Past and future changes of streamflow in Poyang Lake Basin, Southeastern China

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Abstract. To understand the causes of the past water cycle variations and the influence of climate variability on the streamflow, lake storage, and flood potential, we analyze the changes in streamflow and the underlying drivers in four typical watersheds (Gaoshan, Meigang, Saitang, and Xiashan) within the Poyang Lake Basin, based on the meteorological observations at 79 weather stations, and datasets of streamflow and river level at four hydrological stations for the period of 1961–2000. The contribution of different climate factors to the change in streamflow in each watershed is estimated quantitatively using the water balance equations. Results show that in each watershed, the annual streamflow exhibits an increasing trend from 1961–2000. The increases in streamflow by 4.80 m\textsuperscript{3}s\textsuperscript{−1}yr\textsuperscript{−1} and 1.29 m\textsuperscript{3}s\textsuperscript{−1}yr\textsuperscript{−1} at Meigang and Gaoshan, respectively, are statistically significant at the 5 % level. The increase in precipitation is the biggest contributor to the streamflow increment in Meigang (3.79 m\textsuperscript{3}s\textsuperscript{−1}yr\textsuperscript{−1}), Gaoshan (1.12 m\textsuperscript{3}s\textsuperscript{−1}yr\textsuperscript{−1}), and Xiashan (1.34 m\textsuperscript{3}s\textsuperscript{−1}yr\textsuperscript{−1}), while the decrease in evapotranspiration is the major factor controlling the streamflow increment in Saitang (0.19 m\textsuperscript{3}s\textsuperscript{−1}yr\textsuperscript{−1}). In addition, radiation and wind contribute more than actual vapor pressure and mean temperature to the changes in evapotranspiration and streamflow for the four watersheds.

For revealing the possible change of streamflow due to the future climate change, we also investigate the projected precipitation and evapotranspiration from of the Coupled Model Intercomparison Project phase 3 (CMIP3) under three greenhouse gases emission scenarios (SRESA1B, SRESA2 and SRESB1) for the period of 2061–2100. When the future changes in the soil water storage changes are assumed ignorable, the streamflow shows an uptrend with the projected increases in both precipitation and evapotranspiration (except for the SRESB1 scenario in Xiashan watershed) relative to the observed mean during 1961–2000. Furthermore, the largest increase in the streamflow is found at Meigang (+4.31 %) and Xiashan (+3.84 %) under the SRESA1B scenario, while the increases will occur at Saitang (+6.87 %) and Gaoshan (+5.15 %) under the SRESB1 scenario.

1 Introduction

Water resources are influenced by many aspects of environment (especially climate change, such as precipitation, evapotranspiration and temperature), economy and society (Kundzewicz et al., 2007; Zhang et al., 2007; Nash and Gleick, 1991; Liu and Fu, 1993; Milly et al., 2005; Gedney et al., 2006; Oki et al., 1995). Meanwhile, water resources also have a potential to severely affect environmental quality, economic development and social well-being (Kundzewicz et al., 2007; Zhang et al., 2007). As an important part of the water cycle, streamflow changes can significantly affect water resources, society safety and ecosystem health (Oki and
Kanae, 2006). Therefore, it can be used as an indicator of climate change owing to the intimate linkage between the water cycle and climate.

It was shown in the Intergovernmental Panel on Climate Change (IPCC, 2007) that the average global temperature increased by 0.74 ± 0.18°C in the past 100 yr, which impacted on the natural ecosystems and environment significantly. In addition, climate change may even be speeded up in the future, consequently leading to an increase in probabilities of floods and droughts. Therefore, changes in water resources and the underlying driving forces due to climate changes have become research focuses (Andréasson et al., 2004; Christensen and Lettenmaier, 2007; Frederick and Major, 1997; Gíl et al., 2010; Lins and Slack, 1999; Liu and Cui, 2009; Null et al., 2010; Piao et al., 2007; Thodsen, 2007; Vörösmarty et al., 2000; Xu et al., 2010; Zhang et al., 2001). Andréasson et al. (2004) discussed the impacts of climate change on streamflow under three anthropogenic CO₂ emission scenarios with a hydrology model (HBV) and concluded that the influences of climate change based on hydrology cycle varied geographically. Lins and Slack (1999) applied the non-parametric Mann-Kendall test to study temporal trends of streamflow at 395 gauging stations across the USA and suggested that streamflow increased in most regions, except for the northwest and the southeast Pacific. Zhang et al. (2001) pointed out that streamflow decreased significantly in most months, especially in August and September from 1947-1996 in southern Canada. Xu et al. (2010) analyzed the trends of major hydroclimatic variables from 1960–2007 in the Tarim River Basin of China and concluded that the impacts of increasing air temperature on streamflow showed different characteristics, depending on location and seasons. An increase in temperature tends to increase surface runoff, especially in mountainous regions due to the enhanced snowmelt and glacier melt in the spring, but to decrease the runoff in plain areas because of the increase in the actual evapotranspiration in the summer.

Over the last century, the average temperature in China has experienced a dramatic increase (Ding and Dai, 1994; Zhai et al., 2004; Zhang et al., 2005; Yang et al., 2010; Li et al., 2010), leading to an increase in areas with severe water stress (the Standing Committee of the National People’s Congress – NPC – of the People’s Republic of China, 1994). The Yangtze River Basin is one of the most advanced economic regions in China and has been affected by flooding seriously and frequently. As the temperature and precipitation increased, the frequency and intensity of floods in this basin showed dramatic increase over the past few years (Hu et al., 2007), resulting in serious economic losses. The projected increase of 2.7°C in temperature in the 21st century will induce precipitation and streamflow to increase by 10% (Gao et al., 2001) and 37% (IPCC, 2001), respectively, and the extreme events of rainfall may take place more frequently. Moreover, the occurrence probability of the most serious floods occurring in periods of 10-, 100-, 1000-yr, and even longer (e.g. floods in 1870, 1954 and 1998) may increase. Poyang Lake, as the largest freshwater lake in China, is the reservoir of floods in the middle and lower reaches of the Yangtze River. Its capacity of flood diversion has decreased continuously due to ecological and environmental degradation (e.g. the serious soil and water losses, and the reduction of lake areas and volume). The surface area of this lake shrunk by 25% and its capacity decreased by 22% from 1954–1998 (Jiang, 2007), consequently resulting in high vulnerability of the basin to floods. Both droughts and floods have occurred frequently and alternatively over the basin in recent decades. Furthermore, floods have increased in severity since 1990. In the summers of 1998, 1996, and 1995, the basin experienced three of its most severe floods (in descending order) in the last 50 yr (Wang and Dong, 2000; Jiang and Shi, 2003; Shankman et al., 2006).

