Vegetation dynamics and climate seasonality jointly control the interannual catchment water balance in the Loess Plateau under the Budyko framework

Tingting Ning[1,2], Zhi Li[3], and Wenzhao Liu[1,2]

[1] *State Key Laboratory of Soil Erosion and Dryland Farming on the Loess Plateau, Institute of Soil and Water Conservation, CAS & MWR, Yangling, Shaanxi 712100, China*

[2] *University of the Chinese Academy of Sciences, Beijing 100049, China*

[3] *College of Natural Resources and Environment, Northwest A&F University, Yangling, Shaanxi 712100, China*

Correspondence to: Wenzhao Liu, wenzhaoliu@hotmail.com

Abstract. Within the Budyko framework, the controlling parameter (ω in the Fu equation) is widely considered to represent landscape conditions in terms of vegetation coverage (M); however, some qualitative studies have concluded that climate seasonality (S) should be incorporated in ω. Here, we discuss the relationship between ω, M, and S, and further develop an empirical equation so that the contributions from M to actual evapotranspiration (ET) can be determined more accurately. Taking 13 catchments in the Loess Plateau as examples, ω was found to be well correlated with M and S. The developed empirical formula for ω calculations at the annual scale performed well for estimating ET by the cross-validation approach. By combining the Budyko framework with the semi-empirical formula, the contributions of changes in ω to ET variations were further decomposed as those of M and S. Results showed that the contributions of S to ET changes ranged from 0.1% to 65.5 % (absolute values); therefore, the impacts of climate seasonality on ET cannot be ignored. Otherwise, the contribution of M to ET changes will be estimated with a large error. The developed empirical formula between ω, M, and S provides an effective method to separate the contributions of M and S to ET changes.

KEYWORDS: Budyko framework; Controlling parameter; Vegetation dynamics; Climate seasonality; Loess Plateau

1. Introduction

The water cycle has been influenced greatly by human activities and climate change since the 1960s, and considerable variability in hydrological processes has been observed in many basins around the world; this has led to a series of problems concerning essential water resources (Stocker et al., 2014). Analyses of the mechanisms of the interactions among the water balance, climate, and catchment surface conditions are important for understanding these complex processes at different spatio-temporal scales (Zhang et al., 2008), and such work has practical significance in regard to the improvement of water resources and land management (Rodriguez-Iturbe, 2000; Xu et al., 2014).

Budyko (1948, 1974) postulated that precipitation (P, represents the water supply from the atmosphere) and potential evapotranspiration (ET_0, represents the demand by the atmosphere) are the two dominant variables that control the long-term average water balance. The Budyko framework is considered one of the most abiding frameworks linking climatic conditions to the runoff (R) and actual evapotranspiration (ET) of a catchment (Donohue et al., 2007), and it has been used successfully to investigate interactions between hydrological processes, climate variability, and landscape characteristics (e.g. (Milly, 1994; Woods, 2003; Yokoo et al., 2008; Yang et al., 2009)). A series of empirical formulas have been developed for the Budyko curve based on theoretical research and case studies of regional water balance over the past 50 years. Among them, the Fu (Fu, 1981; Zhang et al., 2004) and Choudhury–Yang equations (Choudhury, 1999; Yang et al., 2008) have been used widely; furthermore, the controlling parameters ω (in the Fu equation) and n (in the Choudhury–Yang equation) are related linearly (Yang et al., 2008).

Deviations from the Budyko curve have been detected in previous studies, which indicates that in addition to climate conditions, other variables can also influence the variability of regional water balances (Yang et al., 2007; Wang and Alimohammadi, 2012). Two kinds of factors have been identified to be responsible for the deviations. The first type of factors are related to land surface conditions, and these include vegetation dynamics (Donohue et al., 2007; Yang et al., 2009; Donohue et al., 2010; Li et al., 2013; Zhang et al., 2016), soil properties, and topography (Yang et al., 2007; Peel et al., 2010). The second type of factors include seasonal climate variability (in addition to P and ET_0), such as storm depth (Donohue et al., 2012; Shao et al., 2012; Li et al., 2014), frequency of daily rainfall (Milly, 1994), and differences in the timing of P and ET_0 (Budyko, 1961; Potter et al., 2005). All of these factors can be encoded into the controlling parameter of the Budyko equations (e.g. ω in the Fu equation and n in the Choudhury–Yang equation). So far, a great deal of attention has been paid to the relationships between land surface conditions and the controlling parameter. Based on satellite products of vegetation such as the Normalized Difference Vegetation Index (NDVI), vegetation has been found to correlate well with the controlling parameter, and some empirical relationships have been successfully developed (Yang et al., 2009; Li et al., 2013). In particular, the controlling parameter can be better represented by vegetation when higher spatiotemporal resolution products are used. Therefore, the impacts of dynamic changes in vegetation on hydrology can be effectively quantified.

Many current studies attribute any effects of the controlling parameter to landscape characteristics (Roderick and Farquhar, 2011; Zhou et al., 2015; Zhang et al., 2016). However, both empirical evidence and modelling tests have demonstrated the important function of climate seasonality on catchment water yield, and thereby, evidence exists that climate seasonality also strongly affects the controlling parameter in the Budyko equations (Berghuijs and Woods, 2016). Some indices and models have thus been developed to address this issue, and several potential solutions have been discussed (Milly, 1993, 1994; Potter et al., 2005; Yokoo et al., 2008; Feng et al., 2012; Li, 2014). Yang et al. (2012) introduced the climate seasonality index into the Budyko framework and proposed an empirical equation to include its effect in the estimation of the long-term controlling parameters; however, by focusing on the mean annual scale, the effects of vegetation dynamics were not considered. Therefore, how the vegetation dynamics and climate seasonality jointly control the interannual variability in the controlling parameters needs further interpretation.

Therefore, the primary motivation behind this study was to detect the potential linkages between the controlling parameter and surface condition change, as well as climate seasonality at an annual scale. The specific objectives were to derive an appropriate analytic formula between parameter ω in the Fu equation and the above two factors for typical catchments in the Loess Plateau, and then, quantify the impacts of vegetation change and climate seasonality variability on the catchment water balance.

2. Methods

2.1. Annual water balance definition

The Budyko framework assumes that the long-term average water balance is in a steady state (Wang and Alimohammadi, 2012), and the water storage change in a catchment can be negligible. The interannual variability of the water balance in individual basins can also be studied by overlooking the interannual variation of the catchment water storage (Sankarasubramanian and Vogel, 2002; Yang et al., 2007; Potter and Zhang, 2009). However, water storage change can be great when analysing the interannual variability of the water balance (Wang, 2012). To minimize the potential errors introduced by neglecting water storage variation, the hydrological year (Sivapalan et al., 2011; Carmona et al., 2014) and moving windows (Jiang et al., 2015) were introduced to the time series of annual hydrological variables. Similar to Sivapalan et al. (2011) and Carmona et al. (2014), the hydrological year rather than the calendar year is introduced to calculate the annual *ET*, and this is called the ‘‘measured’’ *ET* in the subsequent discussion. Specifically, as the study area has a semiarid climate with most rainfall occurring in summer and autumn (July–September), a hydrological year is defined as July to June of the following year. In this way, the water input occurs mainly at the beginning of the year and the water is consumed within that year.

