Trend Analysis of Annual and Seasonal River Runoff by Using Innovative Trend Analysis with Significant Test

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Abstract: This study investigated the temporal patterns of annual and seasonal river runoff data at 13 hydrological stations in the Lake Issyk-Kul basin, Central Asia. The temporal trends were analyzed using the innovative trend analysis (ITA) method with significance testing. The ITA method results were compared with the Mann-Kendall (MK) trend test at a 95% confidence level. The comparison results revealed that the ITA method could effectively identify the trends detected by the MK trend test. Specifically, the MK test found that the time series percentage decreased from 46.15% in the north to 25.64% in the south, while the ITA method revealed a similar rate of decrease, from 39.2% to 29.4%. According to the temporal distribution of the MK test, significantly increasing (decreasing) trends were observed in 5 (0), 6 (2), 4 (3), 8 (0), and 8 (1) time series in annual, spring, summer, autumn, and winter river runoff data. At the same time, the ITA method detected significant trends in 7 (1), 9 (3), 6 (3), 9 (3), and 8 (2) time series in the study area. As for the ITA method, the “peak” values of 24 time series (26.97%) exhibited increasing patterns, 25 time series (28.09%) displayed increasing patterns for “low” values, and 40 time series (44.94%) showed increasing patterns for “medium” values. According to the “low”, “medium”, and “peak” values, five time series (33.33%), seven time series (46.67%), and three time series (20%) manifested decreasing trends, respectively. These results detailed the patterns of annual and seasonal river runoff data series by evaluating “low”, “medium”, and “peak” values.

Keywords: river runoff; innovative trend analysis; Mann-Kendall test; significant test; Lake Issyk-Kul basin

1. Introduction

Climate change affects regional water resources [1]. Water resources are an essential foundation for society’s sustainable development, the economy, and the ecological environment [2–4]. As a unique geographic component, the mountain area is the birthplace of most rivers in arid regions. Almost all rivers originate in the mountainous regions [5]. Central Asia’s water resources are mainly recharged by melting ice and snow in the mountainous areas, making them more sensitive to climate change [6–8]. Human activities have led to global environmental changes [9]. Due to population growth and disproportional spatial distribution, developing industries, agricultural expansion, and increased urban construction, there is increased pressure on water and other natural resources because the anthropogenic demand for water is growing [10]. The river runoff is one of the primary forms and components of water resources and the most critical water source for production and living [11,12]. Affected by climate variability and anthropogenic activities, the river
runoff of many rivers worldwide has undergone significant changes, seriously threatening the state of regional water resources [13–18].

Under the influence of natural conditions such as climate and landforms, along with human activities, the evolution of river runoff obeys specific laws [19,20]. Anthropogenic activities such as river water intake, reservoir water storage, and cross-basin water transfer consume runoff and affect water resources’ spatiotemporal distribution of [21–24]. These activities also cause different degrees of spatiotemporal variation in the hydrological sequence of the Lake Issyk-Kul Basin [25], leading to inconsistencies and varying frequencies of the basin’s hydrological characteristics [26–28], which subsequently make the results of hydrology and water resource analysis subject to calculation deviations and distortions of hydrological laws [29]. Therefore, to further grasp the evolution of water resources in the basin, it is necessary to study the evolution of river runoff. In order to support water management, the first phase in determining the impact of the climate system on hydrology is to estimate the trends (direction and magnitude) of hydrological variables [30–32].

Trend analysis is used to evaluate the rate or direction of changes in a variable over a period of time [33,34]. Due to climate change, many researchers have paid great attention to trend identification and river runoff analysis over the past several decades [35,36]. Although there are classical trend identification methods, such as Sen’s slope Sen [37], regression analysis [38], Spearman’s rho [39], the Mann-Kendall (MK) trend test [40,41], and the others, these techniques search for monotonic trends as a whole within a given time. Atta-ur-Rahman and Dawood [42] used the Sen slope method to perform a spatial statistical analysis of temperature trends in northern Pakistan. Satanand, Saravanavan [43] have developed runoff forecasting models that utilize data mining tools and regression analysis techniques. Sharma, Patel [44] employed Spearman’s Rho to investigate the long-term changes in river flows from 1973–2013 in the Tapi basin of west-central India. Research has revealed that the river basin’s river flow is generally decreasing, except for the Tapi River itself. This downward trend is primarily due to human activities, such as changes in land-use patterns. Wang, Peng [45] investigated the characteristics of runoff variations and the respective impacts of climate variability and anthropogenic activities on runoff variations from 1961–2015 in the five sub-basins of the Tapi Basin by applying several approaches, including Hurst exponent analysis, the MK trend test, and the cumulative anomaly method. These well-known trend techniques limit the possibility of partial trend interpretation, information extraction, and knowledge. However, to draw meaningful inferences from the hydrological time series trends, the development of robust methods that require less stringent assumptions remains a considerable challenge.

The innovative trend analysis (ITA) method with more information functions can illuminate future potential climate change impacts, planning, management, design, and research [46]. For the benefit of researchers, the ITA method is rapidly spreading around the world. It is being utilized to assess hydrometeorological trends [47–51] since it can objectively identify “low”, “medium”, and “peak” data series values visually, rather than identifying trends as a whole [52–55]. Trend research has been the focus of tremendous efforts by researchers in the past decade. Demir and Kisi [56] employed an innovative trend analysis method to determine the possible annual total precipitation trends in the Turkish provinces in Tokat, Ordu, Sinop, Corum, Samsun, and Amasya. Huang, Liu [57] applied the ITA method to investigate the water quality parameters’ temporal evolutionary laws at six different aquifers in the New Three coal mine. In Turkey, monthly streamflows in the Yesilirmak, Western Black Sea, and Eastern Black Sea basins were examined using observed data from nine stations, as reported by Kisi, Santos [58]. They concluded that trends of “low” and “peak” streamflow data could be identified by the ITA method, which has some advantages over Şen’s method and MK test. In England, Alashan [59] investigated the temporal variations of Maximum temperatures from 1853–2017 using a modified Mann-Kendall (MMK) approach, Şen’s slope estimator (SSE), an innovative trend analysis. This study indicated that MMK_ITA (modified Mann-Kendall_Innovative trend analysis) has a more substantial influence on serial correlation than MK and MMK_SSE. MMK_ITA can be
used in trend calculations and is better than traditional methods, even when relying on time series.

