Seasonal variation and influence factors of river water isotopes in the East Asian monsoon region: A case study in Xiangjiang River basin spanning 13 hydrological years

Xiong Xiao¹, Xinping Zhang¹,², Zhuoyong Xiao¹, Zhiguo Rao¹, Xinguang He¹,², Cicheng Zhang¹

¹ College of Geographic Science, Hunan Normal University, Changsha 410081, China
² Key Laboratory of Geospatial Big Data Mining and Applications in Hunan Province, Hunan Normal University, Changsha 410081, China

ABSTRACT: Seasonal variation and influencing factors of water isotopes were investigated in the Xiangjiang River basin, located in the East Asian monsoon region. This involved comprehensive sampling of precipitation and river water, as well as observing hydrometeorological factors spanning 13 hydrological years from January 2010 to December 2022. Key findings are as follows: River water δ²H (δ²H_R) exhibited significant seasonal variation, with the most positive and negative δ²H_R occurring in the spring flood period and summer drought, respectively, and generally aligned with those observed in precipitation. The correlations of the δ²H_R with the corresponding hydrometeorological factors were generally weak and the reasons can be attributed to the seasonality of precipitation isotopes and mixing of various water bodies within the basin, but the changes in the runoff (ΔR) and δ²H_R (Δδ²H_R) between two contiguous samplings showed significant responses to the corresponding accumulated precipitation and evaporation. These results underscore the potential of Δδ²H_R as a variable that reflects the seasonal variations in local environments, valuable for paleoclimate reconstruction. Prolonged rainless intervals with high evaporation rates in 2013 and 2022, as well as significant precipitation events in
major flood periods in 2011 and 2017, notably had a significant impact on the $\delta^2$H$_R$ and runoff discharge. The most positive $\delta^2$H$_R$ values were primarily influenced by the precipitation input with the most enriched isotopes in the spring flood period, while the moderately isotope-depleted precipitation during limited basin wetness conditions led to the most negative $\delta^2$H$_R$, thus caution is advised when interpreting extreme isotopic signals in river water. The spatial correlation analysis between water isotopes and hydrometeorological factors at the observing site and in the surrounding regions supported the representation of the Changsha site in the Xiangjiang River basin. Overall, these findings provide insights into the seasonal variation and influencing factors of $\delta^2$H$_R$ in the study area, shedding light on the complex dynamics of river water isotopes under different hydrometeorological conditions.

**Keywords:** Stable isotopes; River water; Precipitation input; Evaporation; Seasonal variation.

1. **Introduction**

Stable isotopes of natural water possess exceptional sensitivity and serve as remarkable recorders of environmental change, for instance, water bodies undergo phase changes throughout the water cycle, resulting in stable isotope fractionation—an occurrence where light or heavy stable isotope molecules are distributed unequally between phases (Scholl et al., 2015; Xiao et al., 2022a). Generally, during water phase changes, light isotope molecules tend to evaporate more readily than heavy isotope molecules, while heavy isotope molecules preferentially condense compared to their lighter counterparts (Craig, 1961; Dansgaard, 1964). This isotope fractionation phenomenon contributes to variations in stable isotopic compositions among different water bodies within the water cycle. As a result, the relative ratios of stable isotopes in water serve as natural indicators of the
water cycle processes (Boral et al., 2019; Xiao et al., 2020; Wu et al., 2021) and find extensive application in hydrometeorology and meteorological diagnosis (e.g., Aggarwal et al., 2016; Sinha et al., 2019; Zhiňa et al., 2022) and paleoclimate reconstruction (e.g., Steinman et al., 2010; Jiménez-Iñiguez et al., 2022; Emmanouilidis et al., 2022). For instance, the stable river water isotopes primarily reflect the characteristics of precipitation, as precipitation input acts as the primary water source (Sprenger et al., 2022; Wang et al., 2023). Moreover, due to varying degrees of evaporation enrichment and mixing processes experienced by different water bodies within a basin, the river water isotopes markedly differ from that of the precipitation input and exhibit distinct seasonality (Jiang et al., 2021; Sun et al., 2021; Das and Rai, 2022). This disparity forms the basis for employing stable isotope techniques to investigate runoff generation processes in basins.

River water is commonly recognized as a natural integrator of basin hydrological processes, offering insights into the effects of hydrometeorological factors like air temperature, evaporation, precipitation input, and runoff discharge/water level (Yang et al., 2020; von Freyberg et al., 2022). Extensive efforts have been made to investigate the extent of variations in river water isotopes and examine the relationship between stable isotopes in river water and specific environmental factors (e.g., Yang et al., 2020; Das and Rai, 2022; Ren et al., 2023). However, in regions where new water mixes thoroughly with old water, the river water isotopes exhibit dampened signals, indicating that old water dominates the composition of stream water and that the response of river water isotopes to hydrometeorological factors is sluggish (Munoz-Villers and McDonnell, 2012; Streletskiy et al., 2015). Furthermore, extreme precipitation and drought events have become more frequent under the background of global climate changes, as evidenced in numerous regions worldwide.
These events introduce additional complexities and intricate seasonality in river stable isotopes on a basin scale, thus the identification of the controlling factors that influence river water isotopes becomes challenging (Uchiyama et al., 2017; Boutt et al., 2019; Saranya et al., 2020). The East Asian monsoon region, characterized by complex water vapor sources, substantial seasonal and inter-annual temperature and precipitation variations, as well as frequent floods and seasonal droughts, further contributes to the hydrological complexity in this region (Huang et al., 1998; Zhou et al., 2019; Wang et al., 2023). Hence, long-term observations of river water isotopes and hydrometeorological factors, along with comprehensive analyses of the influencing factors, are crucial to enhance our understanding of how climate change impacts hydrological regimes in basins within the East Asian monsoon region.

Long-term observations of water isotopes are crucial as they enable the capture of extreme precipitation and drought events, facilitating an analysis of their influences on river water isotopes, while also unveiling patterns of seasonal and inter-annual variation (Rode et al., 2016; von Freyberg et al., 2022; Ren et al., 2023). However, the long-term observations of river water isotopes with a high sampling frequency are relatively challenging and rare due to logistical constraints (von Freyberg et al., 2017). Therefore, in this study, the Xiangjiang River basin was selected as the study area to investigate the seasonal variation and controlling factors of river water isotopes under the influence of the monsoon. Extensive sampling of river water and precipitation, along with the monitoring of hydrometeorological factors, was conducted over 13 complete hydrological years from January 2010 to December 2022. This study aims to achieve the following objectives: (1) Identify the factors influencing the seasonality of
river water isotopes; (2) Assess the influences of extreme drought and precipitation events on river water isotopes; (3) Interpret the environmental significance implied by the seasonality of river water isotopes.