2.2. Identification of factors determining parameter ω in Fu's equation

The Fu equation is used in this study with the following expressions:

$$\frac{ET}{P} = 1 + \frac{ET_0}{P} - \left[1 + \left(\frac{ET_0}{P}\right)^{\omega}\right]^{1/\omega} \text{ or}$$

$$\frac{ET}{ET_0} = 1 + \frac{P}{ET_0} - \left[1 + \left(\frac{P}{ET_0}\right)^{\omega}\right]^{1/\omega} \quad (1)$$

where ω is the controlling parameter of the Budyko curve. ET_0 is calculated by using the equation of Priestley and Taylor (1972).

The important issue regarding the parameterization of ω in Fu's equation is to choose factors with physical meanings. According to the results from related studies, land surface conditions can be mainly represented by vegetation, which was also true in this study. With an arid to semiarid climate, water availability is the key factor that controls vegetation dynamics. Although soil properties and topography also influence vegetation growth, their impacts can be ignored on an annual scale because they would be expected to be almost constant over a year. Therefore, vegetation dynamics (i.e. vegetation coverage) were chosen to represent the variations in surface conditions. The vegetation coverage (M) was estimated by the following equation (Yang et al. (2009)):

$$M = \frac{NDVI - NDVI_{\min}}{NDVI_{max} - NDVI_{min}} \quad (2)$$

where $NDVI_{max}$ and $NDVI_{min}$ are the NDVI values of dense forest (0.80) and bare soil (0.05), respectively.

Two limiting conditions were used to illustrate the effects of seasonal variations in coupled water and energy on the regional water balance. If P and ET_0 are in phase, the intra-annual distribution of precipitation is very symmetrical, and thus, $R \to 0$ in non-humid regions and $ET \to P$. However, if P and ET_0 are out phase, the total precipitation of one year is concentrated at a certain moment, and thus, $R \to P$ and $ET \to 0$. Therefore, the impacts of seasonal variations in coupled water and energy on the regional water balance cannot be neglected, and they can only be reflected by the controlling parameter.

Solar radiation was considered as the dominant factor that controls the climate seasonality and thus the seasonality of P and ET_0 can be can be expressed by sine functions (Milly, 1994;Woods, 2003):

$$P(t) = \bar{P}(1 + \delta_P sin\omega t) \quad (3a)$$

$$ET_0(t) = \overline{ET_0}(1 + \delta_{ET_0} sin\omega t) \quad (3b)$$

where δ_P and δ_{ET_0} are the seasonal amplitude of precipitation and potential evapotranspiration, respectively. The values of δ_P and δ_{ET_0} might both range from -1 to 1 because P and ET_0 always have positive value on physical grounds. Larger absolute values of δ_P and δ_{ET_0} mean larger variability of climate seasonality. φ is the duration of the seasonal cycle, $2\pi\varphi$ equal to 1 year. Woods (2003) summarized the modelled climate of Eqs.(3a) and (3b) in dimensionless form and defined the climate seasonality index (S) and here it was used to reflect the non-uniformity in the annual distribution of water and heat in our study:

$$S = \left|\delta_P - \delta_{ET_0}\emptyset\right| \quad (4)$$

where $\emptyset$ is the dryness index, $\emptyset = \overline{ET_0}/\bar{P}$. If S=0, there is no seasonal fluctuation of the difference between P and ET_0. Larger values of S indicate that the larger changes in the balance between P and ET_0 during the seasonal cycle.

2.3. Evaluating the contributions of climate change and surface condition alterations to *ET* changes

Based on the climate elasticity method, which was introduced by (Schaake and Waggoner, 1990) and improved by (Sankarasubramanian et al., 2001), the contribution of change for each climate factor to runoff was defined as the product of the sensitivity coefficient and the variation of the climate factor (Roderick and Farquhar, 2011):

$$dR = \frac{\partial R}{\partial P}dP + \frac{\partial R}{\partial ET_0} + \frac{\partial R}{\partial \omega}d\omega \quad (5)$$

However, due to ignoring the higher orders of the Taylor expansion in equation (5), this method will result in high errors (Yang et al., 2014b). Recently, Zhou et al. (2016) proposed a new method to partition climate and catchment effect on the mean annual runoff based on the Budyko complementary relationship, called "the complementary method". The algebraic identities in their work can ensure that the change in runoff can be decomposed into two components precisely without any residuals. Here, we extend "the complementary method" to conduct attribution analysis of ET changes for each basin by further incorporating the effects of vegetation coverage and climate seasonality:

$$\Delta ET = \alpha\left[\left(\frac{\partial ET}{\partial P}\right)_1 \Delta P + \left(\frac{\partial ET}{\partial ET_0}\right)_1 \Delta ET_0 + P_2\Delta\left(\frac{\partial ET}{\partial P}\right) + ET_{0,2}\Delta\left(\frac{\partial ET}{\partial ET_0}\right)\right]$$

$$+(1-\alpha)\left[\left(\frac{\partial ET}{\partial P}\right)_2 \Delta P + \left(\frac{\partial ET}{\partial ET_0}\right)_2 \Delta ET_0 + P_1\Delta\left(\frac{\partial ET}{\partial P}\right) + ET_{0,1}\Delta\left(\frac{\partial ET}{\partial ET_0}\right)\right] \quad (6)$$

where α is a weighting factor that varies from 0 to 1, which can determine the upper and lower bounds of the climate and the controlling parameter effect. In this study, we defined α=0.5 according to the recommendation of Zhou et al. (2016).The difference operator (Δ) refers to the difference of a variable from period 1 (1981 to the changing point detected by Pettitt's test (Pettitt, 1979)) to period 2 (period-1 end to 2012) ,e.g., $\Delta ET_0 = ET_{0,2} - ET_{0,1}$. Then the contributions of P, ET_0, and ω changes to the ET changes can be expressed as follows:

$$C\_(P) = \alpha\left[\left(\frac{\partial ET}{\partial P}\right)_1 \Delta P\right] + (1-\alpha)\left[\left(\frac{\partial ET}{\partial P}\right)_2 \Delta P\right] \quad (7a)$$

$$C\_(ET_0) = \alpha\left[\left(\frac{\partial ET}{\partial ET_0}\right)_1 \Delta ET_0\right] + (1-\alpha)\left[\left(\frac{\partial ET}{\partial ET_0}\right)_2 \Delta ET_0\right] \quad (7b)$$

$$C\_(\omega) = \alpha\left[P_2\Delta\left(\frac{\partial ET}{\partial P}\right) + ET_{0,2}\Delta\left(\frac{\partial ET}{\partial ET_0}\right)\right] + (1-\alpha)\left[P_1\Delta\left(\frac{\partial ET}{\partial P}\right) + ET_{0,1}\Delta\left(\frac{\partial ET}{\partial ET_0}\right)\right] \quad (7c)$$

After obtaining the contribution of parameter ω to the ET change, the contributions of vegetation coverage (M) and climate seasonality (S) to ET change can be further decomposed as follows.