As one of the crucial links in the water cycle, river runoff is affected by both climate (precipitation, temperature, evaporation) and human activities (projects involving the underlying surface, topography, and water conservancy, among others), and is highly complex, dynamic, and nonlinear [60,61]. Lake Issyk-Kul is an endorheic mountain lake [62]. Accordingly, climate change and human activities are affecting the lake’s surface runoff [63]. Many researchers have pointed out that surface runoff changes are the main driving factor in the water level fluctuations of Lake Issyk-Kul [64–66]. Exploring river runoff variability of in this endorheic lake basin is essential for developing sustainable water resources in the study area [67]. Still, there have been few studies on the characteristics of river runoff changes. To the best of our knowledge, there is no published research on the use of ITA of river runoff in the Lake Issyk-Kul Basin. Moreover, there are some deficiencies in the existing studies: (1) Most previous research focused on annual river runoff changes, while only a few investigations performed detailed analyses on monthly runoff changes; (2) The collected river runoff data durations have not been long, and the number of hydrological stations used in the studies has been limited; and (3) The “low”, “medium”, and “peak” trends of river runoff were not explored. Keeping the above reviews in mind, this study was conducted with the following specific objectives: (i) to analyze the temporal trend in seasonal and annual river runoff time series data using the MK test, Theil-Sen Median (TSM), and new ITA method with significance test; (ii) to compare the results of MK method with the new ITA.

2. Study Area and Data
2.1. Study Area
Lake Issyk-Kul (Figure 1) is an inland saltwater lake located in the basins of the Teskey Ala-Too Range and Kungey Ala-Too mountain ranges in the Western Tian Shan Mountains in northeastern Kyrgyzstan [68]. The center of the lake is 42.42° N, 77.33° E. Lake Issyk-Kul is 60.1 km from north to south and 178 km from east to west, with an area of 6236 km². Its water storage capacity can reach 1730 × 10⁸ m³, the perimeter of the lake shoreline is approximately 669 km, the average lake depth is 278.4 m, and the maximum lake depth is 702 m [64,69]. Among the world’s mountain lakes, Lake Issyk-Kul is second only to Lake Titicaca in South America in terms of area and ranks first of lake depth and lake water volume. The climate in the lake area is mild and dry. The average temperature in July is 15–25 °C, and the average temperature in January is −6 °C. The Lake Issyk-Kul drainage basin occupies a surface area of 22,080 km², and there are 118 large and small rivers with a surface runoff of about 36.7 × 10⁸ m³ [62]. The main river flows are Chong-Koi-Suu, Cholpon-Ata, Chong-Ak-Suu, Ak-Suu, Chong-Uryukty, Pzhergalan, Ak-sai, Tossor, Ton, Dzhuuiku, Tamga, Karakol, and Chong-Kyzyl-Su [63].

Lake Issyk-Kul is a closed inland lake with no outlet. Many rivers are used for irrigation before flowing into the lake, but rivers are not the only source of lake water. The annual surface water supply is 13 × 10⁸ m³, and the annual total of precipitation and groundwater is 330 mm. The annual evaporation of the lake is 820 mm.

2.2. Data Sources
This paper obtained the monthly and annual time series data (river runoff, covering the 1943–2012 period) selected from 13 hydrological stations surrounding Lake Issyk-Kul, as shown in Figure 1. This study compiled observational data provided by the National Science & Technology Infrastructure Center-Data Sharing Infrastructure of Earth System Science and the Department of Water Resources and Irrigation of the Ministry of Agriculture and Land Reclamation of the Kyrgyz Republic and the National Statistical Committee of the Kyrgyz Republic [63]. In this study, data from the four prominent seasons—spring (March-April-May), summer (June-July-August), autumn (September-October-November), and winter (December-January-February)—were analyzed. Also, the trends of seasonal
and annual data series. Table 1 lists the arithmetical evaluation results, including mean, maximum, minimum, standard deviation, kurtosis, and skewness of annual and seasonal river runoff time series (m$^3$/s). These results indicate that the river runoff in spring and autumn generally exhibits positive skewness (the river runoff has a highly skewed distribution, e.g., Ton annually, Chong-Urykt in winter), and the annual river runoff also displays positive kurtosis.

### Table 1. Descriptive statistics of annual and seasonal river runoff (m$^3$/s) at study hydrological stations between 1943 and 2012.