2. Study site

The study area was situated in a typical East Asian monsoon region, characterized by distinct climatic variations throughout the four seasons. On average, the annual precipitation and evaporation were 1147 mm and 902 mm, respectively (Xiao et al., 2022b). Notably, there is a significant seasonal disparity in precipitation, with an average of 152 rainy days each year. The period from early March to mid-July experiences abundant precipitation due to the influence of the monsoon, whereas from mid-July to September, drought conditions prevail as a result of the subtropical high-pressure system. The average annual temperature in the region is 17.4 °C, and the duration of the plant growing period spans approximately 330 days.

The Xiangjiang River basin, originating in the Guangxi Zhuang Autonomous Region, encompasses a drainage area of 94,660 km². It stretches for 856 km northward across Hunan Province, passing through cities like Hengyang, Xiangtan, Zhuzhou, and Changsha (Fig. 1). The altitudinal range of the Xiangjiang River basin varies from 1902 to 10 m, with higher elevations in the southern region characterized by multiple terraces and valley landforms, while the northern part is relatively lower. The middle and lower reaches of the Xiangjiang River basin are predominantly hilly basins, surrounded by the Xuefeng Mountains and Nanling Mountains (Fig. 1).
3. Methods and materials

3.1 Water samples collection and analysis

From January 2010 to December 2022, the collection of river water and precipitation samples was conducted.

River water sampling took place 929 times at the center of Orange Island, with a regular sampling interval of five days. Specifically, samples were obtained on the 1st, 6th, 11th, 16th, 21st, and 26th days of each month. The sampling depth of river water was relatively deep to avoid the influences of human activity; moreover, this operation can avoid the evaporation fractionation of surface water during the sampling and ensure adequate mixes of the river water.

At the College of Geographic Science, Hunan Normal University, Changsha (Fig. 1).
(1, 112°56′28″ E, 28°11′30″ N), the sampling of precipitation and the measurements of precipitation amount were conducted at the altitude of 55 m, adjacent to the Yuelu Mountains, throughout the sampling period of this study (2010–2022). For precipitation sampling, a siphon rain gauge was repurposed as a collector, consisting of a 6-cm diameter funnel connected to a glass bottle via a plastic pipe. Both rainfall and snowfall were collected and measured at 8:00 and 20:00 local time on the precipitation day, and the volume-weighted value of the two samplings was used to represent the precipitation isotopic values of the day, while the precipitation amount of the day was calculated by the sum of the two samplings. Snowfall samples were carefully packed in sealed plastic bags, which were later melted at room temperature. A total of 1668 precipitation isotopic values were obtained over 1668 precipitation days. Furthermore, considering that the sampling interval for river water was set at five days, and previous studies have indicated that it may take 3-5 days for precipitation (new water) to significantly contribute to river water (Yao et al., 2016; Xiao et al., 2022a), the precipitation isotopes were volume-weighted in the 5-day interval within this study.

To ensure proper preservation, both the river water and precipitation samples were transferred to clean, sealed, polyethylene bottles (30 ml) and stored in a refrigerator at 0 °C. However, few precipitation and river water samples were lost, resulting in some missing data. Further details regarding the sample collection procedures can be found in Xiao et al. (2022a). The isotopic composition of the samples was determined using the off-axis integrated cavity output spectroscopy method, specifically conducted with equipment from Los Gatos Research in the USA. The stable isotopic composition in the water samples is reported in ‰ (per mil). For a comprehensive description of the analytical procedures employed, please refer to the
detailed account provided by Xiao et al. (2022b).

3.2 Hydrometeorological observations

The daily air temperature and evaporation data used in this study were obtained from the National Meteorological Reference Station in Changsha (station code: 57687), specifically utilizing the large evaporator model E-601B. It is worth noting that the evaporation recorded by the E-601B evaporator closely approximates the actual evaporation experienced in small water bodies such as lakes and rivers. As such, it reliably represents the quantity and temporal variations of evaporation within the study area (Hua et al., 2019).

Daily runoff discharge data were obtained from the Xiangtan hydrological station (station code: 61102000). Following the national standard “Code for liquid flow measurement in open channels” (GB 50179-93, 1993) issued by the Ministry of Water Resources of the People’s Republic of China, the daily discharge values are calculated by applying the weighted average of intraday instantaneous discharge. This calculation is based on water level observations and a specific stage-discharge curve. As per the guidelines outlined in GB 50179-93, the relative errors in instantaneous discharge measurements range from 2% for high water levels, 5% for normal water levels, and 9% for low water levels. On average, the estimated annual discharge falls within 5% of the actual value. To facilitate the comparison of runoff discharge with precipitation and evaporation, the daily runoff discharge data (in m$^3$/d) are normalized by dividing them by the basin area (in m$^2$) of the measuring cross-section. Consequently, the runoff depth (in m/d or mm/d) and runoff discharge data (in m$^3$/d) are computed and utilized in the subsequent analysis.

4. Results

4.1 Stable isotopic characteristics of precipitation and river water
Table 1 presents the monthly average values of air temperature (°C), precipitation (mm), evaporation (mm), and runoff discharge ($10^8$ m³), and the results illustrate the uneven distribution of these factors throughout the year. Based on the monthly patterns of these hydrometeorological factors (Table 1) and the previous findings (Qin et al., 2006; Yao et al., 2016), four distinct runoff periods have been identified: the rainless period, spring flood period, major flood period, and summer drought period. The rainless period spans from October to the following February, characterized by low air temperature, minimal precipitation, evaporation, and runoff discharge. In this period, the runoff discharge exhibits slight fluctuations except for isolated peaks resulting from major rainfall events; The spring flood period, occurring in March and April, marks an increase in runoff discharge, this can be attributed to relatively higher precipitation amounts, as well as moderate air temperature and evaporation in spring; The major flood period, extending from May to mid-July, exhibits a rapid surge in runoff discharge, often reaching peak values due to intensive precipitation events; The summer drought period, which spans from mid-July to September, while a significant decrease in runoff discharge is always observed. This decline can be attributed to high air temperature leading to increased evaporation in this period. Additionally, the scarcity of precipitation events and relatively low precipitation amounts contribute to the reduced runoff discharge in this period. Overall, the analysis highlights the seasonal variations in air temperature, precipitation, evaporation, and runoff discharge, leading to distinct runoff periods throughout the year.
Table 1. Monthly average air temperature (℃), precipitation (mm), evaporation (mm), and runoff discharge ($10^8$ m$^3$)