First, the contributions of M and S to parameter ω are calculated by using the sensitivity method similar to Eq. (5) based the relationship between ω and M as well as S we built:

$$\Delta\omega = \frac{\partial\omega}{\partial M}\Delta M + \frac{\partial\omega}{\partial S}\Delta S \quad (8)$$

Furthermore, the individual relative contributions (RC) of M and S to ω can be calculated. Then, the contributions of M (C_(M)) and S (C_(S)) to ET changes can be obtained as follows:

$$C\_(M) = C\_(\omega) \times RC\_(M) \quad (9a)$$

$$C\_(S) = C\_(\omega) \times RC\_(S) \quad (9b)$$

3. Study area and data

The Loess Plateau, which is located in the middle reaches of the Yellow River in China, experiences a sub-humid and semiarid continental monsoon climate (Ning et al., 2016). Frequent heavy summer storms, sparse vegetation coverage, easily erodible wind-deposited loess soil, and a long agricultural history have all contributed to severe drought and soil erosion problems in this region (Li et al., 2012). To recover and preserve the ecosystem, the Chinese government has launched numerous soil and conservation measures since the 1950s, and these include biologic measures (“Grain to Green” Project) and engineering measures (building terraces and sediment trapping dams) (Mu et al., 2007). As a result, the hydrological processes of this area have undergone significant changes (Huang and Zhang, 2004; Zhang et al., 2008). Thirteen catchments on the Loess Plateau were selected as our study area (Figure 1).

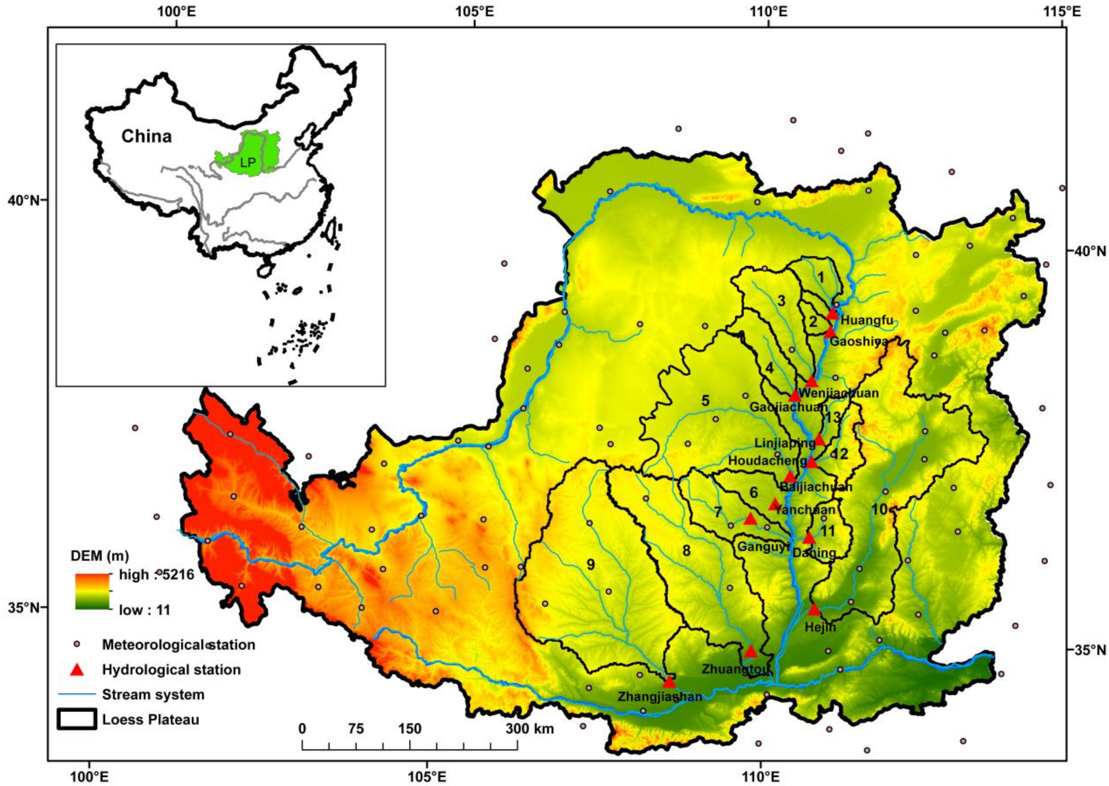

Figure 1. Locations of the study area and hydrometeorological stations.

Monthly runoff data for the 13 catchments were supplied by the Yellow River Conservancy Commission. Detailed information about the catchment characteristics and data durations are shown in Table 1. Daily meteorological data (1960–2012) comprised of precipitation, daily maximum and minimum temperatures, atmospheric pressures, wind speeds, mean relative humidity values, and sunshine durations, which were recorded at 96 stations, were provided by the China Meteorological Administration. The new NDVI third generation (NDVI3g) dataset was used to represent the vegetation characteristics of the study area, and detailed information about this dataset was presented earlier by Fensholt and Proud (2012). The maximum value compositing (MVC) procedure (Holben, 1986) was applied to produce the annual NDVI values.

Table 1. Long-term hydrometeorological characteristics and vegetation coverage (1981-2012)[a].

ID	Basin	Data length, year	*P*, mm/yr	ET_0, mm/yr	*ET*, mm/yr	ω	*M*	*S*
1	Huangfu	32	372	972	347	3.15	0.42	0.94
2	Gushan	32	424	1078	394	2.74	0.47	0.90
3	Kuye	32	375	1018	333	2.45	0.43	0.99
4	Tuwei	32	383	1031	308	1.99	0.41	0.95
5	Wuding	32	385	1045	356	2.68	0.46	0.95
6	Qingjian	32	451	1009	417	3.00	0.60	0.60
7	Yan	32	462	984	433	3.21	0.70	0.51
8	Beiluo	28	502	960	475	3.76	0.88	0.34
9	Jing	32	529	936	497	3.74	0.59	0.51
10	Fen	29	465	982	452	4.21	0.87	0.43
11	Xinshui	32	478	992	458	3.77	0.87	0.45
12	Sanchuan	24	444	998	397	2.70	0.57	0.58
13	Qiushui	23	442	1006	418	3.33	0.67	0.60

[a]Because a few runoff data points were missing for several basins, the data length in these basins was less than 32. Each item represents the mean annual value.

4. Results

4.1. The variability of parameter ω

The Budyko framework is usually used for analyses of long-term average data on catchment-scale water balances; however, in this study, it was employed for the interpretation of the interannual variability of the water balances by using the hydrological year approach described earlier. To validate the feasibility of using Fu's equation for interannual variability, the evapotranspiration ratio (ET/P) and dryness index (ET_0/P) on an annual scale for 13 basins are presented in the supporting information (Figure S1), and it can be seen that almost all points are focused on Fu's curves in each basin. Therefore, Fu's equation was considered adequate for the analysis of the interannual variability of the water balance.

If the controlling parameter ω on an annual scale can reflect the combined impacts of vegetation change and climate seasonality, it should also exhibit interannual variability with the seasonal variation in vegetation and climate, especially in those basins affected significantly by climate change and human activities. Obviously, this is true for basins in Loess Plateau (Figure 2). During 1961–2012, ω values in all 13 basins had an upward trend. Along with such a changing trend in ω, *ET* should increased for the same levels of *P* and ET_0. Before the 1980s, the variation in ω for each basin was relatively gentle; however, since that time, it has increased dramatically.