| Statistical Parameters | StatheroGal | Cholpon-Ala | Chongo-Ak-Su | Ak-Su | Chongo-Koi-Su | Ak-Sai | Karakol | Chong-Kory-Su | Dzhunuk | Tosor | Ton | Chong-Urykt | Tanga |
|------------------------|------------|------------|--------------|-------|--------------|-------|---------|--------------|---------|-------|-----|-------------|-------|
| **Annual**             |            |            |              |       |              |       |         |              |         |       |     |             |       |
| Min                    | 11.7       | 2.17       | 6.59         | 5.50  | 2.51         | 6.05  | 8.23    | 7.92         | 9.90    | 3.67  | 1.80| 2.79        | 6.39  |
| Max                    | 34.00      | 1.41       | 5.31         | 3.50  | 1.51         | 3.25  | 6.81    | 5.35         | 6.81    | 2.23  | 0.68| 1.51        | 4.22  |
| Mean                   | 21.43      | 0.29       | 0.59         | 0.83  | 0.41         | 0.92  | 0.86    | 1.02         | 1.08    | 0.50  | 0.29| 0.38        | 0.69  |
| Stdev                  | 4.22       | 0.62       | −0.28        | 0.37  | −0.08        | 0.92  | −0.54   | 0.49         | 0.26    | 0.04  | 0.40| 1.55        | 0.77  |
| Skew                   | 0.70       | 0.19       | −0.71        | −0.90 | −0.81        | 0.92  | −0.54   | 0.49         | 0.26    | 0.04  | 0.40| 1.55        | 0.77  |
| Kurt                   | 0.62       | 0.91       | 4.18         | 2.09  | 0.86         | 1.71  | 4.76    | 3.48         | 4.43    | 1.25  | 0.25| 0.75        | 2.93  |
| **Spring**             |            |            |              |       |              |       |         |              |         |       |     |             |       |
| Min                    | 10.8       | 0.41       | 1.82         | 1.38  | 0.83         | 0.43  | 0.87    | 2.25         | 1.65    | 1.47  | 0.77| 0.22        | 0.81  |
| Max                    | 32.00      | 1.64       | 4.77         | 4.23  | 1.87         | 2.76  | 5.81    | 5.35         | 4.77    | 2.36  | 1.94| 2.52        | 3.22  |
| Mean                   | 16.09      | 0.98       | 2.82         | 2.47  | 1.09         | 1.67  | 3.62    | 2.93         | 2.40    | 1.37  | 0.66| 1.44        | 1.62  |
| Stdev                  | 3.88       | 0.29       | 0.53         | 0.74  | 0.48         | 0.42  | 0.83    | 0.86         | 0.56    | 0.30  | 0.30| 0.42        | 0.54  |
| Skew                   | 1.56       | 0.22       | 0.60         | 0.31  | 0.58         | 0.86  | 0.59    | 0.89         | 1.41    | 0.87  | 2.02| 0.85        | 0.30  |
| Kurt                   | 3.73       | −0.35      | 0.99         | −0.87 | −1.21        | 0.17  | −0.05   | 0.69         | 3.90    | 1.25  | 0.26| 0.26        | 0.03  |
| **Summer**             |            |            |              |       |              |       |         |              |         |       |     |             |       |
| Min                    | 16.2       | 1.63       | 8.54         | 4.12  | 1.49         | 2.24  | 10.48   | 8.24         | 9.08    | 1.91  | 0.45| 0.96        | 0.69  |
| Max                    | 59.07      | 4.55       | 14.93        | 14.77 | 4.01         | 11.19 | 20.47   | 20.23        | 24.27   | 7.58  | 2.14| 5.18        | 17.64 |
| Mean                   | 32.51      | 2.79       | 11.55        | 7.74  | 2.50         | 5.61  | 16.50   | 13.11        | 16.55   | 4.66  | 1.12| 2.93        | 10.72 |
| Stdev                  | 9.50       | 0.65       | 1.56         | 2.10  | 0.53         | 1.82  | 2.25    | 2.38         | 3.10    | 1.43  | 0.43| 0.91        | 3.18  |
| Skew                   | 0.43       | 0.49       | 0.11         | 0.84  | 0.93         | 0.92  | −0.45   | 0.46         | 0.24    | −0.32| 0.51| 0.13        | −1.42 |
| Kurt                   | −0.27      | −0.17      | −0.68        | 0.59  | 0.60         | 1.30  | 0.03    | 0.34         | 0.00    | −0.56| −0.56| −0.16       | 3.59  |
Table 1. Cont.

| Statistical Parameters | Pzhergalan | Cholpon-Ata | Chong-Ak-Suu | Ak-Suu | Chong-Koi Suu | Karakol | Chong-Kyzyl-Suu | Dzhukuu | Tossor | Ton | Chong-Ury kty | Tamga |
|------------------------|------------|-------------|--------------|--------|---------------|---------|----------------|---------|-------|-----|---------------|-------|
| Autumn                 | 11.5       | 0.71        | 2.98         | 1.17   | 0.76          | 1.43    | 3.43           | 2.16    | 3.66  | 1.06| 0.76          | 1.43  |
| Max                    | 37.07      | 2.48        | 7.35         | 3.85   | 3.45          | 6.85    | 7.87           | 7.73    | 10.36 | 3.65| 3.45          | 6.85  |
| Mean                   | 22.26      | 1.28        | 4.68         | 2.48   | 1.49          | 3.42    | 5.10           | 3.81    | 6.17  | 1.74| 1.49          | 3.42  |
| Stdev                  | 4.93       | 0.36        | 1.00         | 0.64   | 0.67          | 1.25    | 0.91           | 1.06    | 1.29  | 0.52| 0.67          | 1.25  |
| Skew                   | 0.79       | 1.15        | 0.54         | 0.17   | 1.16          | 0.95    | 0.40           | 1.38    | 0.71  | 1.32| 0.95          | 0.40  |
| Kurt                   | 1.26       | 1.82        | -0.01        | -0.77  | 0.24          | 0.05    | 0.15           | 2.27    | 0.53  | 2.12| 1.82          | -0.01 |
| Winter                 | 7.4        | 0.31        | 1.43         | 0.75   | 0.33          | 1.25    | 1.19           | 0.88    | 1.41  | 0.71| 0.33          | 1.25  |
| Max                    | 23.97      | 1.10        | 3.43         | 2.16   | 2.11          | 4.06    | 2.76           | 3.07    | 3.95  | 1.95| 3.07          | 2.76  |
| Mean                   | 14.87      | 0.56        | 2.19         | 1.30   | 0.96          | 2.31    | 2.06           | 1.52    | 2.12  | 1.15| 1.52          | 2.06  |
| Stdev                  | 2.75       | 0.15        | 0.44         | 0.29   | 0.59          | 0.67    | 0.40           | 0.41    | 0.33  | 0.28| 0.59          | 0.67  |
| Skew                   | -0.32      | 0.96        | 0.79         | 0.30   | 0.97          | 1.12    | -0.38          | 0.92    | 0.34  | 0.28| 0.30          | 0.97  |
| Kurt                   | 1.84       | 1.80        | 0.49         | -0.25  | -0.82         | 0.42    | -0.72          | 0.09    | 0.50  | 1.19| 0.49          | -0.25 |

3. Methodology

This study used the MK trend test [40,41] and selected the Šen’s trend method [46,55] to evaluate the annual and seasonal, temporal trends and features of the river runoff. The data analysis is developed in the DPS (Data Processing System) and SPSS (Statistical Product and Service Solutions) software. These methods are described in the following sections.