Recently, the responses of hydrological cycle in the Yangtze River Basin to climate change have been causing more attention. The trend test and change-point analysis have been carried out using the annual maximum, annual minimum and annual mean discharge rates recorded at the Yichang gauging station during the period of 1882–2001 by Xiong and Guo (2004). They reported that at the 5% significance level, the annual maximum discharge rate did not have any statistically significant trend, but the annual minimum and the annual mean discharge rates significantly decreased by 8% and 6%, respectively. Applying the SWAT (Soil and Water Assessment Tools) model in Poyang Lake Basin of China, Guo et al. (2008) studied the annual and seasonal responses of streamflow to climate and the land-use changes and revealed that climate had a dominant effect on annual streamflow compared with the impacts of land-use changes, but the land-use changes could strongly influence seasonal variations of streamflow and alter the annual hydrograph of this basin. Chen et al. (2007) found that the mean annual, spring and winter runoff decreased at the 5% significance level in the Hanjiang Basin, caused by the integrated effects of changes in both precipitation and temperature. They also projected the increasing trends of runoff during the period of 2021–2050 under three climate scenarios of greenhouse gases emissions using a two-parameter water balance model. Zhao et al. (2009) declared that streamflow was more sensitive to precipitation variations than to potential evapotranspiration variations in Poyang Lake Basin.

In summary, most of the previous studies focused on qualitatively analyzing the effects of long-term variability of climate variables, particularly precipitation and temperature on water resources. However, the influences of other climatic variables, such as radiation, wind speed, and vapor pressure, on the past water cycle have not been studied thoroughly. Understanding the causes of water cycle variations clearly and systematically, it is necessary to examine the impacts of each climate variable on the streamflow variation. Knowing these responses, we can address the questions on how the on-going climate change may have influenced the streamflow, lake
storage, and flood potential in the past, and how the water cycle will vary with the future climate change. These problems are particularly important for water resources exploitation and utilization, agriculture production, and economy development. Therefore, the present study aims to: (1) quantify the contributions of various climate variables to the past (1961–2000) streamflow trends in Poyang Lake Basin on the basis of water balance equations, and (2) project the percentage changes of the streamflow in the future (2061–2100) relative to the past, using the precipitation and evapotranspiration data projected by different global coupled atmosphere-ocean general circulation models (AOGCMs) under three greenhouse gases emission scenarios.

2 Study region, data used and methods

2.1 Study region

Poyang Lake Basin is located in the middle reaches and the south bank of the Yangtze River, China, covering in total an area of \(1.6 \times 10^5 \text{ km}^2\), occupying nearly 96.85 % of the land mass of Jiangxi Province and accounting for 9 % of the Yangtze River Basin (Fig. 1). The size of the lake water body changes seasonally. It can exceed a maximum area of 4000 km\(^2\) in the summer and shrink to less than 3000 km\(^2\) in the fall and the winter. This lake receives water primarily from Ganjiang River, Xiushui River, Fuhe River, Raohe River, and Xinjiang River. The topography in Poyang Lake Basin is diverse, including mountains, hills, and alluvial plains. Mountains are mainly located in the western and eastern parts with a maximum elevation of 1800 m a.s.l. (above the sea level), while low alluvial plains are primarily in its central areas, mainly distributed in areas along Ganjiang River.

Four typical watersheds within Poyang Lake Basin were selected for studying the historical trends of streamflow. They are Meigang, Xiashan, Saitang, and Gaosha watersheds, located in the northeast, southeast, middle-west, and northwest parts of this basin (Fig. 1), respectively. The boundaries of watersheds were delineated using the hydrological analysis tools of ArcGIS 9.2 software package based on the 90 m STRM Version1 (http://dds.cr.usgs.gov/srtm/) Digital Elevation Model (DEM) data. The drainage areas for Meigang, Saitang, Gaosha and Xiashan, are \(1.53 \times 10^4 \text{ km}^2\), \(3.07 \times 10^3 \text{ km}^2\), \(5.22 \times 10^3 \text{ km}^2\) and \(1.59 \times 10^4 \text{ km}^2\), respectively.

The study area belongs to the subtropical monsoon climate zone, and it has a temperate and humid climate with abundant sunlight. Temperature and precipitation both exhibit distinct seasonality (Fig. 2). Among the four watersheds, monthly mean temperature (the left panel of Fig. 2a) increases from January to July and then decreases. The annual mean temperature during 1961–2000 was 16.6 °C in Saitang watershed, while it was above 17.9 °C in the other three watersheds. Monthly total precipitation (the left panel of Fig. 2b) increases quickly from January to June and then decreases sharply. During 1961–2000, the annual precipitation in Meigang and Saitang watersheds is about 1640 mm, while in Gaosha and Xiashan watersheds is about 1690 mm.
2.2 Data

2.2.1 Meteorological and hydrological data

The daily meteorological data during 1961–2000 from 79 weather stations (6 stations are outside Poyang Lake Basin) are used in this study (Fig. 1), including daily precipitation (mm), 20 cm caliber pan evaporation (mm), sunshine percentage (%), wind speed (m s$^{-1}$), maximum temperature ($^\circ$C), minimum temperature ($^\circ$C), mean temperature ($^\circ$C), actual water vapor pressure (kPa), and relative humidity of air (%). As there are only 2 weather stations with radiation observed in the study region, the methods proposed by Wang (2006) and Tong (1989) are used to calculate daily total incoming solar radiation (MJ m$^{-2}$ day$^{-1}$) and long-wave radiation (MJ m$^{-2}$ day$^{-1}$), respectively. The net radiation (MJ m$^{-2}$ day$^{-1}$) is calculated as the difference between the total incoming solar radiation and the long-wave radiation. The Spline Function Method in the ArcGIS 9.2 platform is employed to interpolate the annual mean/total values of climate variables at 79 stations into a dataset at the resolution of 1×1 km. The time series of regional means of climate variables for each watershed are calculated for the period from 1961–2000 to assess the impacts of climate on the streamflow.