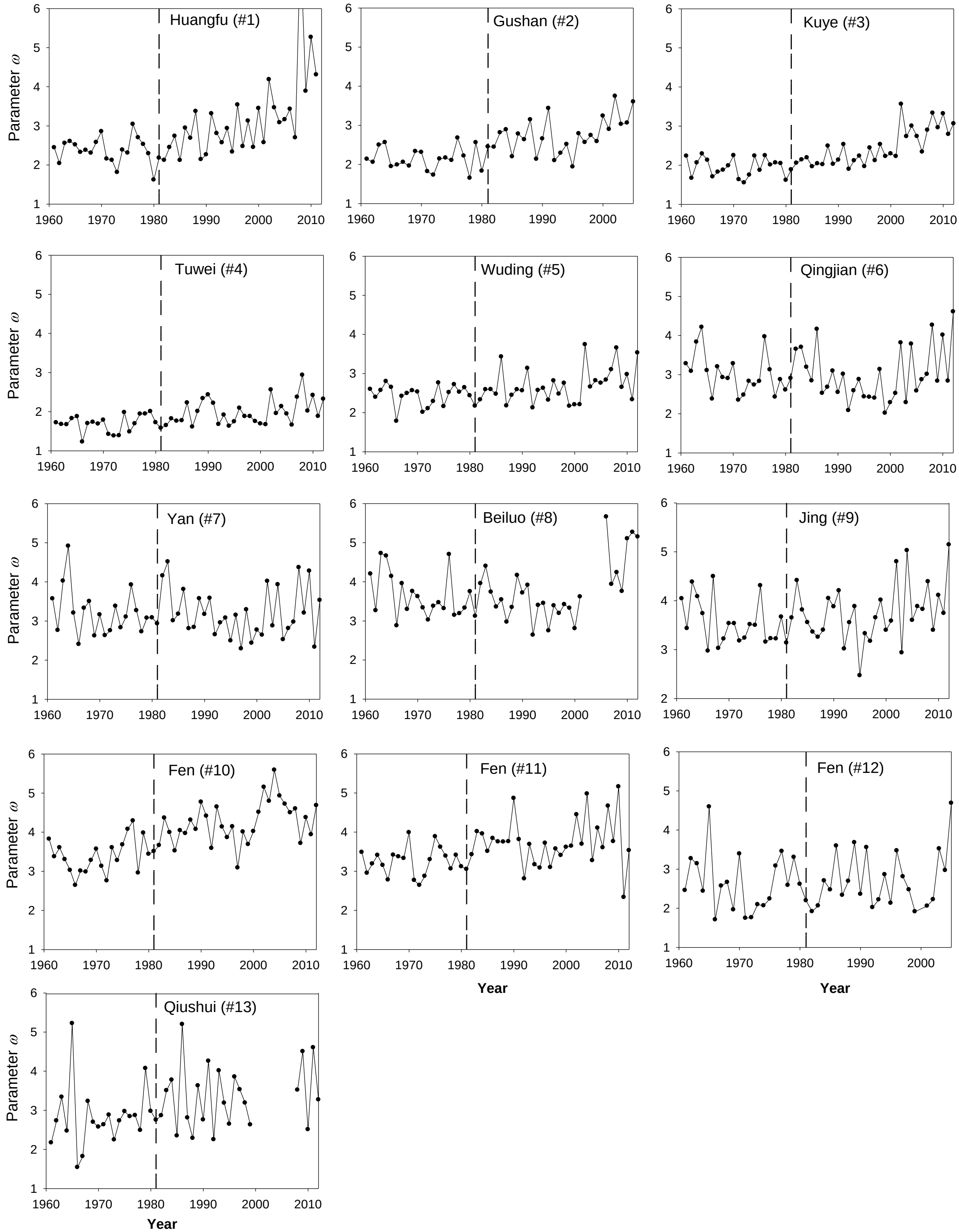
Huangfu (#1)
Gushan (#2)
Kuye (#3)
Tuwei (#4)
Wuding (#5)
Qingjian (#6)
Yan (#7)
Beiluo (#8)
Jing (#9)
Fen (#10)
Fen (#11)
Fen (#12)
Qiushui (#13)
Parameter ω
Year

Figure 2. The variability of parameter ω for 13 basins during 1961 to 2012.

4.2. Development of the semi-empirical formula for parameter ω

The relationships between the annual parameter ω and vegetation coverage M as well as the climate seasonality index S were first explored in each study basin during the period 1981–2012, and the results are shown in Figures S2 and S3. We can see that the parameter ω generally had a positive correlation with M, which implies that evapotranspiration increased with improvements in the vegetation conditions. However, ω was correlated negatively with S, which means that larger seasonal variations of coupled water and energy resulted in less evapotranspiration in this area. The relationships between ω and M as well as S imply that the annual variation in parameter ω can be estimated by the changes in vegetation dynamics and climate seasonality.

To expand the sample size and span a wider range of climate conditions, as well as to make the derived semi-empirical formula of parameter ω more representative, relationships were then developed based on the combined dataset from the 13 basins (Figure 3). These results also indicate a good relationship between ω and M ($R^2 = 0.40$, $p < 0.01$) as well as S ($R^2 = 0.28$, $p < 0.01$).

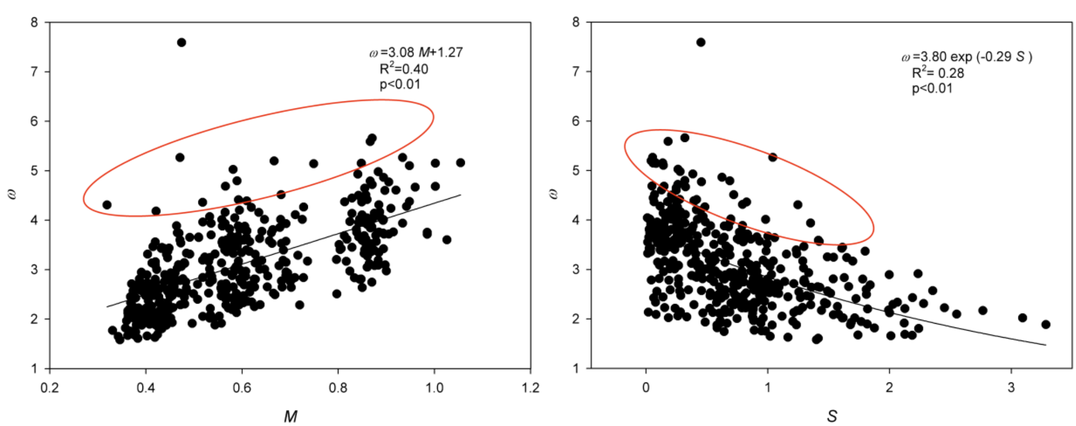

Figure 3. Relationships between the (a) annual ω and vegetation coverage (M) and (b) ω and climate seasonality index (S) based on the combined dataset from 13 basins.

To develop the semi-empirical formula of parameter ω, the limiting conditions of the two variables were considered as follows:

(1) If $M \to 0$, i.e. the land surface was bare, which indicates that the climate was extremely dry, $P \to 0$, $ET \to 0$, and thus, $\omega \to 1$;

(2) If $S \to \infty$, i.e. $\emptyset \to \infty$ and $\delta_{ET_0} \neq 0$ in the equation (3), which means monthly ET_0 is not uniform distributed within a year and $P \to 0$, thus $ET \to 0$, and $\omega \to 1$.

Considering the relationships shown in Figure 3 and given the above limiting conditions, the general form of parameter ω can be expressed as follows:

$$\omega = 1 + a \times M^b \times \exp(cS) \tag{10}$$

where a, b, and c are constants. Using the least linear square regression method, the semi-empirical formula of parameter ω is derived as follows:

$$\omega = 1 + 3.525 \times M^{0.783} \times \exp(-0.218\, S) \tag{11}$$

The coefficient of determination R^2 and the statistics for the F test of the modelled ω were 0.51 and 218.94, respectively.