<table>
<thead>
<tr>
<th>Month</th>
<th>Air Temperature</th>
<th>Precipitation</th>
<th>Evaporation</th>
<th>Runoff discharge</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>℃</td>
<td>mm</td>
<td>mm</td>
<td>$\times 10^8$ m$^3$</td>
</tr>
<tr>
<td>Jan.</td>
<td>5.6</td>
<td>68.9</td>
<td>33.0</td>
<td>35.5</td>
</tr>
<tr>
<td>Feb.</td>
<td>7.6</td>
<td>82.3</td>
<td>33.8</td>
<td>35.7</td>
</tr>
<tr>
<td>March</td>
<td>12.7</td>
<td>150.0</td>
<td>46.4</td>
<td>65.1</td>
</tr>
<tr>
<td>April</td>
<td>17.8</td>
<td>145.6</td>
<td>60.8</td>
<td>85.5</td>
</tr>
<tr>
<td>May</td>
<td>22.2</td>
<td>230.6</td>
<td>71.9</td>
<td>111.7</td>
</tr>
<tr>
<td>June</td>
<td>26.1</td>
<td>212.3</td>
<td>80.6</td>
<td>115.0</td>
</tr>
<tr>
<td>July</td>
<td>29.4</td>
<td>156.7</td>
<td>132.4</td>
<td>71.9</td>
</tr>
<tr>
<td>Aug.</td>
<td>28.9</td>
<td>87.6</td>
<td>141.7</td>
<td>36.0</td>
</tr>
<tr>
<td>Sep.</td>
<td>24.8</td>
<td>84.2</td>
<td>105.3</td>
<td>26.8</td>
</tr>
<tr>
<td>Oct.</td>
<td>18.7</td>
<td>61.2</td>
<td>83.9</td>
<td>22.0</td>
</tr>
<tr>
<td>Nov.</td>
<td>13.6</td>
<td>82.7</td>
<td>53.9</td>
<td>38.1</td>
</tr>
<tr>
<td>Dec.</td>
<td>7.5</td>
<td>54.6</td>
<td>47.5</td>
<td>31.8</td>
</tr>
</tbody>
</table>

Because the $\delta^2$H and $\delta^{18}$O of river water and precipitation samples were very similar, for the aims to keep with the previous analysis of the Xiangjiang River water isotopes (i.e. Xiao et al., 2022a) the temporal variations of $\delta^2$H values are mainly discussed in this paper. From 2010 to 2022, the 5-day volume-weighted precipitation $\delta^2$H ($\delta^2$H$^p$) exhibited large seasonality throughout the year, as depicted in Fig. 2 and Fig. 3. Notably, the 5-day volume-weighted $\delta^2$H$^p$ ranged from $-133.0‰$ to $39.1‰$, with a standard deviation of $27.6‰$. The 5-day volume-weighted $\delta^2$H$^p$ showed the maximum values in March, while the minimum values occurred in September (Fig. 2c). Furthermore, the 5-day volume-weighted $\delta^2$H$^p$ followed the following order: spring flood period > rainless period > major flood period > summer drought period (Fig. 3b). The seasonal variations in the precipitation isotopes primarily result from different vapor sources, upstream effects, circulation patterns, and local meteorological factors in different seasons (Zhou et al., 2019; Xiao et al., 2023). The $\delta^2$H values of river water ($\delta^2$H$^r$) ranged from $-63.7‰$ to $-21.7‰$, with a standard deviation of $6.1‰$ (Figs. 2 and 3).
Figure 2. Temporal variation of the daily runoff discharge and precipitation (a), daily air temperature and evaporation (b), daily volume-weighted precipitation $\delta^{2}H$ (c), and river water $\delta^{2}H$ with 5-day interval (d).
Figure 3. Stable isotope composition (δ²H and δ¹⁸O) of daily volume-weighted precipitation and river water samples (a) and box plots of precipitation and river water δ²H (b) in different periods (rainless period, spring flood period, major flood period, and summer drought period). LMWL (black solid line) and RWL (red solid line) represent the local meteoric water line and river water line based on all the precipitation and river water isotopic data from January 2010 to December 2022, respectively. The letters (RL, SF, MF, and SD) in the x-axis represent the rainless period, spring flood period, major flood period, and summer drought period, respectively.

The magnitude of δ²Hₚ was ranked as follows: spring flood period, major flood period, rainless period, and summer drought period (Fig. 3b). The most depleted and enriched isotopic values in river water in the sampling period occurred on September 1, 2013, and May 21, 2014, respectively (Fig. 2). The seasonal variations in river water isotopes generally aligned with those observed in precipitation, indicating that river water directly derives from the precipitation input and is influenced by its isotopic composition. Moreover, the local meteoric water line (LMWL) and river
water line (RWL) have a relatively close slope of 8.3 and 8.05, respectively (Fig. 3a). However, the $\delta^2$HR exhibited a relatively smaller range of variation and more attenuated temporal pattern compared to the $\delta^2$HP. This is likely due to the processes that precipitation undergoes before recharging river water, such as evaporation and mixing with older waters, which significantly reduce the variability in $\delta^2$HR (Xiao et al., 2022a).

### 4.2 Relationship between river water isotopes and various hydrometeorological factors

The relationships between the $\delta^2$HR and the corresponding 5-day accumulated precipitation and evaporation and average runoff discharge and air temperature in different periods are illustrated in Fig. 4. Based on the relationship between the 5-day volume-weighted $\delta^2$HP and the corresponding accumulated precipitation in different runoff periods (Fig. S1; left panel), it is evident that an “amount effect” is observed in the precipitation isotopes across various runoff periods. Specifically, the increased precipitation amounts consistently result in more isotope-depleted precipitation. Furthermore, the $\delta^2$HR exhibits a positive correlation versus the corresponding 5-day volume-weighted $\delta^2$HP, with a correlation coefficient of 0.55 and $p < 0.001$ (Fig. S2). This suggests that precipitation input is the primary factor influencing $\delta^2$HR. However, as shown in Fig. 4, the $\delta^2$HR exhibited relatively weak correlations with the corresponding 5-day accumulated precipitations in the rainless period, major flood period, and summer drought period, as indicated by low correlation coefficient values and $p > 0.05$ (Fig. 4a, 4i, 4m). Although a positive correlation between $\delta^2$HR and the corresponding 5-day accumulated precipitations in the spring flood period, their correlation coefficients were not exceptionally high (Fig. 4e). The weak correlation between the $\delta^2$HR and precipitation amount or runoff discharge can be attributed to the...
expansive area and water reserves in the Xiangjiang River basin—that is, after precipitation falls in the basin and the new input precipitation within 5 days may influence the $\delta^2$HR to some extent, however, it tends to mix with old waters, such as groundwater, soil water, and river water consists of a high proportion of older water components, thereby attenuating the impact of precipitation input and predominantly shape the river water isotopes.