Hydrological data used in this study include daily streamflow (m$^3$ s$^{-1}$) and river level (m) measured at Meigang, Saitang, Gaosha, and Xiashan hydrological stations (Fig. 1).

2.2.2 Data for the future climate scenarios

Simulations of AOGCMs for 20th century (20C3M) and 21st century climate were collected from the Couple Model Intercomparison Project phase 3 of (CMIP3). The 21st century simulations are used to project the changes of streamflow and climate during the period of 2061–2100. The 20th century climate (20C3M) was simulated with the contemporary climate scenario, whereas the future climate was projected under three different scenarios of greenhouse gases emission, including medium greenhouse gases emission scenario (SRESA1B), high greenhouse gases emission scenario (SRESA2), and low greenhouse gases emission scenario (SRESB1). For the future climate simulations, there is no detailed data for some variables (e.g. wind speed and vapor pressure). Therefore, only the monthly mean precipitation and evapotranspiration (converted from monthly latent heat fluxes) from the different AOGCMs under the three emission scenarios are used to estimate the percentage changes of future streamflow relative to the past. We chose only those model outputs with precipitation and latent heat flux and having data up to 2100. The selected AOGCMs and their resolutions are listed in Table 1. The details about these models and their outputs can be found at the website of http://www-pcmdi.llnl.gov/ipcc/model_documentation.ipcc_model_documentation.php.

2.2.3 Other data

The land use/land cover dataset in 1995 was downloaded from Environmental & Ecological Science Data Center for West China, National Natural Science Foundation of China (http://westdc.westgis.ac.cn). The SPOT VGT-NDVI datasets in 1999 and 2000 were from the VITO archive (http://www.vgt.vito.be). The spatial resolutions of these two datasets are 1×1 km. Additionally, a global long-term (1983–2006) 10-day evaportranspiration record with the resolution of 8×8 km (http://www.ntsg.umn.edu/project/et) was collected. Zhang et al. (2009, 2010) indicated that this dataset can capture observed spatial and temporal variations at the global and watershed scale. It can be used as the observational data for evaluating the simulations of actual evapotranspiration with Eq. (3) in the present study.

2.3 Methods

2.3.1 Temporal trends detection

The trends of hydrometeorological variables are fitted using the linear equation:

$$\hat{x}_i = f_0 + f_1 \cdot t (t = 1, 2, ..., n)$$

(A)
where $\hat{\beta}_1$, $f_0$, $f_1$, and $t$ represent the fitted value of the variable, intercept, temporal variability, and time, respectively; $n$ ($n=40$) is the sample size. A positive value of $f_1$ indicates an increasing trend, and vice versa. A larger magnitude of $f_1$ denotes a stronger increasing or decreasing trend. The Student’s t-test is used to examine the significance level of a trend. The p-value tells the probability of whether the linear trend value is statistically significantly different from zero.

### 2.3.2 Water balance for a watershed

The study region is located at subtropical climate zone with rare snowfall. Therefore, the water balance for a watershed is calculated as

$$ R = a \cdot P - E + \Delta W + q $$  \hspace{1cm} (2)

where $R$ is the streamflow (the sum of surface and underground runoff) measured at the outlet of a watershed. $a$ is the ratio of throughfall to total precipitation above canopy, and it depends on canopy density and rainfall intensity. Because a certain amount of rainfall is intercepted by vegetation canopy (Crockford and Richardson, 2000; Hölsher et al., 2004; Huang et al., 2005), the intercepted rainfall is not involved in the process of runoff yield. Only the throughfall affects the streamflow. $P$ is the precipitation amount. $\Delta W$ is the change of water storage in the watershed. $q$ is the water consumption from the watershed. In reality, $q$ is small in a closed watershed and it is assumed to be zero here for similarity. $E$ is the actual evapotranspiration, and can be calculated from the 20 cm caliber pan evaporation measurements:

$$ E = b \cdot E_{\text{pan}} $$  \hspace{1cm} (3)

where $b$ is the coefficient converting pan evaporation to actual evapotranspiration; $E_{\text{pan}}$ is the evaporation measured with the 20 cm caliber pan.

In the study region, as precipitation shows considerable interannual variability due to the monsoon climate. The interannual variations of water storage can not be ignored in the calculation of water balance for a watershed using Eq. (2). However, there is not any observation of water storage available at the watershed level. It is known that a tight linkage exists between long mean soil water storage and river level at the outlet in a watershed. In wet periods, both soil water storage and river level are expected to increase, and vice versa. Therefore, the change in river level can be used as a proxy for the change of soil water storage in a watershed to some extent. As an approximation, we use the intra-annual variability of river level ($\DeltaWL$) as a surrogate of $\Delta W$. To weaken the intense low-frequency turbulence in daily river level observations, $\DeltaWL$ is defined as the difference between the mean water level of the last 10-day in December and that of the first 10-day in January in the same year. $\Delta W$ is calculated as

$$ \Delta W = c \cdot \DeltaWL. $$  \hspace{1cm} (4)

When the study period ($n$) is long enough, $\Delta W$ satisfies the following equation:

$$ \frac{1}{n} \sum_{i=1}^{n} \Delta W_i = \frac{1}{n} \sum_{i=1}^{n} (c \cdot \DeltaWL_i) \approx 0. $$  \hspace{1cm} (5)

After the units conversion, Eq. (2) can be rewritten as

$$ R_{yr} = \left( a \cdot P_{yr}/1000 - b \cdot E_{\text{pan},yr}/1000 + c \cdot \DeltaWL_{yr} \right) \cdot A/(\text{yd} \cdot 24 \cdot 3600) $$  \hspace{1cm} (6)

where $R_{yr}$ (m$^3$s$^{-1}$), $P_{yr}$ (mm), and $E_{\text{pan},yr}$ (mm) represent the annual mean streamflow, the annual total precipitation and pan evaporation, respectively. $\DeltaWL_{yr}$ (m) denotes the intra-annual variability of river level at one year; $A$ (m$^2$) is

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**Table 1.** Global coupled atmosphere-ocean general circulation models (AOGCMs) and their 20th century climate (20C3M) simulations and 21st century climate projections used in this study.  