A cross-validation approach was chosen to calibrate and test the above semi-empirical formula for parameter ω. Specifically, the dataset for the 13 basins in our study was separated into two groups. One was applied to build the semi-empirical formula, and it consisted of 12 basins for each time; the other was used for testing the performance of the semi-empirical formula, and it consisted of the remaining 1 basin. In total, the cross-validation process was conducted 13 times. After building the semi-empirical formula by using the vegetation coverage data and climate seasonality index data for the 12 basins, the parameter ω for the validated basin was modelled by using this fitted formula, and the annual ET for the validated basin was evaluated with the modelled ω, which is referred to as the "modelled" ET. Then, the “modelled” ET was compared with the ‘‘measured’’ ET.

Table 2 shows the cross-validation results for each basin. The model coefficients of each calibration formula for parameter ω were very close with the coefficients of Eq. (11), which means the relationship between ω and M as well as S we built is stable. Except for the basin #4 and #12, the MAE (mean absolute error) and RMSE (square root of the mean square error) values for each cross-validation process were relative low, with mean values of 13.5 mm and 16.8mm, respectively. The NSE coefficient (Nash–Sutcliffe coefficient of efficiency) for each process was greater than 0.8, thus suggesting that vegetation changes and climate seasonality can well explain the variation in the controlling parameter of the catchment water balance on the shorter time scale.

Table 2. Cross-validation results for each basin.

ID	Validated basin	Model coefficients			ET estimation accuracy		
		a	b	c	MAE	RMS	NSE
1	Huangfu	3.597	0.868	-0.228	22.3	23.8	0.88
2	Gushan	3.525	0.787	-0.231	16.3	21.3	0.90
3	Kuye	3.490	0.743	-0.233	17.4	22.7	0.88
4	Tuwei	3.350	0.627	-0.224	33.4	37.5	0.84
5	Wuding	3.525	0.803	-0.211	8.3	12.5	0.97
6	Qingjian	3.525	0.794	-0.206	13.9	18.1	0.96
7	Yan	3.560	0.803	-0.210	11.3	14.0	0.98
8	Beiluo	3.633	0.826	-0.213	10.2	11.9	0.97
9	Jing	3.456	0.814	-0.188	23.1	25.8	0.87
10	Fen	3.421	0.738	-0.223	6.3	8.9	0.98
11	Xinshui	3.560	0.803	-0.216	6.6	9.0	0.99
12	Sanchuan	3.561	0.782	-0.215	25.6	31.0	0.88
13	Qiushui	3.525	0.800	-0.204	12.5	16.4	0.96

4.3. Quantitative attribution of the variation in *ET*

The impacts of vegetation changes on ET have been widely studied with the Budyko framework by assuming surface conditions can be represented by the controlling parameter. However, according to the developed relationships in our study, the controlling parameter is not only related to surface condition change, but also to climate seasonality. The contributions of changes in climate (P, ET_0, and S) and vegetation (M) to the ET change were thus estimated by using the semi-empirical formula for parameter ω in the context of Fu's framework.

Trend in hydrometeorological variables and vegetation coverage were first analyzed for each basin (Table 3). ET_0 and S in all basins exhibited an upward trend, though with different significances. Similarly, M in most basins increased during past several decades. Based on the sensitivity coefficients of ET (Table S1) and the changes in mean annual P, ET_0, ω, M and S from period I to period II (Table 4), the changes in ET due to those in P, ET_0, M and S were estimated using the method described in Section 2.3. The contributions of four variables to ET change for each basin were presented in Table 4. In basin #1, 3-4 and #6, the ET changes were controlled by vegetation improvement; however, in the other basins, the dominant factor was precipitation. Except for basin #6, #9 and #12, elevated vegetation in most basins positively contributed to ET changes, which is consistent with Feng et al. (2016). ET in several basins showed a downward trend even though M positively contributed to ET changes; which is due to the offsetting effect of the other factors.

Table 3. Trend analysis for the hydrometeorological variables and vegetation coverage[b].

ID	Basin	ET,mm yr^{-2}	ET_0,mm yr^{-2}	P,mm yr^{-2}	M	S
1	Huangfu	1.89	1.16	0.61	0.002*	0.001
2	Gushan	0.76	3.85**	-0.01	0.004**	0.012
3	Kuye	2.34*	2.04*	0.53	0.004**	0.006
4	Tuwei	1.87	2.33**	0.53	0.005**	0.006
5	Wuding	0.88	1.17	0.31	0.006**	0.004
6	Qingjian	-0.45	1.78*	-0.94	0.007**	0.006
7	Yan	-1.62	2.03*	-1.99	0.005**	0.006
8	Beiluo	-5.4*	4.6*	-6.2*	0.0001	0.017
9	Jing	-0.97	1.47*	-1.79	0.002**	0.001
10	Fen	-0.72	1.93*	-1.16	0.002*	0.003
11	Xinshui	0.33	1.80	-0.12	0.003**	0.005
12	Sanchuan	1.49	1.84	0.09	-0.0004	0.004

13	Qiushui	-0.50	1.79	-0.83	0.002	0.008

[b]* and ** indicate the trend is significant at the level of $p = 0.05$ and $p = 0.01$ by the Mann–Kendall test, respectively.

It should be noted that the climate seasonality (represented by *S*) played an important role in the catchment *ET* variation. The contributions of *S* to *ET* changes ranged from 0.1% to 65.5% (absolute values). Besides basin #6, #9 and #12, the climate seasonality had a negative effect on *ET* variation in most of the basins, which means that larger seasonality differences between seasonal water and heat will lead to smaller amounts of evapotranspiration. Accordingly, if ω is supposed to only represent the landscape condition, the effects of landscape condition change on *ET* variation will be underestimated in basin #1, #3, #6-7, #9 and #11. Except for basin #9, the area of these basins is relative smaller; while its effects will be overestimated in the other basins, and the error would be equal to the contributions of *S* to *ET* changes.

Table 4. Attribution analysis for ET changes for each basin [c]