3.1. Trend Test

The MK trend test is a rank-based nonparametric test employed to detect significant trends in meteorological and hydrological time series [40,41]. The MK trend test statistic $S$ is defined as:

$$S = \sum_{i=1}^{n-1} \sum_{k=i+1}^{n} \text{sgn}(x_k - x_i)$$

where, $x_k$ is the rank of the $k$th observation, $x_i$ is the rank of the $i$th observations, $n$ is the number of observations, and $\text{sgn} (\theta) = \text{sign} (\text{sgn})$, recorded as $\text{sgn} (x_k - x_i)$ and measured as:

$$\text{sgn}(\theta) = \begin{cases} 
1, & \theta > 0 \\
0, & \theta = 0 \\
-1, & \theta < 0 
\end{cases}$$

Assuming that the series with sample size $n \geq 8$ is the test statistic $S$, then if $H_0$ is valid, $S$ is approximately normally distributed, and $P$ is the number of tied groups, and $t_i$ denotes the number of data values in the $i$th tied group ($i = 1, 2, 3 \ldots n$). A tied group is a set of sample data having the same value. With variance $\text{var}[S]$ computed as:

$$\text{var}[S] = \frac{n(n-1)(2n+5) - \sum_{i=1}^{P} t_i(t_i-1)(2t_i+5)}{18}$$

where, $n$ is the number of observations in the sample, and $t_i$ is the number of tied groups.
After calculating the \( \text{var}[S] \) of the data series, the standardized statistic \( Z_c \) value is calculated using the following formula:

\[
Z_c = \begin{cases} 
\frac{S-1}{\sqrt{\text{var}(S)}}, & S > 0 \\
0, & S = 0 \\
\frac{S+1}{\sqrt{\text{var}(S)}}, & S < 0 
\end{cases}
\]  

(4)

The standardized statistical \( Z_c \) value follows the standard normal distribution \((Z)\), with a variance of one \((\sigma^2 = 1)\) and a mean value of zero \((\mu = 0)\). The null hypothesis \((H_0)\) assumes the river runoff observations have no trend, while the alternate hypothesis \((H_1)\) assumes the river runoff data series have no trend. Therefore, under the two-sided condition, \( H_0 \) and \( H_1 \) are tested at a significant level \((\alpha)\): If \(|Z| > 2.58\), the trend is significant at the confidence level of \(\alpha = 99\%\); if \(|Z| > 1.96\), the trend is significant at the confidence level of \(\alpha = 95\%\). If \(\pm Z_c > \pm Z_{\alpha/2}\), then \( H_1 \) is accepted, and \( H_0 \) is rejected. A negative (positive) value of \( Z_c \) indicates a downward (upward) trend in the observed data.

3.2. Innovative Trend Analysis Method

In this approach, a recorded observation time series is separated into two halves from the initial data to the final data, and the two sub-series are then sorted in ascending order. Based on the two-dimensional Cartesian coordinate system, the first half-series \((x_i)\) is located on the horizontal axis \((x\text{-axis})\), and the second half-series \((x_j)\) is located on the vertical axis \((y\text{-axis})\). The range of the two axes should be equal, as shown in Figure 2a. The 1:1 \((45^\circ)\) line divides the diagram into two similar triangles. If the data series accumulates on the 1:1 line, it is concluded that there is no trend (trendless). If the data series falls above the 1:1 line (upper triangular area), it can be assumed that the data series exhibits an upward trend. If the data series accumulates below the 1:1 line (lower triangular area), there is a downward trend in the time-series data [46,55]. Therefore, the variable domain of each semi-sequence can be classified into different patterns (“low”, “medium”, and “peak”). The scope of each group can be determined either qualitatively from experts’ opinions or quantitatively by dividing the range of change into subcategories. The closer the scatter points are to the 1:1 \((45^\circ)\) line, the weaker the trend slope [46,55].

![Figure 2. Illustration of the Şen’s method [46,55]](image)

(a) The increasing, decreasing, and no trend (trendless) in the ITA method; (b) The innovative trend slope.

The straight-line trend slope calculated by the following equation [46]:

\[
s = \frac{2(y_2 - y_1)}{n}
\]  

(5)
where \( s \) is the straight-line slope of the trend line represented by the ITA method; \( n \) is the number of data year; \( \overline{y}_1 \) and \( \overline{y}_2 \) are the arithmetic averages of the first half and the second half (\( y \), sequence) of the dependent variable. Figure 2b is a graphical representation for understanding the concept of trend slope. As shown in Figure 2b, the substitution of the numerical values as \( n = 70 \) and the arithmetic averages from Figure 2b as \( \overline{y}_2 = 36 \) and \( \overline{y}_1 = 42 \) in to Equation (5), yields \( s = 2 \times (36 - 42)/70 = -0.1714 \).

