represent the watershed's overall soil water storage capacity.

One limitation of this study lies in the indirect estimation of field capacity through observed SWC thresholds rather than direct measurement or modeling. Although this approach aligns with observed patterns and simplifies the analysis, it does not fully capture the spatial variability of field capacity or its dependence on soil depth. Future work should incorporate direct field capacity measurements or modeling to refine the relationship between SWC and delayed stormflow initiation, thereby improving the accuracy of threshold predictions.

520 **4.4 Conceptual model of runoff generation in XEW**

521 This section presents a conceptual model elucidating the runoff generation mechanisms in XEW,
522 with a particular focus on the interplay between soil water storage and GWL dynamics. Soil water
523 storage is identified as the critical factor driving the transition from initial to delayed runoff generation.
524 Once the soil water deficit is replenished, the slowly rising GWL becomes the dominant control in
525 the delayed stormflow process. Fig 17 illustrates the conceptual framework, which incorporates
526 transmissivity feedback mechanisms to explain the formation of distinct hydrograph patterns.

527 **a) Runoff generation under dry conditions (Fig. 17b):**

528 In dry watershed conditions, characterized by low antecedent moisture and light rainfall,
529 rainwater primarily infiltrates and is retained in the soil profiles. Streamflow during such events
530 consists of two primary components: (1) a rapid yet modest streamflow peak driven by direct rainfall
531 onto the channel and (2) a relatively stable baseflow originating from the gradual release of deep
532 groundwater reservoirs.

533 Under these conditions, the baseflow reflects the slow release of groundwater stored in deeper
534 aquifers, while the limited rainfall input is insufficient to trigger significant connectivity between
535 hillslopes and the channel.

536 **b) Delayed stormflow during moderate storms (Fig. 17c):**

537 As rainfall intensity and duration increase, moderate storm events lead to the replenishment of
538 soil water deficits, resulting in the exceedance of soil storage capacity. Initially, the response
539 resembles that of dry conditions, with a rapid streamflow peak generated by direct channel rainfall.
540 However, as rainfall continues, excess water infiltrates deeper, elevating the GWL and expanding the
541 saturated zone.

542 This process enhances the hydraulic connectivity between the stream channel and adjacent
543 hillslopes, facilitating the lateral transport of shallow groundwater to the channel. As the GWL
544 intersects more conductive soil layers, a delayed stormflow peak is observed, typically occurring after

545 rainfall ceases. This secondary peak reflects the combined effects of deep infiltration, gradual GWL
546 rise, and increased transmissivity in the subsurface, which accelerates shallow groundwater
547 movement towards the channel.

548 **c) Runoff generation during extreme storm events (Fig. 17d):**

549 Extreme storm events, characterized by high rainfall intensity and large volumes, result in a
550 sharp and widespread rise in GWL across the entire watershed. During such events, the rapid
551 expansion of saturated areas and the increased hydraulic conductivity of the subsurface enable the
552 swift mobilization of shallow groundwater. This synchronous response generates a pronounced flood
553 peak within a short timeframe.

554 In the riparian zones, GWLs may rise into the soil profile or even reach the ground surface,
555 facilitating direct water flow into the channel via subsurface pathways. Observational data from
556 extreme events corroborate this mechanism, as significant increases in SWC are recorded in the
557 deeper soil layers of riparian zones after rainfall ends. This pattern suggests that groundwater from
558 adjacent hillslopes contributes to the replenishment of soil water in these zones, reinforcing lateral
559 subsurface flow pathways.

