A two-parameter Budyko function to represent conditions under which evapotranspiration exceeds precipitation

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Received: 8 June 2015 – Published in Hydrol. Earth Syst. Sci. Discuss.: 16 July 2015
Revised: 24 May 2016 – Accepted: 26 May 2016 – Published: 8 June 2016

Abstract. A comprehensive assessment of the partitioning of precipitation (P) into evapotranspiration (E) and runoff (Q) is of major importance for a wide range of socio-economic sectors. For climatological averages, the Budyko framework provides a simple first-order relationship to estimate water availability represented by the ratio E/P as a function of the aridity index (Ep/P, with Ep denoting potential evaporation). However, the Budyko framework is limited to steady-state conditions, being a result of assuming negligible storage change in the land–water balance. Processes leading to changes in the terrestrial water storage at any spatial and/or temporal scale are hence not represented. Here we propose an analytically derived modification of the Budyko framework including a new parameter explicitly representing additional water available to evapotranspiration besides instantaneous precipitation. The modified framework is comprehensively analyzed, showing that the additional parameter leads to a rotation of the original water supply limit. We further evaluate the new formulation in an example application at mean seasonal timescales, showing that the extended framework is able to represent conditions in which monthly to annual evapotranspiration exceeds monthly to annual precipitation.

1 Introduction

The Budyko framework serves as a tool to estimate mean annual water availability as a function of aridity. It is widely used and well-established within the hydrological community, due both to its simplicity and long history, combining experience from over a century of hydrological research. Budyko (1958, 1974) derived a formulation of the function based on findings of Schreiber (1904) and Ol’Dekop (1911), and several other formulations have also been postulated which are numerically very similar (Schreiber, 1904; Ol’Dekop, 1911; Turc, 1954; Mezentsev, 1955; Pike, 1964; Fu, 1981; Choudhury, 1999; Zhang et al., 2001, 2004; Porporato et al., 2004; Yang et al., 2008; Donohue et al., 2012; Wang and Tang, 2014; S. Zhou et al., 2015). Many of these formulations are empirically derived and only few are analytically determined from simple phenomenological assumptions (Fu, 1981; Milly, 1994; Porporato et al., 2004; Zhang et al., 2004; Yang et al., 2007; S. Zhou et al., 2015). Numerous studies further assess controls determining the observed systematic scatter within the Budyko space. This scatter is, however, inherent, being also justified by the existence of free parameters within analytically derived formulations of the Budyko curve (Fu, 1981; Choudhury, 1999; Zhang et al., 2004; Yang et al., 2007). A variety of catchment and climate characteristics such as vegetation (Zhang et al., 2001; Donohue et al., 2007; Williams et al., 2012; Li et al., 2013; G. Zhou et al., 2015), seasonality characteristics (Milly, 1994; Potter et al., 2005; Gentine et al., 2012; Chen et al., 2013; Berguijs et al., 2014), soil properties (Porporato et al., 2004; Shao et al., 2012; Donohue et al., 2012) and topographic controls (Shao et al., 2012; Xu et al., 2013) have been proposed to exert a certain influence on the scatter within the Budyko space. Also more complex approaches to combine various controls (Milly, 1994; Gentine et al., 2012; Donohue et al., 2012; Xu et al., 2013) have been considered. Nonetheless, until present no conclusive statement on controls determining the scatter within the Budyko space has been made. In a recent assessment, Greve et al. (2015) further suggested a probabilistic Budyko framework by assuming that the combined influence
of all possible controls is actually nondeterministic and follows a probability distribution instead.

In this study we make use of the formulation introduced by Fu (1981) and Zhang et al. (2004). They derived a functional form between \( E/P \) and \( \Phi = E_p/P \) at mean annual catchment scales analytically from simple physical assumptions,

\[
\frac{E}{P} = 1 + \Phi - (1 + (\Phi)^\omega)^{\frac{1}{\omega}},
\]

where \( \omega \) is a free model parameter. The original formulation introduced by Budyko (1958, 1974) is best represented by setting \( \omega = 2.6 \) (Zhang et al., 2004). The obtained function is subject to two physical constraints constituting both the water demand and supply limits. The water demand limit represents \( E \) being limited by \( E_p \), whereas the water supply limit determines \( E \) to be limited by \( P \) (see Fig. 1). To maintain the supply limit, steady-state conditions are required. Therefore, the storage term \( (dS/dt) \) in the land–water balance equation at catchment scales,

\[
\frac{dS}{dt} = P - E - Q
\]