The standard deviation of sampling slope is defined as:

\[
\sigma_s^2 = \frac{8}{n^2} \frac{\sigma_y^2}{n} \left( 1 - \rho_{\overline{y}_1, \overline{y}_2} \right) \tag{6}
\]

The standard deviation of the sampling slope is \( \sigma_s \); the standard deviation of whole the time series is \( \sigma_y \), and \( \rho_{\overline{y}_1, \overline{y}_2} \) is the correlation coefficient between the ascendingly sorted two halves arithmetic averages calculates as:

\[
\rho_{\overline{y}_1, \overline{y}_2} = \frac{E(\overline{y}_1 \overline{y}_2) - E(\overline{y}_1)E(\overline{y}_2)}{\sigma_{\overline{y}_1} \sigma_{\overline{y}_2}} \tag{7}
\]

Finally, the lower and upper confidence limits (\( CL \)) of a standard normal PDF with zero mean and standard deviation are \( s_{cri} \), then the confidence limits (\( CL \)) of the trend slope at the \( \alpha \) level of significance are calculated by applying the following equation [70]:

\[
CL_{1-a} = 0 \pm s_{cri} \times \sigma_s \tag{8}
\]

where \( CL \) represents the lower and upper confidence limits at \( \alpha \) level of significance; \( s \) is the slope value, and \( \sigma_s \) is the standard deviation of sampling slope. A negative (positive) value of \( s \) indicates a decreasing (increasing) trend in the data points.

4. Results and Analysis

4.1. Trend Variations of Annual River Runoff in the Lake Issyk-Kul Basin

This study used the annual river runoff data from 1943–2012 at the 13 hydrological stations in the Issyk-Kul Basin to evaluate the annual and seasonal trends by employing the MK test and TSM, as presented in Table 2. These results show that the magnitudes of the trends at most of the selected river stations were positive (with the exceptions of the Pzhergalan and Tossor stations) and varied from 0.001–0.031 m\(^3\)/s. Notably, the stations in the northern portion of the study area (i.e., Chong-Koi-Suu, Cholpon-Ata, Ak-Suu, Chong-Urykty, and Chong-Ak-Suu) exhibited large slopes (the mean slope was 0.011 m\(^3\)/s). In contrast, the stations in the basins’ southeastern parts (i.e., Pzhergalan, Karakol, Chong-Kyzyl-Suu, and Dzhukuu) displayed higher slopes (the mean absolute slope was 0.028 m\(^3\)/s).

The southern parts of the basin (i.e., Ak-Sai, Ton, Tossor, and Tamga) mainly presented weak slopes (from \(-0.007\) m\(^3\)/s to 0.007 m\(^3\)/s, with a mean absolute slope of 0.0075 m\(^3\)/s). As for the MK trend test, the positive values were more dominant. At Ak-Suu, Chong-Koi-Suu, Chong-Urykty, Chong-Kyzyl-Suu, and Ton stations (38.5%), exhibiting statistically significant trends were shown at the \( \alpha = 99\% \) confidence level, confirming the results of the TSM. No significant negative trends were found at the Pzhergalan and Dzhukuu stations (\( Z_c = -1.93 \) and \( Z_c = -1.03 \), respectively), comprising 15.4% of the 13 stations. All other stations (i.e., Cholpon-Ata, Chong-Ak-Suu, Karakol, Ak-sai, Tossor, and Tamga, comprising 46.1%) displayed no significant positive trends.

After the MK test, the annual river runoff trends were determined by applying the ITA method, as listed in Table 3. It can be seen from this table that the slope (s) of yearly river runoff was dominated by positive values, i.e., Cholpon-Ata (0.74), Chong-Ak-Suu (0.83), Ak-Suu (3.73), Chong-Koi-Suu (1.98), Chong-Urykty (0.51), Chong-Kyzyl-Suu (3.67), Dzhukuu (3.55), Ak-Sai (1.93), Ton (2.83), and Tamga (1.52). Meanwhile, negative trends were found at Pzhergalan (\(-4.71\)), Karakol (\(-0.19\)), and Tossor (\(-0.18\)). These results imply that some decreasing trends ignored by the MK trend test can be detected by the ITA.
method, verifying its ability to detect the hidden patterns in data series. The ITA method’s graphical results for all 13 stations from 1943–2012 are presented in Figure 3. The increasing tendency of annual river runoff at most stations is indicated by the data points falling above the 1:1 (45°) line in the Cartesian coordinate system.

Table 3. Analysis of MK trend test and TSM trend analysis for the temporal runoff dynamics (1943–2012).

| Station Name       | Annual TSM | Spring MK | Summer TSM | Autumn MK | Winter TSM |
|--------------------|------------|-----------|------------|-----------|------------|
| Cholpon-Ata       | 0.001      | 1.48      | 0.004      | 2.42 °    | 0.021      |
| Chong-Ak-Suu      | 0.008      | 1.40      | −0.002     | −0.55     | 0.009      |
| Ak-Suu            | 0.016      | 7.08 °    | 0.018      | 6.57 °    | −0.005     |
| Chong-Koi-Suu     | 0.027      | 4.74 °    | 0.022      | 4.95 °    | 0.057      |
| Chong-Urykty      | −0.05      | −1.93     | −0.012     | −0.67     | −0.141     |
| Pzhergalan        | 0.001      | 0.31      | 0.005      | 0.91      | 0.002      |
| Karakol           | 0.031      | 5.95 °    | 0.026      | 5.94 °    | 0.061      |
| Chong-Kyzyl-Suu   | 0.029      | 1.03      | 0.007      | 2.38 °    | 0.086      |
| Ak-Sai            | 0.008      | 1.89      | 0.006      | 2.00 °    | 0.017      |
| Tossor            | −0.007     | 1.72      | −0.007     | −2.02 °   | −0.034     |
| Ton               | 0.008      | 7.01 °    | 0.008      | 5.90 °    | 0.005      |
| Tamga             | 0.007      | 1.68      | −0.007     | −2.02 °   | −0.008     |

** Trends at the 99% confidence level. ° Trends at the 95% confidence level; The slope by TSM is in m³/s.

4.2. Trend Variations of Spring River Runoff in the Lake Issyk-Kul Basin

The river runoff trends in the spring season were identified using the classical MK trend test and TSM, and the test statistics are summarized in Table 2. These findings demonstrate that the river runoff was dominated by positive trends, with seven stations, namely Ak-Suu ($Z_c = 4.95$), Chong-Koi-Suu ($Z_c = 6.57$), Chong-Kyzyl-Su ($Z_c = 5.94$), Ton ($Z_c = 5.90$), Cholpon-Ata ($Z_c = 2.42$), Ak-sai ($Z_c = 2.00$), and Dzuukuk ($Z_c = 2.38$) (53.8%) exhibiting significant increasing trends. In comparison, significant negative trends were found at Tossor ($Z_c = −2.02$) and Tamga ($Z_c = −2.02$) (15.4% of the 13 stations).