Figure 4. Relationships between the river water $\delta^2$H and the corresponding 5-day accumulated precipitation (the first column), 5-day average runoff discharge (the second column), 5-day average air temperature (the third column), and 5-day accumulated evaporation (the fourth column) in different periods, and the first raw (a, b, c, and d), the second raw (e, f, g, and g), the third raw (i, j, k, and l), and the forth raw (m, n, o, and p) represent the rainless period, spring flood period, major flood period, and summer drought period, respectively. The $\delta^2$HR exhibits a consistent relationship with the corresponding 5-day average.
air temperature and accumulated evaporation within each respective runoff period. However, positive or negative correlations may be observed in these relationships across the four runoff periods. Specifically, in the spring flood period and summer drought period, the $\delta^2$H$_R$ demonstrates either a significant ($p < 0.001$) or non-significant ($p > 0.05$) positive correlation with the corresponding 5-day average air temperature and accumulated evaporation (Fig. 4g-h and 4o-p). Conversely, in the rainless period and major flood period, a significant ($p < 0.001$ or $p < 0.01$) negative correlation is observed (Fig. 4c-d and 4k-i). The negative relationship between the $\delta^2$H$_R$ and the corresponding 5-day average air temperature or accumulated evaporation may seem counterintuitive, as river water isotopes typically become enriched with increasing air temperature and progressing evaporation (Gibson et al., 2016; Jiang et al., 2021). This discrepancy can be explained by considering the seasonality of the relationship between precipitation isotopes and air temperature in different runoff periods (Fig. S1; right panel). For instance, the 5-day volume-weighted $\delta^2$H$_P$ gradually decreases as the corresponding average air temperature increases in the major flood period, while increasing as the corresponding average air temperature decreases in the rainless period (Fig. 2 and 3). This would lead to the negative relationship between the 5-day volume-weighted $\delta^2$H$_P$ and the corresponding average air temperature in the rainless period and major flood period (Fig. S1b and S1f), subsequently resulting in the negative relationship between the $\delta^2$H$_R$ and the corresponding 5-day average air temperature in these two periods (Fig. 4c-d and 4k-i). Moreover, as there is strong consistency between evaporation and air temperature (Allen et al., 2005), this alignment causes evaporation to also exhibit a negative correlation with $\delta^2$H$_R$ in these two periods (Fig. 4). Therefore, in the major flood period and rainless period, the influences of air temperature and evaporation on
river water isotopes are somewhat masked by the seasonality of precipitation isotopes. In the spring flood period and summer drought period, the river water isotopes exhibited an enrichment trend with the increasing of the corresponding average air temperature and accumulated evaporation (Fig. 4g-h and 4o-p). However, these relationships may be somewhat misleading due to the positive correlation observed between the 5-day volume-weighted $\delta^2$H$_R$ and the corresponding average air temperature in the spring flood period ($p < 0.05$) and summer drought period ($p > 0.05$) (Fig. S1d and S1h). This suggests that the positive correlation between river water isotopes and air temperature or evaporation may also be influenced by the seasonality of precipitation isotopes. Moreover, in the spring flood period, the increases in precipitation amount and runoff discharge lead to more isotope-enriched river water, as indicated in Fig. 4e and 4f. This phenomenon can be attributed to the most enriched precipitation isotopes occurring in the spring flood period (Fig. 2c and 3b), as the precipitation amount and runoff discharge gradually increase in this period (Fig. 2a), it contributes to the positive relationship observed between the $\delta^2$H$_R$ and the corresponding accumulated precipitation or average runoff discharge. Consequently, based on the findings of this section, when interpreting the relationships between the river water isotopes and hydrometeorological factors, it is important to consider the uncertainty arising from the seasonality of precipitation isotopes.

4.3 Relationship between the changes in river water isotope and the precipitation and evaporation

As shown in Fig. 4, the relationship between the $\delta^2$H$_R$ and various factors may not be significant or may contradict common sense—that is, precipitation input and evaporation are likely the major driving factors that change the isotopic compositions of river water. For instance, the river water isotopes tend to be more depleted and the
slope of RWL tends to be high in heavy precipitation events, while the river water generally exhibits gradual enriched stable isotopes and lower evaporation line slope on rainless days with high air temperature and evaporation (Gibson et al., 2016; Yang et al., 2020; Jiang et al., 2021). Therefore, under the river water sample collection at a 5-day interval, it is essential to analyze the influences of the individual hydrometeorological factors on the changes in the $\delta^2$H$_R$ between two contiguous samplings ($\Delta\delta^2$H$_R$), which was calculated by the $\delta^2$H$_R$ differences between the $\delta^2$H$_R$ of the following and preceding samplings (i.e. $\Delta\delta^2$H$_R = \delta^2$H$_R(t) - \delta^2$H$_R(t-1)$).

The 5-day runoff depth changes (i.e. $\Delta R = R(t) - R(t-1)$) and $\Delta\delta^2$H$_R$ exhibited significant increases and decreases, respectively ($p < 0.001$), in response to the 5-day accumulated precipitation (Fig. 5a and 5b). These findings indicate that precipitation input is the primary factor influencing the variations in runoff discharge and river water isotopes. Furthermore, based on the intersections between the linear fitting line and the x-axis, the thresholds for precipitation amounts influencing the $\Delta\delta^2$H$_R$ and $\Delta R$ at 5-day interval were determined to be 19.0 mm and 19.4 mm, respectively. This suggests that heavier precipitation events are more likely to alter the river water isotopes. The weak correlation ($p > 0.05$) between the $\Delta\delta^2$H$_R$ and corresponding 5-day accumulated precipitation below the threshold value of 19.0 mm supports this observation (Fig. 5c). Conversely when the 5-day accumulated precipitation exceeded 19.0 mm, a significant ($p < 0.001$) negative correlation was observed between the $\Delta\delta^2$H$_R$ and the corresponding 5-day accumulated precipitation (Fig. 5d). In other words, greater 5-day accumulated precipitation led to more negative $\delta^2$H values in the subsequent river water sample, with a correlation coefficient of $-0.34$ and $p < 0.001$.

Therefore, it can be concluded that the variation in river water isotopes reflects the isotopic signal of the input precipitation, and the isotopic composition of river water
exhibits a significant “amount effect” by the precipitation input.

Figure 5. Relationship between the 5-day accumulated precipitation and the changes in 5-day runoff discharge ($\Delta R$) (a) and river water $\delta^2$H ($\Delta$δ$^2$H$_R$) (b), and the relationship between the $\Delta$δ$^2$H$_R$ and the corresponding 5-day accumulated precipitation $<$ 19.0 mm and $>$ 19.0 mm were shown in subplot (c) and (d), respectively.