<table>
<thead>
<tr>
<th>Models</th>
<th>Resolution (20C3M)</th>
<th>20C3M scenarios (2061–2100)</th>
<th>SRESA1B</th>
<th>SRESA2</th>
<th>SRESB1</th>
</tr>
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<tr>
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<tr>
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</tr>
<tr>
<td>miroc3_2_medres</td>
<td>128 x 64</td>
<td>*</td>
<td>*</td>
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<td>144 x 90</td>
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<td>*</td>
<td>*</td>
</tr>
</tbody>
</table>

* denotes that the climate simulated or projected by a model was used in this study.
the watershed area; \( yd \) (day) is the number of days within one year; \( a, b \) and \( c \) are parameters to be optimized using the observed hydrological and climate data.

### 2.3.3 Contribution of different climate variables to the past streamflow changes

Evapotranspiration is a key component of water balance in a watershed. Temperature, radiation, wind speed, and actual vapor pressure are the major climate factors that influence actual evapotranspiration. Based on the Penman equation, the daily evaporation from a pan (Allen et al., 1998; Sun et al., 2010) can be expressed as

\[
E_{\text{pan}} = \frac{ET_R + ET_A}{K_p}, \quad ET_R = \frac{\Delta R_n - G}{\Delta + \gamma}, \quad ET_A = \frac{\gamma}{\Delta + \gamma} f (U_2) (e_s - e_a)
\]

(7)

where \( ET_R \) (mm day\(^{-1}\)) and \( ET_A \) (mm day\(^{-1}\)) are the daily reference evapotranspiration related to the radiation and aerodynamic terms, respectively; \( K_p \) (dimensionless) is the pan coefficient and is chosen as 0.67 (Xu et al., 2006). \( \Delta \) (kPa \(^\circ\)C\(^{-1}\)) is the slope of the saturation vapor pressure curve; \( \gamma \) (kPa) is the psychometric constant; \( \lambda \) (MJ mm\(^{-1}\) represents the latent heat of evapotranspiration. \( R_n \) (MJ m\(^{-2}\) day\(^{-1}\)) is the net radiation; \( G \) (MJ m\(^{-2}\) day\(^{-1}\)) is the soil heat flux density and assumed to be zero at the annual time step. \( f (U_2) \) (mm kPa\(^{-1}\) day\(^{-1}\)) is the function of wind speed (Sun et al., 2010); \( U_2 \) is the wind speed at 2 m height which is converted from the wind speed at 10 m height \((U_{10}); e_s \) (kPa) and \( e_a \) (kPa) is the saturation vapor pressure and the air vapor pressure, respectively. The various items in Eq. (7) are calculated following Allen et al. (1998).

The contribution of different factors to the streamflow changes is quantified by differentiating Eqs. (6) and (7), i.e.

\[
\frac{dR_n}{dt} = \left[ \frac{d(\frac{a}{1000} P_{yr})}{dt} - \frac{d(\frac{b}{1000} E_{\text{pan},yr})}{dt} + \frac{d(c \cdot \Delta WL_{yr})}{dt} \right] \frac{A}{yd \cdot 24 \cdot 3600}
\]

\[
= \frac{a}{1000} \frac{dP_{yr}}{dt} \frac{A}{yd \cdot 24 \cdot 3600} + \frac{b}{1000} \frac{\partial ET_A}{\partial R_n} \frac{dR_n}{dt} \frac{A}{yd \cdot 24 \cdot 3600}
\]

where \( P_{yr} \) is the annual precipitation, \( A \) is the area of the watershed, \( \Delta WL_{yr} \) is the annual change in soil moisture, and \( \frac{dP_{yr}}{dt} \) is the rate of change of annual precipitation.

### 2.3.4 Projected changes of precipitation and evapotranspiration, and their contributions to the future streamflow variations

In the present study, we didn’t utilize the complicated methods (e.g. statistical downscaling and dynamic downscaling) to process the datasets, instead we used a simple and efficient approach (named Delta method) to obtain the projected climate change which may occur in the study region. The Delta method was proposed by the United States Global Change Research Program (USGCRP) and has since been compared with other downscaling methods in the United States (Hay et al., 2000) and Yellow River Basin in China (Zhao and Xu, 2008). Recently, a number of scientists have utilized the same or similar method to evaluate the potential changes of streamflow or other environmental variables (Miller et al., 2003; Ju et al., 2007; Cramer et al., 2001).

The precipitation and evaporation simulated by AOGCMs (including 20C3M and three emission scenarios) were extracted for an area \((110–120^\circ E, 20–35^\circ N)\) and were interpolated into the study region using the Spline Function method in the ArcGIS 9.2 platform. The areal means of the projected precipitation and evapotranspiration during 2061–2100 were calculated for different watersheds. To constrain the effects of single model biases on assessing the future changes of streamflow caused by climate change, precipitation and evapotranspiration during 2061–2100 can be calculated as follows:

\[
\Delta \text{var}^k = \frac{1}{n} \sum_{i=1}^{n} \left( \frac{\text{var}^\text{obs}}{\text{var}^\text{20C3M}} - \frac{\text{var}^k}{\text{var}^\text{20C3M}} \right)
\]

(9)

where \( \text{var} \) denotes the mean of precipitation/evapotranspiration; \( \Delta \text{var}^k \) represents the change of the mean of var during 2061–2100 projected by AOGCMs relative to the mean of
Fig. 3. Comparison of observed and estimated annual mean streamflow during the period from 1991–2000. The solid lines are the 1:1 lines. RME and RMSE were estimated as $\text{RME} = \frac{1}{n} \sum_{i=1}^{n} \frac{R_{\text{OBS},i} - R_{\text{CAL},i}}{R_{\text{OBS},i}}$ and $\text{RMSE} = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (R_{\text{OBS},i} - R_{\text{CAL},i})^2}$, respectively. $R_{\text{OBS},i}$ and $R_{\text{CAL},i}$ are the observed and calculated streamflow values, $n = 10$.

observation during 1961–2000; $k$ denotes the $k$-th future climate scenarios; $n$ is the number of AOGCMs under the $k$-th scenario; $\text{var}_i^k$ is var during 2061–2100 projected by the $i$-th AOGCM under the $k$-th scenario; $\text{var}_{\text{obs}}^\text{20C3M}$ represents the mean of var observed during 1961–2000; $\text{var}_{\text{obs}}^\text{20C3M}$ is the mean of var projected by the $i$-th AOGCM under the contemporary emission.