After applying the MK trend test, the ITA method was used for the spring season for river runoff. The test results are listed in Table 4. The ITA results indicated nine stations (Cholpon-Ata, Ak-Suu, Chong-Koi-Suu, Chong-Uryukty, Chong-Kyzyl-Su, Dzuukuk, Ak-sai, Tossor, and Ton) exhibited significant positive trends. In comparison, three stations (Chong-Ak-Suu, Pzhergalan, and Tamga) displayed significant negative trends. Still, employing the MK trend test, the values had both negative and positive directions. The
graphical ITA results for all the 13 river stations during the spring seasons from 1943–2012 are presented in Figure 4. As can be seen from this figure, for Tossor and Karakol, most of the data points for the spring season runoff fell on the no-trend line.

**Figure 3.** Analysis of the ITA for annual river runoff at 13 stations.

### 4.3. Trend Variations of Summer River Runoff in the Lake Issyk-Kul Basin

The MK trend test and TSM statistics obtained for the summer season river runoff are listed in Table 2. As can be seen from this table, three stations exhibited significant negative trends—Pzhergalan ($Z_c = -2.22$), and Chong-Kois-Suu ($Z_c = -2.01$) at the 95% level, and the Tossor ($Z_c = -3.81$) station at the 99% level. Meanwhile, significant positive values were dominant at four stations (30.8%)—Cholpon-Ata ($Z_c = 5.33$), Ak-Suu ($Z_c = 5.49$), Chong-Kyzyl-Suu ($Z_c = 5.00$), and Dzhuuku ($Z_c = 5.33$). Table 4 also summarizes the ITA test values of all 13 stations, which revealed that the positive values of the ITA slope ($s$) predominated, which only three stations—Pzhergalan, Karakol, and Dzhuuku—exhibiting negative values. The graphical ITA results for all 13 river stations during the summer seasons from 1943–2012 are illustrated in Figure 5.
Table 4. Analysis of ITA (slope s) of seasonal river runoff (1943–2012).

| Station       | Spring | Summer | Autumn | Winter |
|---------------|--------|--------|--------|--------|
| Cholpon-Ata   | 1.98 **| 1.94 **| 1.64 * | 0.13   |
| Chong-Ak-Suu  | 2.78 **| 8.45 **| 2.48 **| 1.20 * |
| Chong-Koi-Suu | 2.31 **| 2.27 **| 1.82 * | −0.59 *|
| Chong-Urykty  | 1.62 * | 0.48   | 1.56 * | −0.55 *|
| Pzhergalan    | −2.49 **| −2.15 **| −1.88 **| 2.14 **|
| Karakol       | −0.05 | −0.77 *| −0.42  | 0.15   |
| Chong-Kyzyl-Suu| 3.20 **| 1.24 * | 2.49 **| 7.65 **|
| Dzhuuku       | 3.50 **| −0.64 *| 1.59 * | 10.75 **|
| Ak-Sai        | 1.55 * | 0.93   | 3.19 **| 2.04 **|
| Tossor        | 1.69 * | 1.24 **| −2.24 **| 1.70 * |
| Ton           | 1.93 **| 0.82   | 1.55 * | 1.85 * |
| Tanga         | −0.68 *| 0.62   | −1.31 **| 0.60   |

** and * represent 99% and 95% significant levels, respectively.

Figure 4. Analysis of the ITA for spring river runoff at 13 stations.
Table 2 summarizes the MK test statistics obtained for the autumn river runoff. As illustrated by these results, the positive values were more dominant, and some were statistically significant at the 99% confidence level. Specifically, significant positive trends were found at Cholpon-Ata ($Z_c = 3.31$), Chong-Ak-Suu ($Z_c = 2.81$), Ak-Suu ($Z_c = 4.99$), Chong-Koi-Suu ($Z_c = 6.37$), Chong-Urykty ($Z_c = 3.26$), Chong-Kyzyl-Suu ($Z_c = 6.10$), Dzhuuku ($Z_c = 4.18$), and Ton ($Z_c = 5.48$) (61.5%), mainly scattered on the northern rivers of the Lake Issyk-Kul Basin. The summer river runoff trends were similar to the annual river runoff results due to the concentrated river runoff during summer in this region. The value at Karakol ($Z_c = 1.95$) was also very close to the positive significance limit ($Z_c = 1.96$). Besides, the ITA values are provided in Table 4 and graphically illustrated in Figure 6. As shown

Figure 5. Analysis of the ITA for summer river runoff at 13 stations.