Because of the confluence processes of the river water from upstream to downstream and the mixing processes between the new and old waters, river water may consist of a certain proportion of old water with a relatively long residence time (Xiao et al., 2022a), thus we analysis the relationship between the river water isotopes and the hydrometeorological factors at longer time interval. The relationships between the $\Delta$δ$^2$H$_R$ and corresponding accumulated evaporation at various time intervals (5-, 10-, 15-, 20-, 25-, and 30-day) were shown in Fig. 6. As there is strong consistency between evaporation and air temperature (Allen et al., 2005), the relationship between the $\Delta$δ$^2$H$_R$ and air temperature was not analyzed in this paper. The results revealed a weak and statistically non-significant correlation between the $\Delta$δ$^2$H$_R$ and
corresponding accumulated evaporation at the 5-, 10-, 15-, and 20-day intervals (Fig. 6a-d). However, a significant \( p < 0.05 \) negative correlation with relatively low correlation coefficients was observed at the 25- and 30-day intervals (Fig. 6e and 6f). The negative relationship between the \( \Delta \delta^2H_R \) and the corresponding accumulated evaporation (Fig. 6c-f) also contradicts the common sense that the river water generally becomes more enriched under the influences of evaporation. This may be due to the negative relationship between the \( \delta^2H_P/\delta^2H_R \) and the corresponding average air temperature/accumulated evaporation in the rainless period and major flood period (Fig. 2 and 3) as discussed earlier, besides, the effect of dilution precipitation input on river water isotopes (Fig. 5 and S1) and the relatively low air temperature and high relative humidity in the heavy precipitation events may be greater than the enrichment effect of evaporation (e.g., similar as the effect demonstrated in Xiao et al., 2022b).

Figure 6. Relationship between the changes in river water \( \delta^2H \) (\( \Delta \delta^2H_R \)) and the corresponding accumulated evaporation (E) at 5-day (a), 10-day (b), 15-day (c), 20-day (d), 25-day (e), and 30-day (f) time intervals.
To highlight the influences of evaporation on the river water isotopes and eliminate the influence of precipitation input, we analyzed the relationship between the $\Delta \delta^2H_R$ and the corresponding accumulated evaporation at different intervals (5-, 10-, 15-, 20-, 25-, and 30-day) without precipitation input (Fig. 7). The findings revealed a significant ($p < 0.01$ or $p < 0.001$) positive correlation between these variables at different time intervals. In other words, as the accumulated evaporation increased, the $\Delta \delta^2H_R$ also increased, with the correlation coefficient generally increasing with longer rainless intervals. Notably, the influence of greater evaporation on the $\Delta \delta^2H_R$ was particularly pronounced at the 30-day interval, exhibiting a high correlation coefficient of 0.35 and a $p$ value less than 0.001, while these intervals occurred exclusively in 2013 and 2022, which were characterized as very dry years with high evaporation and multiple 30-day intervals devoid of precipitation (Fig. 2 and S3).

![Figure 7. Relationship between the changes in river water $\delta^2H$ ($\Delta \delta^2H_R$) and the corresponding accumulated evaporation (E) at 5- (a), 10- (b), 15- (c), 20- (d), 25- (e), and 30-day (f) intervals without precipitation input.](https://doi.org/10.5194/hess-2023-174)
and 30- (f) rainless time intervals.

The decrease in runoff discharge or water level in the basin can be attributed to evaporation within the basin, as evaporation increases the outflow component in the river water balance, leading to a reduction in the amount of water flowing out of the basin. Additionally, as runoff discharge can increase due to precipitation input, we examined the relationship between the $\Delta \delta^2H_R$ and the corresponding changes in the runoff ($\Delta R$) at different intervals (5-, 10-, 15-, 20-, 25-, and 30-day) without precipitation input (Fig. 8). The results indicated a negative correlation between the $\Delta \delta^2H_R$ and the corresponding $\Delta R$ at the 10-, 15-, 20-, and 25-day intervals, demonstrating relatively high correlation coefficients around $-0.4$ and $p$ values less than 0.001. Considering the higher correlation coefficients observed between the $\Delta \delta^2H_R$ and the corresponding $\Delta R$ (Fig. 8) compared to those between the $\Delta \delta^2H_R$ and the corresponding accumulated evaporation (Fig. 7), it can be inferred that the $\Delta R$ serves as a suitable proxy to represent the effects of evaporation on river water isotopes. However, the correlation between the $\Delta \delta^2H_R$ and the corresponding $\Delta R$ was relatively weak at the 5-day intervals, with correlation coefficients of $-0.11$ (Fig. 8a), thus could be attributed to the short time interval and the limited impact of accumulated evaporation.
Figure 8. Relationship between the changes in river water δ^2H (Δδ^2H_R) and runoff depth (ΔR) at 5- (a), 10- (b), 15- (c), 20- (d), 25- (e), and 30- (f) rainless time intervals.

**4.4 Influences of extreme drought and precipitation events on the river water isotopes**

Analysis of the annual accumulated evaporation, average air temperature, accumulated precipitation, and average runoff discharge in the major flood period and summer drought period reveals that 2013 and 2022 experienced severe summer drought conditions (Fig. S3). These periods were characterized by exceptionally high temperatures and evaporation and low precipitation levels compared to the 13-year observations, particularly in 2022. The summer drought period of 2022 recorded only 16.3 mm of precipitation, the lowest among the 13-year observations and significantly lower than the highest precipitation recorded in the summer drought period of 2010 (i.e. 300.5 mm). Accumulated evaporation and average air temperature in the summer
drought period of 2022 reached 385.1 mm and 28.8 °C, second only to the 393.7 mm and 29.8 °C recorded in the summer drought period of 2013 (Fig. 3). Furthermore, the extreme drought events in 2013 and 2022 have been extensively reported, indicating widespread meteorological, hydrological, and soil droughts that pose a significant threat to water resources for domestic, agricultural, ecological, and human needs (Ma et al., 2022; Bonaldo et al., 2023). Therefore, this section primarily focuses on these two extreme drought processes in 2013 and 2022.