Therefore, the future changes of streamflow relative to the observed mean during 1961–2000 caused by the changes in precipitation or evaporation ($\Delta R_i$) are quantified as

$$\Delta R_i = \frac{C}{D} \cdot \text{var}_i^k \cdot R_{\text{obs}} \cdot 100\%$$

where $\Delta R_i$ represents the percentage change caused by the single climate variables (precipitation or evapotranspiration); $R_{\text{obs}}$ is the mean of observed streamflow from 1961–2000; $C$ denotes the coefficient $a$ for the precipitation term or $b$ for the evapotranspiration term; $D$ is 1 and $K_p$ for precipitation and evapotranspiration, respectively.

### 3 Results and analyses

#### 3.1 Optimized parameters in the water balance equation

Based on the least squares method, parameters $a$, $b$ and $c$ in Eq. (6) are optimized using the observed datasets during the period 1961–1990 and listed in Table 2. Evidently these parameters differ in different watersheds (Table 2), and their differences are evaluated in the following discussion section. Validation using the independent climate and streamflow observations during the period from 1991–2000 confirms that the calibrated water balance equation (Eq. 6) is able to capture the interannual variations of streamflow in different watersheds (Fig. 3). The calculated annual mean streamflow is in good agreement with the observation in each watershed, with $r$ (the correlation coefficient between the calculated and observed streamflows) above 0.94 at the 5% significance level, relative mean error (RME) in the range from −3.8 to 0.98%, and root mean square error (RMSE) ranging from 7.24 to 39.21 m$^3$s$^{-1}$. The estimated streamflow is slightly larger than the observation in Meigang watershed, but slightly smaller than observations in the other three watersheds. The validation demonstrates that the model
developed in this study is applicable to calculate streamflow from climate, pan evaporation and water level data at the watershed scale.

Table 2 lists the values of parameters $a$, $b$ and $c$ in Eq. (6) optimized using the observations for two periods during 1961–2000, and indicates that these parameters differ little between the two periods, and thus can be applied to project the future changes of streamflow under different climate scenarios. The optimized parameter values are used to investigate the influences of the different climate variables on the streamflow.

### 3.2 Annual and seasonal variations of streamflow

Figure 4 shows the measured monthly and annual streamflow averaged over the period 1961–2000 for all the watersheds. Overall, the streamflow in each watershed increases from January, peaks in June and then decreases sharply from July, following the seasonal patterns of precipitation (Fig. 2b). The 40-yr means of annual streamflow are $578.35 \text{ m}^3 \text{s}^{-1}$, $84.08 \text{ m}^3 \text{s}^{-1}$, $158.71 \text{ m}^3 \text{s}^{-1}$ and $440.01 \text{ m}^3 \text{s}^{-1}$ for Meigang, Saitang, Gaosha, and Xiashan watersheds, respectively. The large differences in their magnitudes of the streamflow are mainly due to their differences in the scales.

The streamflow shows distinct interannual and decadal variations in all the watersheds (Fig. 5). The decadal means of streamflow are higher in 1990s than in other periods for all the watersheds (Table 3). Meigang and Saitang watersheds have the lowest streamflow in 1980s ($535.65 \text{ m}^3 \text{s}^{-1}$) and 1970s ($72.98 \text{ m}^3 \text{s}^{-1}$), respectively, while the lowest streamflow appears in 1960s for Gaosha and Xiashan watersheds. The streamflow generally shows overall increasing trends during the study period of 40 yr in all the four watersheds. It increases statistically significantly at the 5 % level in the Meigang ($4.78 \text{ m}^3 \text{s}^{-1} \text{ yr}^{-1}$) and Gaosha ($1.29 \text{ m}^3 \text{s}^{-1} \text{ yr}^{-1}$) watersheds. It also increases in Saitang and Xiashan watersheds, but with a small magnitude.

### 3.3 Temporal trends of precipitation, pan evaporation and intra-annual changes of river level

Figure 6 shows the temporal trends of annual precipitation, pan evaporation and intra-annual change of river level in the four watersheds during 1961–2000. Annual precipitation increases in each watershed. In Meigang and Gaosha watersheds, it has a statistically significant increasing trend of $8.05 \text{ mm yr}^{-1}$ and $8.65 \text{ mm yr}^{-1}$ at the 5 % level, respectively. Pan evaporation declines significantly ($p < 0.05$) in all the watersheds, with Saitang having the biggest decline ($-5.86 \text{ mm yr}^{-1}$), followed by Xiashan ($-5.31 \text{ mm yr}^{-1}$). Intra-annual change of river level decreases slowly but statistically insignificantly.

Seen from Eq. (7), it is known that the changes in the total net radiation, actual vapor pressure, mean temperature, and 2 m wind speed can impact the evapotranspiration obviously, further causing the streamflow to change. Hence, the rates of their changes during 1961–2000 are listed in Table 4. Annual total net radiation and wind speed decrease significantly, with a rate of change ranging from $-3.41 \text{ MJ m}^{-2} \text{ yr}^{-1}$ to $-10.21 \text{ MJ m}^{-2} \text{ yr}^{-1}$ and from $-8.00 \times 10^{-3} \text{ m s}^{-1} \text{ yr}^{-1}$ to $-1.48 \times 10^{-2} \text{ m s}^{-1} \text{ yr}^{-1}$ among the different watersheds, respectively. Actual vapor pressure shows small and insignificant increasing trends. Annual mean temperature marginally decreases in Saitang watershed ($-4.40 \times 10^{-3} \text{ C yr}^{-1}$), while it increases at very small rates in the other three watersheds.