ID	Basin	Break point of ET	Change from Period 1 to Period 2					ET_0/ P/M/S induced ET change (mm)					Contribution to ET change (%)			
			ΔET	ΔET_0	ΔP	ΔM	ΔS	C_ (ET_0)	C_ (P)	C_ (ω)	C_ (M)	C_ (S)	φ_ (ET_0)	φ_ (P)	φ_ (M)	φ_ (S)
1	Huangfu	2001(ns)	41.7	7.0	22.2	0.03	0.01	0.28	18.67	22.70	22.73	-0.04	0.7	44.8	54.6	-0.1
2	Gushan	2000(ns)	33.6	64.9	20.6	0.07	-0.10	2.81	17.01	13.77	8.87	4.90	8.4	50.6	26.4	14.6
3	Kuye	2000(**)	51.4	32.0	17.3	0.06	0.05	1.54	13.34	36.48	55.95	-19.47	3.0	26.0	108.9	-37.9
4	Tuwei	2000(**)	43.2	39.6	24.0	0.07	-0.03	2.57	15.28	25.35	21.85	3.49	5.9	35.4	50.6	8.1
5	Wuding	2000(*)	35.2	17.6	26.9	0.09	-0.12	0.77	21.82	12.64	8.24	4.40	2.2	61.9	23.4	12.5
6	Qingjian	1988(**)	-50.1	32.0	-48.0	0.08	0.19	2.06	-37.80	-14.31	-47.09	32.78	-4.1	75.5	94.08	-65.5
7	Yan	1985(**)	-82.3	44.6	-86.9	0.05	0.30	3.19	-69.52	-15.96	22.19	-38.14	-3.9	84.5	-27.0	46.4
8	Beiluo	1985(**)	-65.1	49.4	-79.8	0.01	0.19	4.33	-62.9	-6.75	3.69	-10.43	-6.6	96.3	-5.7	16.0
9	Jing	1990(**)	-33.7	43.0	-47.8	0.03	0.11	4.1	-37.2	-0.61	-8.23	7.61	-12.2	110.3	24.4	-22.6
10	Fen	2005(ns)	23.1	8.5	21.2	0.07	-0.20	0.33	19.00	3.81	2.13	1.68	1.4	82.1	9.2	7.3
11	Xinshui	1990(**)	-19.1	39.7	-24.7	0.02	0.09	2.06	-21.08	-0.14	0.41	-0.55	-10.8	110.1	-2.1	2.9
12	Sanchuan	1996(ns)	-27.0	45.4	-43.4	-0.01	0.22	3.01	-32.52	2.56	0.20	2.36	-11.2	120.6	-0.7	-8.8
13	Qiushui	1996(ns)	-80.3	77.5	-103.5	-0.01	0.68	3.76	-83.68	-0.40	-0.02	-0.37	-4.7	104.2	0.1	0.5

[c]The relative contribution of a certain variable to the ET change ($\varphi(x)$) was calculated as follows: $\varphi(x) = (\mathrm{C}\_(x)/\Delta ET) \times 100\%$, where C_($x$) represents the contribution of each variable.

5. Discussion

Although the controlling parameter ω showed a good relationship with the vegetation change and climate seasonality index, two groups of deviations around the regressed curves were detected (Figure 3). The deviation points for the relationship between ω and M were mainly located at the top of the curve, i.e. corresponding to the same M values, where ω values were greater. We checked those points and found that precipitation and vegetation coverage in those years were normal, but runoff was very low compared to normal years. Excluding abrupt climate change, possible reasons for the extremely low runoff in those years include dam and reservoir operations, as well as irrigation diversions. A study conducted by Liang et al. (2015) on the same basins that we investigated in the Loess Plateau showed that check-dams increased continuously starting from the 1960s. By the year 2006, the numbers of dams along the basin #10 and #5 reached up to 482 and 181, respectively. Dams can intercept stormwater runoff for a short period during flood seasons and allow more time for infiltration (Polyakov et al., 2014). A total of 21 large and 136 medium-sized reservoirs were installed along the Yellow River by 2001. Such infrastructure can also influence the runoff change by controlling the flooding, regulating the water discharge, and diverting the water to other regions (Chen et al., 2005). Agricultural production is heavily dependent on irrigation throughout the entire Yellow River basin, and it has been reported that water consumption by agricultural irrigation accounted for nearly 80.0% of the entire water consumed from 1998 to 2011 (Wang et al., 2014). Thereby, water withdrawn for irrigation also plays an important role in the changing trends in runoff. In this study, the deviation points around the relationship curve between the annual ω and S fell in the upper left, and they were likely influenced by the low runoff. However, separation of the impacts on runoff from vegetation change, climate seasonality, and engineering works will have to await future work.

The relationships of parameter ω with vegetation dynamics and climate seasonality in some single basins were not significant in this study. Similarly, Yang et al. (2014a) also found a weak relationship between parameter n and vegetation coverage in 201 basins in China. This implies that the parameter ω might represent the combined effects of some other factors. For example, strong interactions among vegetation, climate, and soil conditions will lead to specific hydrologic partitioning at the catchment scale. In dry years, with low soil water contents, plants are trying to adapt by making use of hydrological processes, e.g. ground water dynamics and plant water storage mechanisms, etc. (Renger and Wessolek, 2010). Therefore, the relationship between the parameter ω and vegetation dynamics can be influenced by climate and soil conditions. However, it is difficult to separate the climatic and soil components from the vegetation change. Moreover, Zhang et al. (2001) reported that the impact of different vegetation types on catchment water balance can be vastly different, and the plant-available water coefficient in their function, which is similar to parameter ω in Fu's equation, is related to vegetation type. Therefore, the vegetation type may also be an important variable that influences the parameter ω.

Despite that catchment-scale water storage changes are usually assumed to be zero on long-term scale, the interannual variability of storage change can be an important component in annual water budget during dry or wet years (Wang and Alimohammadi, 2012), and cannot be ignored. However, the Loess Plateau has a subhumid to semiarid climate, the water storage and its annual variation are relatively small compared with humid regions (see Figure 5 from Mo et al., 2016). For example, using GRACE (Gravity Recovery and Climate Experiment), the water storage variations in the Yangtze, Yellow and Zhujiang from 2003 to 2008 were analyzed by Zhao et al. (2011), and the values for the Yangtze and Zhujiang basins were 37.8 mm and 65.2 mm, while no clear annual variations are observed in the Yellow River basin (3.0 mm). Furthermore, Mo et al. (2016) found that the water storage in Yellow River kept decreasing from 2004 to 2011, whereas it was changing slowly with a rate of 1.3 mm yr^{-1}. Therefore, considering the small water storage change in study area, ignoring water storage change in a period of hydrologic year is reasonable.

Errors still exhibited in the attribution analysis of ET changes. As the changes in evapotranspiration has been decomposed without residual by the complementary method (Equation 6-7), the errors were induced from the developed empirical formula for w (Equation 11). It suggested that ω cannot be completely explained by M and S, and it might include some other factors. Therefore, discussing more factors influencing ω remains future work.

6. Conclusions

This study explored the concomitant effects of vegetation dynamics and climate seasonality on the variation in interannual controlling parameter ω from Fu's equation within the Loess Plateau. First, to reduce the impact of ignoring the water storage change on annual catchment water balance, the hydrological year approach was introduced to examine the interannual variability of the controlling parameter ω for the 13 basins in the Loess Plateau from 1961 to 2012. The findings showed that parameter ω in all these basins presented an increasing trend, especially after the 1980s. Furthermore, we checked the relationship between ω and vegetation dynamics (represented by the annual vegetation coverage, M) as well as climate seasonality (represented by the climate seasonality index, S). The interannual changes of parameter ω were found to be related strongly to M and S. As such, a semi-empirical formula for the annual value of ω was developed based on these two parameters, and it was proven superior for estimating the actual evapotranspiration (ET) by a cross-validation approach. Finally, based on the proposed semi-empirical formula for parameter ω, the contributions of changes in climate (P, ET_0, and S) and vegetation (M) to ET variations were estimated. The results showed that the improved vegetation conditions in all basins made a positive contribution to the ET change, but these effects were largely offset by other variables in some basins. The contribution of landscape condition changes to *ET* variation will be estimated with a large error if the effects of climate seasonality were ignored.

Acknowledgments

This study was supported by the National Key Research and Development Program of China (No.2016YFC0501602), the National Natural Science Foundation of China (No. 41571036), and the Public Welfare Industry (Meteorological) Research Project of China (No. GYHY201506001).

References

Berghuijs, W. R., and Woods, R. A.: Correspondence: Space-time asymmetry undermines water yield assessment, Nature Communications, 7, 10.1038/ncomms11603, 2016.