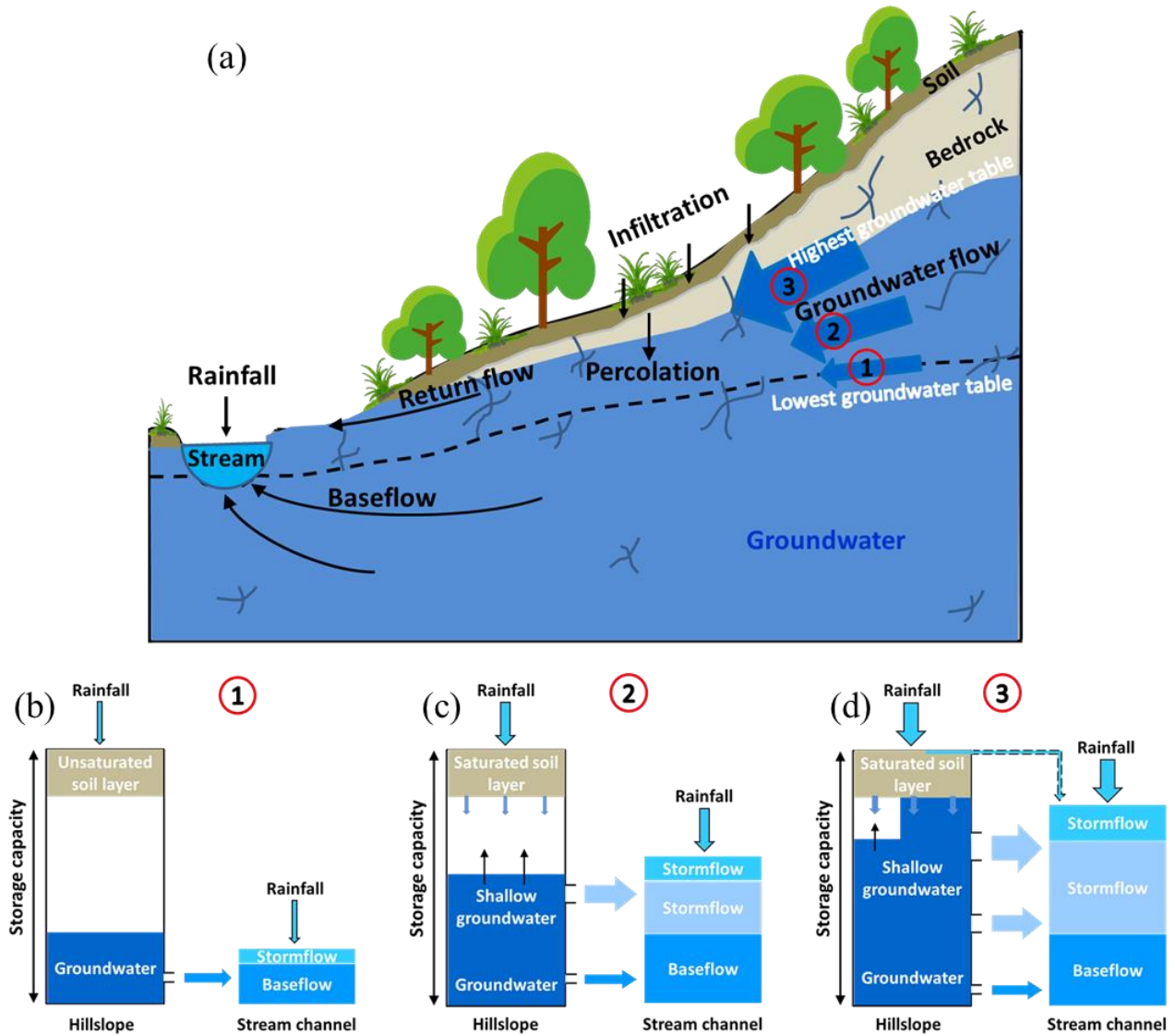


Figure 17. Conceptual model illustrating stormflow generation associated with the transmissivity feedback.

The progression from the runoff generation under dry conditions (Fig. 17b) to moderate storm scenarios (Fig. 17c) and ultimately extreme events (Fig. 17d) reflects the progressive wetting-up of the watershed. Abrupt changes in stormflow volume and timing are initially governed by soil water storage thresholds and subsequently controlled by the hydraulic conductivity of the bedrock and micro-topography.

This conceptual model provides a quantitative framework for understanding how variations in hydrological conditions influence runoff generation processes in XEW. By integrating soil water

570 storage dynamics, GWL responses, and transmissivity feedback mechanisms, the model offers
571 insights into the nonlinear behavior of runoff processes under different rainfall scenarios.

572 **5. Conclusions**

573 Building upon previous work that identified and characterized bimodal streamflow patterns in
574 XEW, this study quantitatively analyzed SWC and GWL dynamics at the event scale to elucidate the
575 mechanisms driving delayed stormflow generation. The findings reveal that when soil water storage
576 surpasses its holding capacity, a secondary increase in streamflow is triggered. This secondary, or
577 delayed, stormflow is primarily governed by GWL dynamics, which dictate both the magnitude of
578 the delayed response and the lag time to its peak.

579 During rainfall events, SWC responds rapidly, increasing until the soil's water storage capacity
580 is reached or exceeded. If the stored water remains within this capacity, SWC stabilizes or decreases
581 gradually following the cessation of rainfall, eventually leveling off near the field capacity. The rate
582 of this decrease is closely linked to the extent of SWC exceeding the field capacity. When SWC
583 begins to decline, excess rainwater percolates deeper into the soil, raising the GWL. Once GWL
584 begins to rise, it becomes the dominant driver of the delayed stormflow process.

585 As GWL rises, hydraulic conductivity increases, facilitating enhanced groundwater flux from
586 hillslopes to the stream channel. This process expands the effective connectivity between the channel
587 and adjacent hillslopes. At specific high GWL thresholds, the synchronization of GWL responses
588 across multiple hillslopes significantly amplifies stormflow volume. This synchronized response
589 shortens the lag time and increases the volume of delayed stormflow, often merging the delayed peak
590 with the direct stormflow peak.