### 3.4 Contributions of different climate factors to the changes of streamflow

The contributions of different factors to streamflow changes are quantified using Eq. (8) and shown in Table 5. Increases of precipitation, decreases of evapotranspiration and intra-annual variation of river level lead to increases in streamflow. Net radiation, actual vapor pressure, temperature and wind speed indirectly impact on streamflow through their roles in evapotranspiration. If evapotranspiration increases with net radiation and wind speed, streamflow will...
Evapotranspiration consequently have a decreasing trend. Evapotranspiration is positively correlated with temperature, and thus it increases with temperature. The streamflow consequently decreases (and vice versa) with temperature. In contrast, the relationship between evaporation and actual vapor pressure are opposite. Decreases of evapotranspiration caused by increases in actual vapor pressure lead to an increase in streamflow, and vice versa. In Meigang, Gaoshan, and Xiashan watersheds, precipitation has the biggest influence on streamflow, followed by evapotranspiration and then intra-annual variation of river level ($P^* > E^* > W^*$, Table 5). Hence, the increase in precipitation contributes most to the streamflow increment. These are consistent with the previous conclusions by Zhao et al. (2009) that precipitation is the major determinant of streamflow in Poyang Lake Basin. In Sai-tang watershed, precipitation increases marginally while actual evapotranspiration decreases significantly, caused by decreasing net radiation and wind speed. The decrease in actual
Table 3. Decadal variations of streamflow for the four watersheds.

<table>
<thead>
<tr>
<th>Watershed</th>
<th>1961–1970 (m³ s⁻¹)</th>
<th>1971–1980 (m³ s⁻¹)</th>
<th>1981–1990 (m³ s⁻¹)</th>
<th>1991–2000 (m³ s⁻¹)</th>
<th>2001–2000 (m³ s⁻¹)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Meigang</td>
<td>538.72</td>
<td>549.05</td>
<td>535.65</td>
<td>689.97</td>
<td>578.35</td>
</tr>
<tr>
<td>Saitang</td>
<td>83.32</td>
<td>72.98</td>
<td>84.54</td>
<td>95.48</td>
<td>84.08</td>
</tr>
<tr>
<td>Gaosha</td>
<td>14.06</td>
<td>151.38</td>
<td>148.49</td>
<td>188.93</td>
<td>158.71</td>
</tr>
<tr>
<td>Xiashan</td>
<td>404.58</td>
<td>443.46</td>
<td>439.97</td>
<td>472.04</td>
<td>440.01</td>
</tr>
</tbody>
</table>

Table 4. Changes of annual total net radiation, mean actual vapor pressure, mean temperature, and mean 2 m wind speed for the period of 1961–2000.

<table>
<thead>
<tr>
<th>Climate variability</th>
<th>Climate variability</th>
<th>Climate variability</th>
<th>Climate variability</th>
<th>Climate variability</th>
</tr>
</thead>
<tbody>
<tr>
<td>dRd/dt (MJ m⁻² yr⁻¹)</td>
<td>dTave/dt (°C yr⁻¹)</td>
<td>dEav/dt (kPa yr⁻¹)</td>
<td>dU/dt (m s⁻¹ yr⁻¹)</td>
<td></td>
</tr>
<tr>
<td>Meigang</td>
<td>–9.03**</td>
<td>–5.00 × 10⁻³</td>
<td>3.90 × 10⁻³</td>
<td>–9.60 × 10⁻³*</td>
</tr>
<tr>
<td>Saitang</td>
<td>–10.21**</td>
<td>–1.00 × 10⁻⁴</td>
<td>–4.40 × 10⁻⁵</td>
<td>–1.43 × 10⁻³*</td>
</tr>
<tr>
<td>Gaosha</td>
<td>–4.41*</td>
<td>–6.00 × 10⁻⁵</td>
<td>2.60 × 10⁻³</td>
<td>–8.00 × 10⁻³*</td>
</tr>
<tr>
<td>Xiashan</td>
<td>–9.26**</td>
<td>–8.00 × 10⁻⁴</td>
<td>1.70 × 10⁻³</td>
<td>–1.48 × 10⁻³*</td>
</tr>
</tbody>
</table>

Note: * and ** represent that the value is statistically significant at the 5% and 1% level, respectively.

Table 5. The variations of streamflow caused by the changes of precipitation, evapotranspiration and intra-annual river levels (units: m³ s⁻¹ yr⁻¹).

<table>
<thead>
<tr>
<th>Watershed</th>
<th>P*</th>
<th>E*</th>
<th>W*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Meigang</td>
<td>3.79</td>
<td>0.82</td>
<td>0.59</td>
</tr>
<tr>
<td>Saitang</td>
<td>0.13</td>
<td>0.19</td>
<td>0.09</td>
</tr>
<tr>
<td>Gaosha</td>
<td>1.12</td>
<td>0.21</td>
<td>0.05</td>
</tr>
<tr>
<td>Xiashan</td>
<td>1.34</td>
<td>0.86</td>
<td>0.66</td>
</tr>
</tbody>
</table>

3.5 Variations of streamflow under three future emission scenarios

Using the Eq. (9), Table 6 lists the changes of projected mean precipitation and evapotranspiration during 2061-2100 under three scenarios of greenhouse gases emission relative to the mean precipitation/evapotranspiration observed during 1961–2000. Multi-model ensemble means of the projected precipitation and evapotranspiration by the different AOGCMs are used to generate their integrated time series, and both of them exhibit considerable differences. For all of the four watersheds, precipitation and evapotranspiration are projected to increase under the three future climate scenarios, except for the evapotranspiration at Xiashan watershed under SRESA2. In addition, the increases in the precipitation as well as evapotranspiration under the scenario of SRESA2 will be the smallest, while evapotranspiration at Xiashan watershed will possibly decrease by 0.28% under this scenario.

Therefore, the future change of streamflow (ΔRᵢ) relative to the observed mean during 1961–2000 due to the changes in precipitation or evaporation is calculated with Eq. (10). Figure 7 depicts the calculated ΔRᵢ values for different climate change scenarios and watersheds. For all the watersheds, the projected precipitation changes will cause streamflow to increase under the three future climate scenarios. Under the SRESA1B scenario, the largest precipitation-induced increase of streamflow will be in the Meigang watershed (+7.16%) and Xiashan watershed (+6.04%). Under the SRESB1 scenario, the largest precipitation-induced increase of streamflow will be under the SRESB1 scenario in Saitang watershed (+9.13%) and Gaosha watershed (+6.93%). However, for the three future climate scenarios, evapotranspiration changes will cause streamflow to decrease in all
watersheds, except for Xiashan under the SRESA2 scenario, which shows an increase in streamflow by 0.16%. The largest decrease of streamflow caused by evapotranspiration appears under the scenario of SRESA1B for Meigang (−2.85%), Saitang (−2.42%), and Gaosha (−1.89%) watersheds. Evapotranspiration-induced decrease of streamflow will be the largest under the scenario of SRESB1 in the Xiashan watershed (−2.23%). The changes of streamflow will differ among watersheds for the same future climate scenario.