Budyko, M. I.: Evaporation under natural conditions, Gidrometeorizdat, Leningrad, English translation by IPST, Jerusalem, 1948.

Budyko, M. I.: Determination of evaporation from the land surface (in Russian), Izvestiya Akad.nauk Sssr.ser.geograf.geofiz, 6, 3-17, 1961.

Budyko, M. I.: Climate and life, Academic, New York, 1974.

Carmona, A. M., Sivapalan, M., Yaeger, M. A., and Poveda, G.: Regional patterns of interannual variability of catchment water balances across the continental US: A Budyko framework, Water Resources Research, 50, 9177-9193, 10.1002/2014wr016013, 2014.

Chen, J. S., Wang, F. Y., Meybeck, M., He, D. W., Xia, X. H., and Zhang, L. T.: Spatial and temporal analysis of water chemistry records (1958-2000) in the Huanghe (Yellow River) basin, Global Biogeochemical Cycles, 19, 10.1029/2004gb002325, 2005.

Choudhury, B. J.: Evaluation of an empirical equation for annual evaporation using field observations and results from a biophysical model, Journal of Hydrology, 216, 99-110, 10.1016/s0022-1694(98)00293-5, 1999.

Donohue, R. J., Roderick, M. L., and McVicar, T. R.: On the importance of including vegetation dynamics in Budyko's hydrological model, Hydrol Earth Syst Sc, 11, 983-995, 10.5194/hess-11-983-2007, 2007.

Donohue, R. J., Roderick, M. L., and McVicar, T. R.: Can dynamic vegetation information improve the accuracy of Budyko's hydrological model?, Journal of Hydrology, 390, 23-34, 10.1016/j.jhydrol.2010.06.025, 2010.

Donohue, R. J., Roderick, M. L., and McVicar, T. R.: Roots, storms and soil pores: Incorporating key ecohydrological processes into Budyko's hydrological model, Journal of Hydrology, 436, 35-50, 10.1016/j.jhydrol.2012.02.033, 2012.

Feng, X., Vico, G., and Porporato, A.: On the effects of seasonality on soil water balance and plant growth, Water Resources Research, 48, W05543, 10.1029/2011wr011263, 2012.

Feng, X., Fu, B., Piao, S., Wang, S., and Ciais, P.: Revegetation in China's Loess Plateau is approaching sustainable water resource limits, Nature Climate Change, 10.1038/NCLIMATE3092, 2016.

Fensholt, R., and Proud, S. R.: Evaluation of Earth Observation based global long term vegetation trends - Comparing GIMMS and MODIS global NDVI time series, Remote Sensing of Environment, 119, 131-147, 10.1016/j.rse.2011.12.015, 2012.

Fu, B.: On the calculation of the evaporation from land surface (in Chinese), Scientia Atmospherica Sinica, 23-31, 1981.

Holben, B. N.: Characteristics of maximum-value composite images from temporal AVHRR data, International Journal of Remote Sensing, 7, 1417-1434, 1986.

Huang, M., and Zhang, L.: Hydrological responses to conservation practices in a catchment of the Loess Plateau, China, Hydrological Processes, 18, 1885-1898, 10.1002/hyp.1454, 2004.

Jiang, C., Xiong, L. H., Wang, D. B., Liu, P., Guo, S. L., and Xu, C. Y.: Separating the impacts of climate change and human activities on runoff using the Budyko-type equations with time-varying parameters, Journal of Hydrology, 522, 326-338, 10.1016/j.jhydrol.2014.12.060, 2015.

Li, D., Pan, M., Cong, Z., Zhang, L., and Wood, E.: Vegetation control on water and energy balance within the Budyko framework, Water Resources Research, 49, 969-976, 10.1002/wrcr.20107, 2013.

Li, D.: Assessing the impact of interannual variability of precipitation and potential evaporation on evapotranspiration, Advances in Water Resources, 70, 1-11, 10.1016/j.advwatres.2014.04.012, 2014.

Li, Z., Zheng, F. L., and Liu, W. Z.: Spatiotemporal characteristics of reference evapotranspiration during 1961-2009 and its projected changes during 2011-2099 on the Loess Plateau of China, Agricultural and Forest Meteorology, 154, 147-155, 10.1016/j.agrformet.2011.10.019, 2012.

Liang, W., Bai, D., Wang, F. Y., Fu, B. J., Yan, J. P., Wang, S., Yang, Y. T., Long, D., and Feng, M. Q.:Quantifying the impacts of climate change and ecological restoration on streamflow changes based on a Budyko hydrological model in China's Loess Plateau, Water Resources Research, 51,6500-6519, 10.1002/2014wr016589, 2015.

Liu, X., Liu, C., Luo, Y., Zhang, M., and Xia, J.: Dramatic decrease in streamflow from the headwater source in the central route of China's water diversion project: Climatic variation or human influence?, Journal of Geophysical Research: Atmospheres, 117, 10.1029/2011JD016879, 2012.

Liu, X. M., Luo, Y. Z., Zhang, D., Zhang, M. H., and Liu, C. M.: Recent changes in pan-evaporation dynamics in China, Geophysical Research Letters, 38, 10.1029/2011gl047929, 2011.

Milly, P. C. D.: An analystic solution of the stochastic storage problem applicable to soil-water, Water Resources Research, 29, 3755-3758, 10.1029/93wr01934, 1993.

Milly, P. C. D.: Climate,soil-water storage, and the average annual water-balance, Water Resources Research, 30, 2143-2156, 10.1029/94wr00586, 1994.

Mo, X., Wu, J. J., Wang, Q., and Zhou, H.: Variations in water storage in China over recent decades from GRACE observations and GLDAS, Natural Hazards and Earth System Sciences, 16, 469-482, 10.5194/nhess-16-469-2016, 2016.

Mu, X., Zhang, L., McVicar, T. R., Chille, B., and Gau, P.: Analysis of the impact of conservation measures on stream flow regime in catchments of the Loess Plateau, China, Hydrological Processes, 21, 2124-2134, 10.1002/hyp.6391, 2007.

Ning, T., Li, Z., Liu, W., and Han, X.: Evolution of potential evapotranspiration in the northern Loess Plateau of China:

recent trends and climatic drivers, International Journal of Climatology, 36, 4019-4028, 10.1002/joc.4611, 2016.

Peel, M. C., McMahon, T. A., and Finlayson, B. L.: Vegetation impact on mean annual evapotranspiration at a global catchment scale, Water Resources Research, 46, 10.1029/2009wr008233, 2010.

Pettitt, A.: A non-parametric approach to the change-point problem, Applied statistics, 126-135, 10.2307/2346729, 1979.

Polyakov, V. O., Nichols, M. H., McClaran, M. P., and Nearing, M. A.: Effect of check dams on runoff, sediment yield, and retention on small semiarid watersheds, Journal of Soil and Water Conservation, 69, 414-421, 10.2489/jswc.69.5.414, 2014.

Potter, N. J., Zhang, L., Milly, P. C. D., McMahon, T. A., and Jakeman, A. J.: Effects of rainfall seasonality and soil moisture capacity on mean annual water balance for Australian catchments, Water Resources Research, 41, W06007, 10.1029/2004wr003697, 2005.

Potter, N. J., and Zhang, L.: Interannual variability of catchment water balance in Australia, Journal of Hydrology, 369, 120-129, 10.1016/j.jhydrol.2009.02.005, 2009.