4.4. Trend Variations of Autumn River Runoff in the Lake Issyk-Kul Basin

Table 2 summarizes the MK test statistics obtained for the autumn river runoff. As illustrated by these results, the positive values were more dominant, and some were statistically significant at the 99% confidence level. Specifically, significant positive trends were found at Cholpon-Ata ($Z_c = 3.31$), Chong-Ak-Suu ($Z_c = 2.81$), Ak-Suu, ($Z_c = 4.99$), Chong-Koi-Suu ($Z_c = 6.37$), Chong-Urykty ($Z_c = 3.26$), Chong-Kyzyl-Suu ($Z_c = 6.10$), Dzhuuku ($Z_c = 4.18$), and Ton ($Z_c = 5.48$) (61.5%), mainly scattered on the northern rivers of the Lake Issyk-Kul Basin. The summer river runoff trends were similar to the annual river runoff results due to the concentrated river runoff during summer in this region. The value at Karakol ($Z_c = 1.95$) was also very close to the positive significance limit ($Z_c = 1.96$). Besides, the ITA values are provided in Table 4 and graphically illustrated in Figure 6. As shown
in Table 4, the positive values of slope (s) were predominated, with negative trends only found at Pzhergalan, Karakol, Tossor, and Tamga.

![Graphs of river runoff trends for different stations](image)

**Figure 6.** Analysis of the ITA for autumn river runoff at 13.

### 4.5. Trend Variations of Winter River Runoff in the Lake Issyk-Kul Basin

The MK test statistics computed for winter river runoff are listed in Table 2. These results revealed that, significant positive trends were found at nine stations (69.23%), including Pzhergalan ($Z_c = 6.10$), Ak-Suu ($Z_c = 5.39$), Chong-Koi-Suu ($Z_c = 6.10$), Chong-Kyzyl-Suu ($Z_c = 6.28$), Tossor ($Z_c = 3.44$), Ton ($Z_c = 8.14$), Chong-Urykty ($Z_c = 3.34$) at the 99% level, and the Chong-Ak-Suu ($Z_c = 2.43$) and Ak-Sai ($Z_c = 6.28$) stations at the 95% level. In contrast, a significant negative trend was detected at Dzhuuku ($Z_c = -2.54$) at the 95% level. The value of Cholpon-Ata ($Z_c = 1.90$) was close to the positive significance limit ($Z_c = 1.96$). The other two stations (Karakol and Tamga) exhibited negative trends, but neither was significant.
These results indicate that the increasing annual river runoff trends at most stations were mainly due to river runoff in autumn and winter. It is worth noting that, with the exceptions of Dzhuuku, Karakol, and Tamga, winter river runoff at all stations displayed significant increasing trends. The ITA results for all 13 river stations are graphically illustrated in Figure 7 and summarized in Table 4. Negative trends were identified at Chong-Koi-Suu and Chong-Urykt, while all other stations showed positive tendencies.

Verify the reliability of the seasonal and annual river runoff trends in the study area, the ITA method (Tables 3 and 4), and two well-known methods of TSM and MK trend test (Table 2) were used. According to the temporal distribution of the MK test, significantly increasing (decreasing) trends were observed in 5 (0), 6 (2), 4 (3), 8 (0), and 8 (1) time series in annual, spring, summer, autumn, and winter river runoff data. At the same time, the ITA method detected significant trends in 7 (1), 9 (3), 6 (3), 9 (3), and 8 (2) time series in the study area. The significant positive and negative trends of the seasonal and annual river runoff

Figure 7. Analysis of the ITA for winter river runoff at 13.

4.6. Comparison of Trend Analysis Methods

Verify the reliability of the seasonal and annual river runoff trends in the study area, the ITA method (Tables 3 and 4), and two well-known methods of TSM and MK trend test (Table 2) were used. According to the temporal distribution of the MK test, significantly increasing (decreasing) trends were observed in 5 (0), 6 (2), 4 (3), 8 (0), and 8 (1) time series in annual, spring, summer, autumn, and winter river runoff data. At the same time, the ITA method detected significant trends in 7 (1), 9 (3), 6 (3), 9 (3), and 8 (2) time series in the study area. The significant positive and negative trends of the seasonal and annual river runoff
datasets were found to display similar decreases from north to south using either the ITA method (39.2% to 29.4%; Tables 3 and 4) or the MK trend test (46.15% to 25.64%; Table 2). Corresponding graphs of these results are displayed in Figures 3–7. For the summer and winter seasons, the data series (orange and yellow points) of most hydrological stations fell above the 1:1 (45°) line (Figures 5 and 7, respectively).

Besides, in Figure 8, the analysis of the ITA results (Figure 8f–j, and l) has compared with the MK trend test (Figure 8a–e, and k). Among the 65 annual (13 gauging stations annual data series) and seasonal data series (13 gauging stations spring, summer, autumn, and winter data series) studied, significant trends of 39 data series (60%) were found through the MK trend test. However, using the ITA method, 51 data series (78.46%) showed a significant trend. All significant trends identified by the MK trend test can also be detected by the ITA method. It can be seen from Figure 8a–e, and k and Figure 8f–j,l that the trend results of the MK trend test and ITA in the spring, summer, winter, and annual runoff series are similar. However, the ITA method can be used to effectively identify many significant trends that cannot be identified by the MK trend test (especially in spring and autumn). It should be said that the ITA method is beneficial for analyzing many hidden changing trends of river runoff data series in the study area.
Table 5 summarizes the results of the ITA method for “low”, “medium”, and “peak.” As presented in Table 5, the “low”, “medium”, and “peak” values of all 13 river stations were individually evaluated by the ITA method. As provided in Table 2, there is trendless (no trend) in annual and seasonal data series according to the MK test, but increasing and decreasing trends are seen with the ITA method (Table 5). The ITA method results revealed that increasing trends occurred in 89 of the 195 data series at the 13 hydrological stations (45.64%), while decreasing trends were found in 15 data series (7.69%). In comparison, 24 data series (26.97%) exhibited increasing patterns for “peak” river runoff, 25 data series (28.09%) displayed increasing patterns for “low” river runoff, and 40 data series (44.94%) showed increasing patterns for “medium” river runoff. According to the “low”, “medium”, and “peak” river runoff, five data series (33.33%), seven data series (46.67%), and three data series (20%) manifested downward trends, respectively. These results provided detailed information concerning the annual and seasonal river runoff patterns by evaluating the “low”, “medium”, and “peak” values.