With the assumption that the future changes in soil water storage is ignorable, the simultaneous changes in both averaged precipitation and evapotranspiration will cause streamflow to increase in all of the watersheds (Green bars in Fig. 7). The changes in precipitation and evapotranspiration together will result in the largest increase in Meigang (+4.31%) and Xiashan (+3.84%) watersheds under the SRESA1B scenario, but in Saitang (+6.87%) and Gaosha (+5.15%) under the SRESB1 scenario.

4 Discussions

4.1 Causes of differences of parameters $a$, $b$, and $c$ among the four watersheds

Parameter $a$ represents the ratio of throughfall to total precipitation above canopy, while $1 - a$ is relative to the effectiveness of the interception capacity of vegetation (e.g. canopy interception and stem interception) and the intercepted water evaporation. Wen and Liu (1995) quantitatively analyzed
Table 6. Percentage changes of precipitation and evapotranspiration projected under three greenhouse gases emission scenarios.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Watershed</th>
<th>SRESA1</th>
<th>SRESA2</th>
<th>SRESB1</th>
</tr>
</thead>
<tbody>
<tr>
<td>Precipitation</td>
<td>Meigang</td>
<td>−2.07%</td>
<td>−2.86%</td>
<td>−0.50%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>4.78</td>
<td>2.30</td>
<td>4.42</td>
</tr>
<tr>
<td></td>
<td>Saitang</td>
<td>−5.69%</td>
<td>−4.60%</td>
<td>−0.31%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>3.83</td>
<td>2.44</td>
<td>6.00</td>
</tr>
<tr>
<td></td>
<td>Gaoshan</td>
<td>−5.36%</td>
<td>−3.1%</td>
<td>−1.56%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>4.95</td>
<td>2.67</td>
<td>5.04</td>
</tr>
<tr>
<td></td>
<td>Xiashan</td>
<td>−5.36%</td>
<td>−6.85%</td>
<td>−3.32%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>3.88</td>
<td>1.36</td>
<td>3.11</td>
</tr>
<tr>
<td>Evapotranspiration</td>
<td>Meigang</td>
<td>−0.03%</td>
<td>−2.44%</td>
<td>−1.39%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>5.8</td>
<td>1.81</td>
<td>5.27</td>
</tr>
<tr>
<td></td>
<td>Saitang</td>
<td>−0.37%</td>
<td>−7.56%</td>
<td>−1.11%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>4.67</td>
<td>1.30</td>
<td>4.36</td>
</tr>
<tr>
<td></td>
<td>Gaoshan</td>
<td>0.26%</td>
<td>−7.99%</td>
<td>−1.52%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>5.02</td>
<td>1.91</td>
<td>4.75</td>
</tr>
<tr>
<td></td>
<td>Xiashan</td>
<td>−0.76%</td>
<td>−7.58%</td>
<td>−0.73%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>3.91</td>
<td>−0.28</td>
<td>3.96</td>
</tr>
<tr>
<td>Streamflow</td>
<td>Meigang</td>
<td>−9.73%</td>
<td>−4.92%</td>
<td>−3.52%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>4.31</td>
<td>2.56</td>
<td>4.03</td>
</tr>
<tr>
<td></td>
<td>Saitang</td>
<td>−7.80%</td>
<td>−5.55%</td>
<td>−0.33%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>3.41</td>
<td>3.05</td>
<td>6.87</td>
</tr>
<tr>
<td></td>
<td>Gaoshan</td>
<td>−7.48%</td>
<td>−7.64%</td>
<td>−3.07%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>4.93</td>
<td>2.96</td>
<td>5.15</td>
</tr>
<tr>
<td></td>
<td>Xiashan</td>
<td>−8.95%</td>
<td>−8.89%</td>
<td>−5.78%</td>
</tr>
<tr>
<td></td>
<td>Average</td>
<td>3.84</td>
<td>2.28</td>
<td>2.61</td>
</tr>
</tbody>
</table>

the characteristics of rainfall interception of dormant forest ecosystems in China, and found that the interception coefficient (1 − a) differed considerably in various forest ecosystems and the mean values usually ranged from 11.4 to 36.5%. Crockford and Richardson (2000) and Fan et al. (2007) pointed out that a number of factors could influence the canopy interception coefficient, such as rainfall characters (quantity, intensity and duration), wind speed, environment, vegetation types and their canopy density. Generally, the interception coefficient is larger if the canopy is denser and the wind is weaker. There are no big differences in monthly rainfall among all the four watersheds (Fig. 2). Figure 8 shows the monthly and annual means of leaf area index (LAI) derived from the SPOT VGT-NDVI datasets during 1999–2000, and 10m wind speed from 1961–2000 for each watershed, respectively. For Gaoshan and Saitang watersheds, the seasonal variations of rainfall and LAI are asynchronous. Among all the four watersheds, Gaoshan watershed has the larger annual mean LAI and weakest wind speed, resulting in the smallest value of a (0.79) and highest interception coefficient (1 − a = 0.21) for the period 1961–2001. Meigang watershed has the highest wind speed and the second lowest LAI (only slightly higher than LAI in Xiashan watershed). It has the most precipitation in the months of March to June (Fig. 8) when its LAI is low owing to the high percentage (22.60%) of farmland in this watershed with two rotations of rice cultivation. The early rotation of rice is planted in the middle of April and harvest in late July. During the months from January to June, wind is much stronger here than in the other watersheds. The value of parameter a is highest among the four watersheds. The interception coefficient of this watershed during the period 1961–2000 is only 0.02.