Priestley, C., and Taylor, R.: On the assessment of surface heat flux and evaporation using large-scale parameters, Monthly weather review, 100, 81-92, 1972.

Renger, M., and Wessolek, G.: Quantifying the impact of groundwater depth on evapotranspiration in a semi-arid grassland region, Hydrology & Earth System Sciences, 7, 787-806, 10.5194/hess-15-787-2011 , 2010.

Roderick, M. L., and Farquhar, G. D.: A simple framework for relating variations in runoff to variations in climatic conditions and catchment properties, Water Resources Research, 47, 10.1029/2010wr009826, 2011.

Rodriguez-Iturbe, I.: Ecohydrology: A hydrologic perspective of climate-soil-vegetation dynamics, Water Resources Research, 36, 3-9, 10.1029/1999wr900210, 2000.

Sankarasubramanian, A., Vogel, R. M., and Limbrunner, J. F.: Climate elasticity of streamflow in the United States, Water Resources Research, 37, 1771-1781, 10.1029/2000wr900330, 2001.

Sankarasubramanian, A., and Vogel, R. M.: Annual hydroclimatology of the United States, Water Resources Research, 38, 19-11-19-12, 10.1029/2001WR000619, 2002.

Schaake, J. C., and Waggoner, P.: From climate to flow, Climate change and US water resources., edited by P. E. Waggoner, chap. 8, pp. 177–206, John Wiley, N. Y, 1990.

Shao, Q. X., Traylen, A., and Zhang, L.: Nonparametric method for estimating the effects of climatic and catchment characteristics on mean annual evapotranspiration, Water Resources Research, 48, W03517, 10.1029/2010wr009610, 2012.

Sivapalan, M., Yaeger, M. A., Harman, C. J., Xu, X. Y., and Troch, P. A.: Functional model of water balance variability at the catchment scale: 1. Evidence of hydrologic similarity and space-time symmetry, Water Resources Research, 47, W02522, 10.1029/2010wr009568, 2011.

Stocker, T., Qin, D., Plattner, G.-K., Tignor, M., Allen, S. K., Boschung, J., Nauels, A., Xia, Y., Bex, V., and Midgley, P. M.: Climate change 2013: The physical science basis, Cambridge University Press Cambridge, UK, and New York, 2014.

Wang, D.: Evaluating interannual water storage changes at watersheds in Illinois based on long-term soil moisture and groundwater level data, Water Resources Research, 48, 10.1029/2011wr010759, 2012.

Wang, D., and Alimohammadi, N.: Responses of annual runoff, evaporation, and storage change to climate variability at the watershed scale, Water Resources Research, 48, W05546, 10.1029/2011wr011444, 2012.

Wang, F., Zhao, G. J., Mu, X. M., Gao, P., and Sun, W. Y.: Regime Shift Identification of Runoff and Sediment Loads in the Yellow River Basin, China, Water, 6, 3012-3032, 10.3390/w6103012, 2014.

Woods, R.: The relative roles of climate, soil, vegetation and topography in determining seasonal and long-term catchment dynamics, Advances in Water Resources, 26, 295-309, 10.1016/s0309-1708(02)00164-1, 2003.

Xu, X., Yang, D., Yang, H., and Lei, H.: Attribution analysis based on the Budyko hypothesis for detecting the dominant cause of runoff decline in Haihe basin, Journal of Hydrology, 510, 530-540, 10.1016/j.jhydrol.2013.12.052, 2014.

Yang, D., Sun, F., Liu, Z., Cong, Z., Ni, G., and Lei, Z.: Analyzing spatial and temporal variability of annual water-energy balance in nonhumid regions of China using the Budyko hypothesis, Water Resources Research, 43, W04426, 10.1029/2006WR005224, 2007.

Yang, D., Shao, W., Yeh, P. J. F., Yang, H., Kanae, S., and Oki, T.: Impact of vegetation coverage on regional water balance in the nonhumid regions of China, Water Resources Research, 45, 10.1029/2008wr006948, 2009.

Yang, H., Yang, D., Lei, Z., and Sun, F.: New analytical derivation of the mean annual water-energy balance equation, Water Resources Research, 44, 10.1029/2007wr006135, 2008.

Yang, H., Lv, H., Yang, D., and Hu, Q.: Seasonality of precipitation and potential evaporationanditsimpact on catchment water-energy balance (in Chinese), Journal of Hydroelectric Engineering, 31, 54-59+93, 2012.

Yang, H. B., Qi, J., Xu, X. Y., Yang, D. W., and Lv, H. F.: The regional variation in climate elasticity and climate contribution to runoff across China, Journal of Hydrology, 517, 607-616, 10.1016/j.jhydrol.2014.05.062, 2014a.

Yang, H. B., Yang, D. W., and Hu, Q. F.: An error analysis of the Budyko hypothesis for assessing the contribution of climate change to runoff, Water Resources Research, 50, 9620-9629, 10.1002/2014wr015451, 2014b.

Yokoo, Y., Sivapalan, M., and Oki, T.: Investigating the roles of climate seasonality and landscape characteristics on mean annual and monthly water balances, Journal of Hydrology, 357, 255-269, 10.1016/j.jhydrol.2008.05.010, 2008.

Zhang, L., Dawes, W. R., and Walker, G. R.: Response of mean annual evapotranspiration to vegetation changes at catchment scale, Water Resources Research, 37, 701-708, 10.1029/2000wr900325, 2001.

Zhang, L., Hickel, K., Dawes, W. R., Chiew, F. H. S., Western, A. W., and Briggs, P. R.: A rational function approach for estimating mean annual evapotranspiration, Water Resources Research, 40, 10.1029/2003wr002710, 2004.

Zhang, L., Potter, N., Hickel, K., Zhang, Y., and Shao, Q.: Water balance modeling over variable time scales based on the Budyko framework - Model development and testing, Journal of Hydrology, 360, 117-131, 10.1016/j.jhydrol.2008.07.021, 2008.

Zhang, S., Yang, H., Yang, D., and Jayawardena, A. W.: Quantifying the effect of vegetation change on the regional water balance within the Budyko framework, Geophysical Research Letters, 43, 1140-1148, 10.1002/2015gl066952, 2016.

Zhao, Q. L., Liu, X. L., Ditmar, P., Siemes, C., Revtova, E., Hashemi-Farahani, H., and Klees, R.: Water storage variations of the Yangtze, Yellow, and Zhujiang river basins derived from the DEOS Mass Transport (DMT-1) model, Science China-Earth Sciences, 54, 667-677, 10.1007/s11430-010-4096-7, 2011.

Zheng, H., Zhang, L., Zhu, R., Liu, C., Sato, Y., and Fukushima, Y.: Responses of streamflow to climate and land surface change in the headwaters of the Yellow River Basin, Water Resources Research, 45, 10.1029/2007wr006665, 2009.

Zhou, G., Wei, X., Chen, X., Zhou, P., Liu, X., Xiao, Y., Sun, G., Scott, D. F., Zhou, S., Han, L., and Su, Y.: Global pattern

for the effect of climate and land cover on water yield, Nature Communications, 6, 10.1038/ncomms6918, 2015.
Zhou, S., Yu, B., Zhang, L., Huang, Y., Pan, M., and Wang, G.: A new method to partition climate and catchment effect on the mean annual runoff based on the Budyko complementary relationship, Water Resources Research, 52, 7163-7177, 10.1002/2016wr019046, 2016.