As depicted in Fig. 9, the period from late June to mid-August 2013 witnessed rare precipitation, high evaporation rates, and elevated air temperatures. Consequently, the runoff discharge gradually decreased from 2699 m³/s on June 30 to 415 m³/s on August 16. Simultaneously, the δ²H_R progressively increased from −42.0% to −34.8% (Fig. 9e). Similarly, from mid-July to early November 2022, only 72.6 mm of precipitation was recorded, resulting in a gradual decline in runoff discharge until early September. In this process, the δ²H_R increased from −53.1% on July 16 to −37.9% on September 6 (Fig. 9f). Subsequently, the Xiangjiang River maintained low runoff discharge and raised δ³H Δ levels until the end of December, and the δ³H Δ increased by up to 20.2‰ from −53.1% on July 16 to −32.9‰ on November 26, 2022 (Fig. 9f). These findings align with the results obtained in the previous section, indicating that decreases in runoff discharge and higher evaporation rates in long rainless days contribute to the gradual enrichment of river water isotopes. However, it is noteworthy that the δ³H Δ range (−63.7‰ to −21.7‰) mentioned earlier includes the most positive δ³H Δ values influenced by the extreme drought events in 2013 and 2022 (Fig. 2d), while the most isotope-enriched river water occurred on May 21, 2014 (Fig. 2d) and most isotope-enriched precipitation occurred in the spring flood period (Fig. 3b), this indicates that the input of relatively enriched spring precipitation isotopes
plays a crucial role in controlling the isotopic enrichment of river water.

By examining the 13-year observations and ranking the 5-day accumulated precipitation, it was found that the maximum accumulated precipitation, totaling 301.6 mm, occurred from June 27 to July 1, 2017 (Fig. 2). Notably, on June 30 and July 1, 2017, daily precipitation reached 146.4 mm and 130.3 mm, respectively.

Additionally, between June 21 and June 26, 2017, the precipitation amount reached 185.5 mm. Another significant precipitation event took place in June 2011, when the total precipitation for the month reached 340.3 mm, with a single-day rainfall of 110.4 mm on June 28. To analyze the impact of extreme precipitation on river water isotopes in the major flood period, the temperature, evaporation, precipitation, runoff discharge, δ²Hᵣ, and δ²Hᵢ in these two extreme precipitation processes are presented in Fig. 10.
Figure 10. Temporal variations of the evaporation and air temperature (a, b), precipitation amount and precipitation δ²H (c, d), and runoff discharge and river water δ²H (e, f) in the extreme precipitation processes and the following summer drought period in 2011 (left panel) and 2017 (right panel).

In the heavy precipitation period from June 5 to June 16, 2011, the runoff discharge increased from 1410 m³/s to 9010 m³/s, while the δ²HR decreased from −30.8‰ to −46.5‰ (Fig. 10e). Similarly, from June 21 to July 1, 2017, the runoff discharge and river water isotopes exhibited significant fluctuations and depletion under the influence of precipitation input (Fig. 10f). For instance, the runoff discharge increased from 3961 m³/s to 18237 m³/s, while the δ²HR decreased from −29.3‰ to −56.8‰, which represents the third lowest δ²HR value among the 13-year observations. In both extreme precipitation processes, the precipitation isotopes were relatively depleted (Fig. 10c and 10d). Specifically, the volume-weighted δ²HR was −50.9‰ from June 5 to June 16, 2011, and the value was −76.7‰ from June 21 to July 1, 2017. This indicates that the input of extreme precipitation leads to rapid decreases in δ²HR. For instance, in the extreme precipitation process of June 2011, the
δ²Hᵣ was reduced by −15.7‰ under a precipitation input of 144.7 mm from June 5 to June 16, while the δ²Hᵣ decreased by −27.5‰ with a precipitation input of 487.1 mm in the extreme precipitation process from June 21 to July 1, 2017. Subsequently, in the summer drought periods of 2011 and 2017, the river water isotope gradually enriched to approximately −38% (Fig. 10e and 10f).

5. Discussion

5.1 Factors that influence the seasonality in river water isotopes

The river water isotopes showed strong seasonality and were influenced by various factors such as precipitation input and evaporation. For instance, in the major flood period and summer drought period, the δ²Hᵣ usually reflected isotope-depleted precipitation inputs, indicating the “amount effect” by the precipitation input. Specifically, the precipitation isotopes were relatively depleted in these two periods, while the river water isotopes captured the precipitation input signal particularly when the 5-day accumulated precipitation exceeded 19.0 mm (i.e. the threshold precipitation amount) (Fig. 5). Additionally, the extreme precipitation events mainly occurred in the major flood period, resulting in relatively isotope-depleted precipitation that was reflected as negative records in the δ²Hᵣ (Fig. 2 and 10). In the summer drought period, the river water isotopes exhibited a gradual enrichment process due to the influence of evaporation and limited precipitation input (Fig. 2 and 9). In the rainless period, the δ²Hᵣ values were more positive compared to the summer drought period (Fig. 3b), possibly influenced by precipitation input, evaporation enrichment, and groundwater recharge. Furthermore, the δ²Hᵣ reached the highest positive values in the spring flood period, influenced by relatively isotope-enriched precipitation inputs as mentioned above (Fig. 2 and 3).

By examining the relationship between the Δδ²Hᵣ and the corresponding
accumulated evaporation, it becomes evident that a stronger correlation between the two variables emerges at time intervals exceeding 10 days (Fig. 6). This suggests that the influence of evaporation on river water isotopes manifests over relatively long time intervals, spanning from tens of days to even several months, particularly in long dry periods without precipitation input (Fig. 9). The influences of evaporation on river water isotopes occurs gradually, making it challenging to capture using short-term analyses. Conversely, as indicated by Xiao et al. (2022a), the relatively weak influence of evaporation on river water isotopes observed at short intervals (Fig. 6 and 7) can be attributed to the significant influx of precipitation input (i.e. new water) rapidly flowing into the river network, which experiences limited evaporation effects in the relatively short residence time. Furthermore, the variation in runoff discharge exhibits a notable relationship with the $\Delta \delta^{2}H_R$ (Fig. 8), however, it is important to note that the decline in runoff discharge may not be solely due to evaporation but could also be influenced by water transport from the Xiangjiang River to the Dongting Lake due to rapid downstream drainage (Zhan et al., 2015), this introduces uncertainties in the analysis based on the changes of runoff discharge and river water isotopes between different time intervals.

River water isotopes serve as a valuable record not only for detecting the isotopic depletion signal of extreme precipitation input in the major flood period (Fig. 10) but also for capturing the influence of moderate precipitation in the summer drought period. In the latter case, the river water isotopes gradually become enriched with the decrease of the runoff discharge in prolonged periods of drought. When the basin is relatively dry—that is, the reserves of soil water, groundwater, and river water are limited, the river water may be influenced by a medium precipitation amount, resulting in a highly depleted river water isotope signal observed in this extended river
water sample series. For instance, following a precipitation input of 53.5 mm with a δ²H_p of −68.1‰ on August 23, 2013 (Fig. 9c), the river water isotopes exhibited a significant depletion (Fig. 9e). Notably, the δ²H_R rapidly declined from −35.7‰ on August 21 to −63.7‰ on September 1, representing the most negative δ²H_R value observed over the 13-year observations. This δ²H_R signal differs significantly from rapid decreases in δ²H_R caused by extreme precipitation input in the river water (Fig. 10), highlighting the importance of careful consideration when reconstructing precipitation based on isotopic signals derived from river water. Overall, the seasonality of river water isotopes in the Xiangjiang River basin is influenced by various complex factors, including precipitation input, seasonal drought, and the basin wetness conditions, such as soil water, groundwater, and river water reserves within the channel system.