591 These findings offer critical insights into the nonlinear processes governing stormflow
592 generation and the formation of bimodal hydrographs. By elucidating the mechanisms underpinning
593 these dynamics, the study advances hydrological theory and provides actionable knowledge for
594 improving flood modeling and prediction.

595 **Data availability**

596 The data supporting this study are available on the Zenodo website at
597 <https://doi.org/10.5281/zenodo.12581739>.

598 **Author contribution**

599 ZC contributed the conceptualization, formal analysis, investigation and writing; FT contributed
600 the conceptualization, formal analysis and revision.

601 **Competing interests**

602 Fuqiang Tian is a member of the editorial board of Hydrology and Earth System Sciences.

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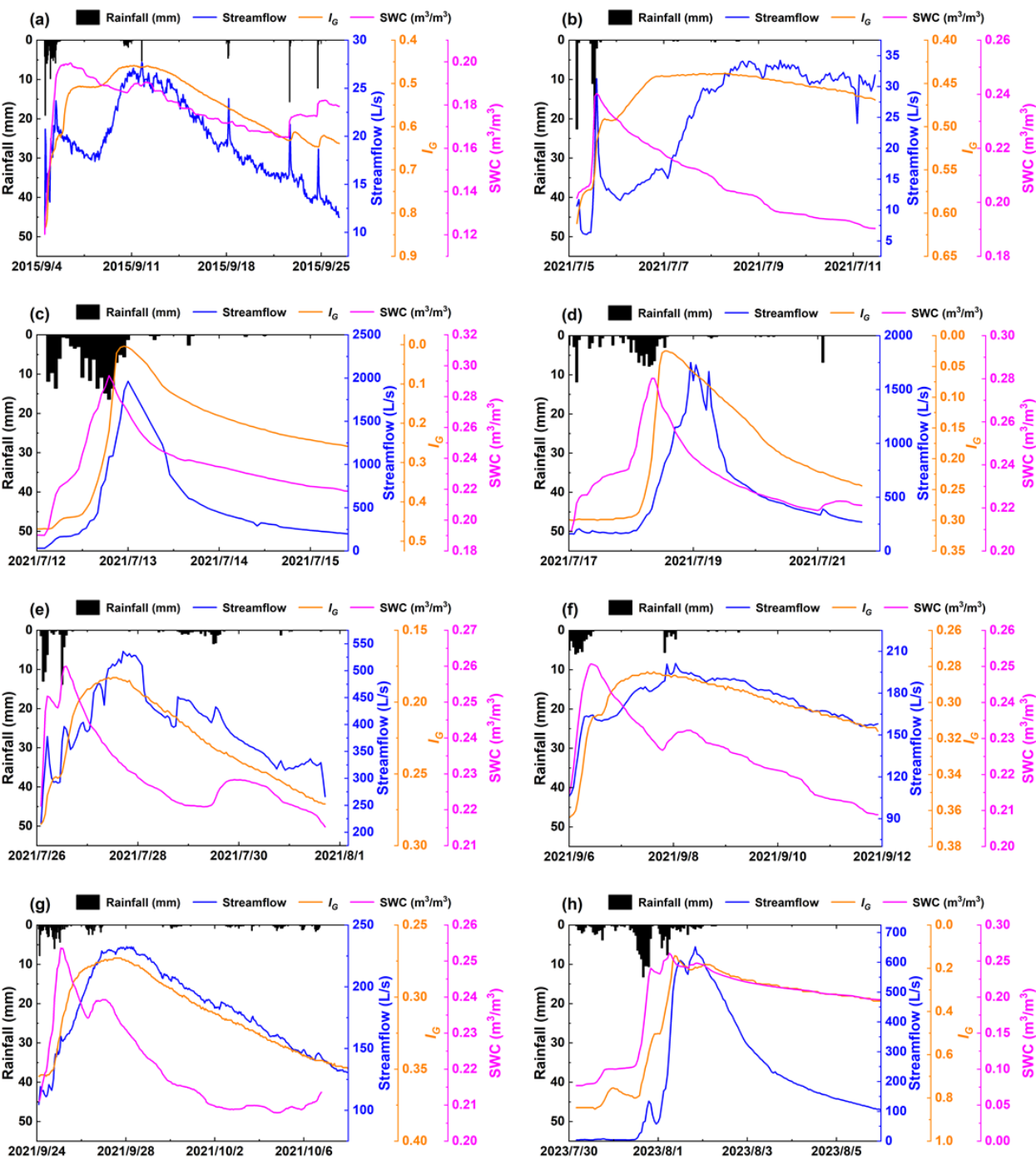
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676



678
679 **Figure A1.** Examples of responses of streamflow, I_G and soil water content to rainfall.