The actual evapotranspiration of a watershed can be estimated based on Eq. (3) using 20 cm caliber pan evaporation observations. The estimated multi-year mean values of actual evapotranspiration differ little with the observations for all of the watersheds (Table 7), indicating that the parameter b is reasonable for simulating the actual evapotranspiration in general. Further analysis of the differences between the estimated actual evapotranspiration and the observations among the four watersheds shows some differences in Xiashan (−44.17 mm) and Meigang (46.22 mm). It is known that variation in the actual evapotranspiration can be influenced by water sources (e.g. precipitation and soil moisture), the radiation and aerodynamic driving factors (e.g. radiation, wind speed, temperature and atmosphere water vapor), and vegetation (e.g. vegetation types and physiological structure). Poyang Lake Basin belongs to a typical humid climate zone with annual precipitation more than 1600 mm. When there is enough water for evaporation and transpiration, the evapotranspiration processes are mainly determined by the radiation and aerodynamic driving factors, and vegetation. To
uncover the estimation versus observation difference, the averaged values of the radiation and aerodynamic terms are analyzed using Eq. (7) and the major driving factors of reference evapotranspiration are listed in Table 8. For Gaoshan, evapotranspiration is the smallest because of the least values of ET_R and ET_A, which are caused by the lowest R_n and U_2, respectively. Meigang and Xiashan have the highest evaporation values owing to the higher values of R_n, U_2 and (e_s - e_a). However, smaller ET_R and ET_A in Saitang result in lower evapotranspiration. In general, the differences in the mean evapotranspiration values for the four watersheds are mainly determined by R_n, U_2 and (e_s - e_a).

Because it is difficult to obtain reliable and long-term soil water storage data, we utilized Eq. (5) to evaluate the parameters c. The 40-yr means of intra-annual changes of river levels in the four watersheds ranged from −6.25 × 10^{-2} m to 1.19 × 10^{-2} m, which is consistent with the hypothesis that the long-term average of intra-annual variation of river level is small. This suggests that Eq. (6) can be used to calculate streamflow according to measurements of precipitation, pan evaporation, and river levels. Additionally, the effects of the soil water storage on streamflow will be investigated in our future work with hydrological models.

4.2 Potential impacts of other factors on streamflow

Long-term changes in streamflow depend on the balance of precipitation and evapotranspiration. The latter is mainly driven by climate factors and vegetation characteristics, such as radiation, wind, actual vapor pressure, temperature, and vegetation types and density. However, previous researchers mainly focused on the response of streamflow to precipitation, temperature and land cover changes. The influences of other climate factors (e.g., radiation, wind and actual vapor pressure) on evapotranspiration have received less attention. In this study, we find that the effect of temperature on the streamflow (seen from Table 5) is limited compared to other climate variables (e.g. radiation and wind) in the four watersheds. Because of the complementation between evapotranspiration and runoff from water balance equation, the contribution of temperature to evapotranspiration is also limited, which is in agreement with the previous findings by other researchers (Roderick and Farquhar, 2002, 2004, 2005; Roderick et al., 2007; Sun et al., 2010). The contribution of radiation and wind should be taken into account in investigating the driving factors of streamflow changes. On the other hand, the observed streamflow trends (dR/dt) cannot be exactly explained by the total contribution from precipitation, evapotranspiration and intra-annual river level. This is mainly due to exclusion of the effects of human activities (e.g. agricultural irrigation, water conservation facilities and land-use change; Guo et al., 2008), and acclimation of plant physiology (e.g. stomatal) and structures (e.g. LAI) to elevated atmospheric CO₂ concentrations (Gedney et al., 2006; Piao et al., 2007; Field et al., 1995; Cowling and Field, 2003). Land-use change and establishments of water conservation facilities can influence the interception of vegetation and the ability of soil infiltration, and thus play important roles on hydrological regimes, and mechanisms of runoff yield and concentration. Some studies suggested that land-use changes has impacted the water cycle and would continue to do so in the next century (Costa and Foley, 1997; Jackson et al., 2005; Foley et al., 2005). Since the late 1960s, Poyang Lake Basin has been used for high head hydropower production and navigation. By the end of 2005, there were 315 hydropower stations operating in Jiangxi Province, which were above 1000 KW (Zhao et al., 2009). On the other hand, the land-use change could also influence the annual and seasonal
flows, although the climate effect is the dominant factor in determining annual streamflow (Guo et al., 2008). After the Chinese economic reform, the hydropower plant construction, urbanization and population increment, etc. would definitely influence the catchments attributes and the water utilization. In order to estimate the contributions of climate changes to the streamflow more accurately, we will consider the effects of vegetation growth feedback, land-use change and human activities on the streamflow in the future.

5 Conclusions

Based on the historical streamflow data of the four gauge stations in Poyang Lake Basin, it is shown that the annual streamflow in the four watersheds exhibits different increasing trends during 1961–2000. The streamflows in the Meigang and Gaosha watersheds increase by 4.80 m³ s⁻¹ yr⁻¹ and 1.29 m³ s⁻¹ yr⁻¹, respectively, and these increasing trends are statistically significant at the 5% level.

Climate variability induces considerable changes in the terrestrial water cycle in the Poyang Lake Basin. Increased precipitation is the biggest contributor to the streamflow increment in Meigang, Gaosha, and Xiashan watershed, while decreased evapotranspiration is the main reason of streamflow increment in Saitang watershed. Changes due to the intra-annual changes of river levels are relatively small and can be ignored. Radiation, wind speed, actual vapor pressure and temperature can influence evapotranspiration processes, consequently leading streamflow to change indirectly. The sign of the contribution (positive or negative) to the streamflow depends on the relationships among climatic variables, evapotranspiration and streamflow. In this study, radiation and wind reduction cause the streamflow to increase for each watershed, and thus the decreasing actual vapor pressure results in a decrease in streamflow. Streamflow decreases with the increase in mean temperature in Meigang, Gaosha and Xiashan watersheds, but increases slightly in Saitang watershed due to the decreases in mean temperature. Comparing the contribution of the different climate variables to evapotranspiration and streamflow trends in the four watersheds, radiation and wind have the larger contribution than the actual vapor pressure and mean temperature.

The future climates projected by different AOGCMs under SRESA1B, SRESA2 and SRESB1 scenarios are used to assess the future changes of streamflow in the study region. Ignoring the changes of soil water storage, with an increase in precipitation and evaporation (except for the SRESB1 scenario in the Xiashan watershed), the streamflow shows an upward trend. Furthermore, the most significant increase of the streamflow is found at Meigang (+4.31 %) and Xiashan (+3.84 %) under the SRESA1B scenario. However, the increases in the streamflow at Saitang (+6.87 %) and Gaosha (+5.15 %) are projected under the SRESB1 scenario.

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