5.2 Environmental significance implied by the seasonality of river water isotopes

The Xiangjiang River, serving as a significant inflow water source, exerts influence on the hydrologic and isotope mass balance of Dongting Lake, the second-largest freshwater lake in China (Zhan et al., 2015; Zhou et al., 2019). The isotopic composition of lake water primarily reflects the input waters, including lake surface precipitation and inflowing river water (Steinman and Abbott, 2013; Gibson et al., 2016; Xiao et al., 2022b), while the isotopic information in lake water can also influence the stable isotopic signatures preserved in lake sediment. Consequently, proxy indicators recorded in lake sediments can be utilized for paleoclimate reconstruction, benefiting from the relationships between the input water isotopes and the local environments.

Through the analysis of river water isotopes and various hydrometeorological
factors on a seasonal scale, it becomes evident that the \( \Delta \delta^2H_R \) can reflect the corresponding accumulated evaporation and precipitation input (Fig. 5, 6, and 7) and the decline in runoff discharge (Fig. 8) at the observed time intervals. Moreover, river water isotopes entering the lakes can record signals of extreme precipitation (Fig. 10) or exhibit gradual isotopic enrichment under the influence of evaporation in relatively dry periods spanning tens of days or even several months without precipitation (Fig. 9). Besides, the isotopic characteristics of precipitation are governed by large-scale factors such as moisture sources, upstream effects, and circulation patterns, and are less influenced by local meteorological factors (Aggarwal et al., 2016; Zhou et al., 2019), thus the river water isotopes are better suited to reflect local environments. Consequently, in comparison to the isotopic characteristics of precipitation, the river water isotopes may provide valuable insights into the relationship between the proxy indicators and the local environments.

In previous studies, due to limited data availability, the inflow water isotopes of the lake were often represented by the volume-weighted precipitation isotopes in hydrologic and isotope mass-balance models (e.g., Steinman and Abbott, 2013; Skrzypek et al., 2015; Jones et al., 2016). However, based on the 13-year observations conducted in this study, the annual volume-weighted \( \delta^2H_R \) and \( \delta^2H_P \) were found to closely match only in the period of 2012-2015, with differences within 2‰ (Table 2). In other years, the annual volume-weighted \( \delta^2H_P \) was either more negative or more positive compared to the annual volume-weighted \( \delta^2H_R \). These differences in the annual \( \delta^2H_R \) and \( \delta^2H_P \) were mainly influenced by the seasonality of \( \delta^2H_P \) and precipitation amount. For instance, significant variations were observed between the volume-weighted \( \delta^2H_P \) and \( \delta^2H_R \) in the different runoff periods (Table 2). Therefore, representing the inflow water of the lake solely by the annual volume-weighted
precipitation isotopes can only serve as a rough estimation. To accurately depict the
detailed variations in the lake hydrologic and isotope mass balance, more
comprehensive observations of the inflowing river water are required.

Table 2. Annual volume-weighted precipitation $\delta^2$H and river water $\delta^2$H ($\delta^2$HR
and $\delta^2$HP) and volume-weighted average values in different runoff periods. The letters
(RL, SF, MF, and SD) represent the rainless period, spring flood period, major flood
period, and summer drought period, respectively.

| Year | Annual $\delta^2$HR in RL | Annual $\delta^2$HR in SF | $\delta^2$HR in MF | $\delta^2$HR in SD | $\delta^2$HR in RL | $\delta^2$HR in SF | $\delta^2$HR in MF | $\delta^2$HR in SD | $\delta^2$HP in RL | $\delta^2$HP in SF | $\delta^2$HP in MF | $\delta^2$HP in SD |
|------|----------------------------|----------------------------|--------------------|--------------------|-------------------|-------------------|--------------------|--------------------|-------------------|-------------------|--------------------|--------------------|-------------------|
| 2010 | -34.6                      | -48.0                      | -38.1              | -54.6              | -26.7             | -16.8             | -36.0              | -47.8              | -40.1             | -73.6             |                     |                    |
| 2011 | -36.9                      | -45.3                      | -37.4              | -52.0              | -30.3             | -11.4             | -39.0              | -54.3              | -38.9             | -49.3             |                     |                    |
| 2012 | -36.3                      | -38.2                      | -36.9              | -29.8              | -31.1             | -15.4             | -37.9              | -50.9              | -45.9             | -47.0             |                     |                    |
| 2013 | -33.9                      | -33.3                      | -35.6              | -72.1              | -27.1             | -3.7              | -30.2              | -28.3              | -48.0             | -47.7             |                     |                    |
| 2014 | -29.0                      | -27.7                      | -33.2              | -27.3              | -26.3             | -10.8             | -27.8              | -34.5              | -36.0             | -57.1             |                     |                    |
| 2015 | -34.4                      | -36.1                      | -38.4              | -43.9              | -27.3             | -10.6             | -30.1              | -34.4              | -40.6             | -64.0             |                     |                    |
| 2016 | -31.4                      | -40.0                      | -35.5              | -31.6              | -27.8             | -16.2             | -30.2              | -50.3              | -36.2             | -67.7             |                     |                    |
| 2017 | -36.4                      | -48.2                      | -35.0              | -26.6              | -27.4             | -14.5             | -41.2              | -59.3              | -39.4             | -75.4             |                     |                    |
| 2018 | -34.7                      | -39.1                      | -35.8              | -30.0              | -30.6             | -8.1              | -33.2              | -49.5              | -39.5             | -67.6             |                     |                    |
| 2019 | -31.8                      | -24.0                      | -30.8              | -25.3              | -26.4             | -2.5              | -34.2              | -31.2              | -36.4             | -28.7             |                     |                    |
| 2020 | -28.5                      | -36.9                      | -32.2              | -31.3              | -24.2             | -9.2              | -28.4              | -43.2              | -28.0             | -54.6             |                     |                    |
| 2021 | -30.1                      | -32.4                      | -36.3              | -31.1              | -28.3             | -12.1             | -28.2              | -32.4              | -32.6             | -66.5             |                     |                    |
| 2022 | -36.1                      | -39.2                      | -34.2              | -47.8              | -32.1             | -18.1             | -37.5              | -46.6              | -42.8             | -50.2             |                     |                    |

Nevertheless, according to the analysis on an annual scale based on the 13-year
observations, the volume-weighted $\delta^2$HR values in the different runoff periods did not
exhibit significant correlations with the corresponding total precipitation, average
runoff discharge, average air temperature, and total evaporation (Fig. S4). Although
this study encompasses 13-year observations with a sampling interval of five days,
there is a need for longer and systematic observations of various water types and
hydrometeorological factors, spanning decadal or longer time scales, to better
elucidate the relationships between river water isotopes and local environments on the
annual scale.