| Name               | Annual ITA | Spring ITA | Summer ITA | Autumn ITA | Winter ITA |
|--------------------|------------|------------|------------|------------|------------|
|                    | L | M | P | L | M | P | L | M | P | L | M | P |
| Cholpon-Ata        | (+) | (+) | No | (+) | No | (+) | No | (+) | No | (+) | (+) | No | No | No |
| Chong-Ak-Suu       | No | (+) | No | (+) | No | (+) | No | (+) | No | (+) | (+) | No | (+) | (+) |
| Ak-Suu             | No | (+) | No | (+) | (+) | (+) | (+) | No | (+) | No | (+) | No | (+) | No |
| Chong-Koi-Suu      | (+) | (+) | No | (+) | (+) | (+) | No | No | No | No | No | No | No | No |
| Chong-Urykty       | (+) | (+) | No | (+) | (+) | (+) | (+) | No | No | No | No | No | No | No |
| Pzhergalan         | No | (−) | No | No | (−) | No | No | No | No | No | No | No | No | No |
| Karakol            | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| Chong-Kyzyl-Suu    | No | (+) | (+) | (+) | No | (+) | No | (+) | No | (+) | No | (+) | No | No | No |
| Dzhuuku            | No | (+) | (+) | No | No | (+) | (+) | (+) | (+) | (+) | (+) | No | (−) | No |
| Ak-Sai             | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| Tossor             | No | No | No | No | No | No | No | No | No | No | No | No | No | No |
| Ton                | (+) | (+) | No | (+) | (+) | No | (+) | (+) | (+) | (+) | No | No | No | No |
| Tamga              | No | (+) | (+) | (−) | (+) | (+) | No | No | No | No | No | No | No | No |

Notes: (±) and No represents the increasing (decreasing) and no trends; L-low, P-peak, M-medium.
5. Discussion

The change of river runoff characteristics has always been the focus and hotspot of hydrological and water resource scientific research. In this study, the ITA method presented by Sen [46] was applied at annual and seasonal time series recorded from 13 hydrological stations (Figure 1) surrounding the lake Issyk-Kul. The river is the product of climate. Suppose the weather station data describes the climatic characteristics of different locations. In that case, the river flow describes the interaction between precipitation and temperature in the river catchment area on a larger scale [71].

River discharge is determined by rainfall, which falls in the catchment area, and by evaporation, which depends on air temperature [72,73]. As reported by Alifujiang, Abuduwailli [74], the ITA method’s application at three sites in the Issyk-Kul Basin revealed that the precipitation exhibited significant increasing trends in summer and winter. Moreover, utilization of the MK trend test and ITA method demonstrated that the percentage of time series with significant positive trends of monthly precipitation was decreased from north to south by Alifujiang.et.al [74]. This study revealed that in the summer and winter seasons, the data series of most hydrological stations fell above the 1:1 (45°) line. The significant positive and negative trends of the seasonal and annual river runoff datasets were found to display similar decreases from north to south using either the ITA method or the MK trend test, both of which indicated that river runoff is affected by precipitation. Thus, external hydrological cycle processes mainly determine the precipitation in the Issyk-Kul Basin [66].

Rivers that are mainly supplied by glacial meltwater generally have a significant increase in runoff during the summer (from June to September), with more floods during the flood season and an increase in annual runoff [5]. According to the origin of Lake Issyk-Kul, there are four types of rivers, and most of the rivers are glacial-snow type [75]. Thus, some measures should take in the summer that is vulnerable to high floods in the future. For the Lake Issyk-Kul basin, the river stations on those 13 rivers are located above the water withdrawal to irrigation. Alymkulova, Abuduwailli [65] pointed out that agricultural water demand increased in summer (from May to July or August), while in October, it rises again. Our results showed most of the river stations have an increasing trend in summer. The results revealed that the time of the high volume of water intake is not under decreasing river runoff.

6. Conclusions

In this study, to verify the reliability of the seasonal and annual river runoff trends in the Lake Issyk-Kul basin, Central Asia, the ITA method and the well-known method of MK trend test were used. The ITA method can be used to effectively identify many significant trends that cannot be determined by the MK trend test. It should be said that the ITA method is beneficial for analyzing many hidden changing trends of river runoff data series in the study area. The conclusions as follows:

(1) The MK trend test results show that in 39-time series, there were significant positive and negative trends (among them, northern river in 18-time series, south-eastern rivers in 11-time series, southern rivers in 10-time series) on seasonal and annual river runoff.

(2) The ITA method results revealed that in 51-time series (in which the stations of the northern part of the lake basin in 20-time series, south-eastern parts in 16-time series, and southern regions in 15-time series), there were significant positive and negative trends on seasonal and annual river runoff. Specifically, the MK test found that the time series percentage decreased from 46.15% in the north to 25.64% in the south, while the ITA method revealed a similar rate of decrease, from 39.2% to 29.4%.

(3) According to the temporal distribution of the MK test, significantly increasing (decreasing) trends were observed in 5 (0), 6 (2), 4 (3), 8 (0), and 8 (1) time series in annual, spring, summer, autumn, and winter river runoff data. At the same time, the ITA method detected significant trends in 7 (1), 9 (3), 6(3), 9 (3), and 8 (2) time series in the
study area. The comparison results revealed that the ITA method could effectively identify the trends detected by the MK trend test.

(4) According to the ITA method, the “peak” values of 24 time series (26.97%) exhibited increasing patterns, 25 time series (28.09%) displayed increasing patterns for “low” values, and 40 time series (44.94%) showed increasing patterns for “medium” values. According to the “low”, “medium”, and “peak” values, five time series (33.33%), seven time series (46.67%), and three time series (20%) manifested decreasing trends, respectively. These results detailed the patterns of annual and seasonal river runoff data series by evaluating “low”, “medium”, and “peak” values.

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