5.3 The representation of the Changsha site in the Xiangjiang River
The increases in runoff discharge and water level in the Changsha section of the Xiangjiang River are primarily attributed to the precipitation recharge in the middle and upper reaches, while the influences of evaporation on river water isotopes mainly occur within the basin. It should be noted that the sampling and observation sites for precipitation, air temperature, and evaporation in this study, located at Hunan Normal University and National Meteorological Reference Station in Changsha, respectively, only represent the local conditions in Changsha. Therefore, the extent of their influence on the runoff discharge and water stable isotopes of the entire Xiangjiang River may be limited. To assess whether the sampling and observations in Changsha can adequately represent the Xiangjiang River basin, a spatial correlation analysis was conducted between the Changsha site and the surrounding regions based on data from 1979 to 2021, including precipitation isotopes, precipitation amount, evaporation, and air temperature (Fig. S5). The analysis employed the simulated precipitation isotope data generated by the isotopic Atmospheric Water Balance Model (iAWBM) as detailed by Zhang et al. (2015), which has a spatial resolution of 1.5° × 1.5°, while the air temperature, evaporation, and precipitation amount data from the ERA5 reanalysis dataset (https://cds.climate.copernicus.eu) published by the European Centre for Medium-Range Weather Forecasts (ECMWF), which has a spatial resolution of 1° × 1°. Overall, all the data employed in this spatial correlation analysis was integrated into a 5-day interval.

The results of spatial correlation analysis revealed a high correlation between the reanalysis data of air temperature and evaporation at the Changsha site and those in the surrounding regions, with correlation coefficients above 0.8 and $p < 0.001$ for the grid points in the Xiangjiang River basin (Fig. S5a and S5b). Furthermore, while the
relationship between the reanalysis data of precipitation amount and simulated precipitation isotopes at the Changsha site and in the surrounding regions is not as strong as that for air temperature and evaporation, the correlation is still high (Fig. S5c and S5d). For instance, the correlation coefficients between the Changsha site and the grid points in the Xiangjiang River basin exceed 0.7 with \( p < 0.001 \) for both the reanalysis data of precipitation amount and simulated precipitation isotopes (Fig. S5c and S5d). Overall, the high correlation coefficients for these factors support the time consistency between the Changsha site and the surrounding regions, while the high spatial consistency is characterized by the large area of the high correlation coefficients, which covers the whole Xiangjiang River basin. These spatial correlation analyses provide support for the representation of hydrometeorological factors and precipitation isotopes at the Changsha site with the Xiangjiang River basin to a certain extent. In addition, the next step involves expanding the observation of air temperature, evaporation, and precipitation amount and the sampling of precipitation samples along the middle and upper reaches of the Xiangjiang River, this will help further validate the representativeness of the observations at the Changsha site for the entire Xiangjiang River basin.

6. Conclusions

The main findings of this study are as follows: (1) Both the \( \delta^2H_P \) and \( \delta^2H_R \) displayed significant seasonal variation throughout the year, the average \( \delta^2H_R \) was ranked as follows: spring flood period, major flood period, rainless period, and summer drought period. The temporal pattern of the \( \delta^2H_R \) was similar but attenuated compared to the \( \delta^2H_P \), indicating the influences of direct precipitation input, evaporation, and mixing with older waters. (2) The \( \delta^2H_R \) showed a weak correlation with the corresponding accumulated precipitation amount or average runoff discharge,
which can be attributed to the mixing between the new input precipitation and the old waters within the basin. The relationship between the $\delta^2$H$_R$ and the corresponding average air temperature or accumulated evaporation may be masked by the precipitation inputs, as the relationship between precipitation isotopes and air temperature exhibits strong seasonality. (3) The $\Delta R$ and $\Delta\delta^2$H$_R$ exhibited significant responses to the corresponding accumulated precipitation, with heavier precipitation events more likely to alter runoff discharge and river water isotopes. The $\Delta\delta^2$H$_R$ showed a notably positive correlation with the corresponding accumulated evaporation and average $\Delta R$, particularly at longer time intervals without precipitation input. (4) In the major flood period, the $\delta^2$H$_R$ may rapidly decrease with a maximum decrease range of $-27.5\%$ due to the input of extreme precipitation with relatively depleted isotopes, while the river water isotopes gradually enriched in the subsequent summer drought period. In the summer drought periods of 2013 and 2022, the runoff discharge and $\delta^2$H$_R$ showed gradual decrease and increase, respectively, with the $\delta^2$H$_R$ potentially increasing by up to $20.2\%$ influenced by the extreme drought.

The $\Delta\delta^2$H$_R$ at 5-day or longer intervals serves as a variable that was influenced by the hydrometeorological factors, such as the accumulated evaporation and precipitation input and the decline in runoff discharge, thus providing insights into the relationship between the river water isotopes and the seasonality of local environments, making it valuable for paleoclimate reconstruction. The influences of evaporation on river water isotopes occur gradually at relatively long time intervals. However, the spring flood period shows the most positive $\delta^2$H$_R$ due to the input of enriched precipitation isotopes. In contrast, in the summer drought period, when basin wetness conditions are limited, the input of moderate precipitation leads to the most negative $\delta^2$H$_R$ observed over the 13-year observations. Therefore, caution should be
exercised when using isotopic signals of river water to reconstruct local environments such as precipitation, air temperature, evaporation, and runoff discharge. The spatial correlation analysis confirms the association between the precipitation amount, evaporation, air temperature, and precipitation isotopes at the Changsha site with the Xiangjiang River basin. To enhance the reliability of the observations and ensure their representativeness for the entire Xiangjiang River basin, further comprehensive measurements of hydrometeorological factors and sampling of precipitation samples are needed within the basin itself.

**Code/Data availability**

The data that support the findings of this study are available from the corresponding author upon reasonable request.

**Author contribution**

Xiong Xiao: Data curation, Methodology, Software, Writing – original draft, & editing. Xinping Zhang: Methodology, Writing – original draft, & editing. Zhiguo Rao, Xinguang He, and Cicheng Zhang: Methodology.

**Competing interests**

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

**Acknowledgments**

This study was supported by the Natural Science Foundation of Hunan Province, China (No. 2023JJ40445) and the National Natural Science Foundation of China (No. 42101130). We are grateful to the graduate students who laboriously sampled water samples without interruption and tested water stable isotopes in the 13 hydrological years.

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