is assumed to be zero, which is generally a valid assumption at mean annual scales. It is further important to note that groundwater flow is not included in Eq. (2) and is neglected throughout the following analysis. However, the assumption of negligible storage changes constitutes a major limitation to the original Budyko framework. As a consequence, the framework is not valid under conditions of additional storage water besides instantaneous \( P \) being available to \( E \) and \( E > P \). We note here that by instantaneous \( P \) (from here on just referred to as \( P \)) we mean all \( P \) within the considered time interval. Conditions under which the framework is not valid can occur, e.g., at subannual or interannual timescales due to changes in terrestrial water storage terms such as soil moisture, groundwater or snow storage. Additional water might be also introduced by landscape changes (Jaramillo and Destouni, 2015), human interventions (Milly et al., 2008) or phase changes of water within the system or supplied through precipitation (Jaramillo and Destouni, 2015; Berghuijs et al., 2014). Also long-term changes in soil moisture may happen, e.g., under transient climate change (Wang, 2005; Orlowsky and Seneviratne, 2013). Only few assessments have addressed this limitation and provided further insights on how the Budyko hypothesis could be extended to conditions under which \( E \) exceeds \( P \) (Zhang et al., 2008; Chen et al., 2013). Nonetheless, so far a theoretical incorporation of conditions in which \( E > P \) into the Budyko framework is lacking. Here we aim to address this issue by analytically deriving a new, modified Budyko formulation from basic phenomenological assumptions by using the approach of Fu (1981) and Zhang et al. (2004).

2 Deriving a modified formulation

2.1 Preliminary assumptions

In the following we will make use of the concept of potential evapotranspiration, which provides an estimate of the amount of water that would be evaporated under conditions of a well-watered surface. Fu (1981) and Zhang et al. (2004) suggested that, for a given potential evaporation, the rate of change in evapotranspiration as a function of the rate of change in precipitation \( (\partial E/\partial P) \) increases with residual potential evaporation \( (E_p - E) \) and decreases with precipitation. Similar assumptions were made regarding the rate of change in evapotranspiration as a function of the rate of change in potential evaporation \( (\partial E/\partial E_p) \) by considering residual precipitation \( (P - E) \). Hence, both ratios can be written as

\[
\frac{\partial E}{\partial P} = f(x) \quad (3a)
\]

\[
\frac{\partial E}{\partial E_p} = g(y) \quad (3b)
\]

with

\[
x = \frac{E_p - E}{P} \quad (4a)
\]

\[
y = \frac{P - E}{E_p} \quad (4b)
\]

Considering \( E_p \) to be a natural constraint of \( E \), it follows that

\[
\frac{\partial E}{\partial P} \bigg|_{x=0} = 0. \quad (5)
\]

The original approach of Fu (1981) further assumes that \( P \) is a natural constraint of \( E \), constituting the following boundary condition:

\[
\frac{\partial E}{\partial E_p} \bigg|_{y=0} = 0. \quad (6)
\]

The coupled boundary conditions Eqs. (5) and (6) mathematically represent the supply and demand limit of the Budyko framework.
framework (see Fig. 1). Considering the definitions of \( x \) and \( y \) given by Eqs. (4a) and (4b), \( x = 0 \) yields \( E = E_p \) and \( y = 0 \) yields \( E = P \). Equation (5) thus states that conditional upon \( x = 0 \), i.e., \( E = E_p \), no further change in \( E \) occurs no matter how \( P \) changes, since \( E \) is already limited by \( E_p \) (constituting the demand limit). Equation (6) states that conditional upon \( y = 0 \), i.e., \( E = P \), no further change in \( E \) occurs no matter how \( E_p \) changes, since \( E \) is already limited by \( P \) (constituting the supply limit). If \( x \neq 0 \) or \( y \neq 0 \), the gradients \( \partial E/\partial P \) or \( \partial E/\partial E_p \) are not (necessarily) zero.

The boundary condition Eq. (6) further requires steady-state conditions and is consequently considered to be valid at mean annual catchment scales (such that \( P - E \geq 0 \)) only. However, as mentioned in the Introduction, a wealth of possible mechanisms and processes can induce conditions in which \( E \) exceeds \( P \). In such cases, \( E_p \) remains the only constraint of \( E \). Consequently, since we explicitly aim to account for conditions of \( E \geq P \), the value \( y = (P - E)/E_p \) (see Eqs. 4a and 4b) is not necessarily positive but is larger than \(-1 \) since we assume that \( E \leq E_p \). The minimum value of \( y \), denoted as \( y_{\min} \), thus lies within the interval between \(-1 \) and \( 0 \) and depends on the additional amount of water being available for \( E \) besides water supplied by \( P \). For convenience we define \( y_0 = -y_{\min} \) (and thus \( y_0 \in [0,1] \)). As a consequence the boundary condition Eq. (6) is then redefined as

\[
\frac{\partial E}{\partial E_p} \bigg|_{-y_0} = 0. \tag{7}
\]

2.2 Solution

Solving the system of the differential Eqs. (3a) and (3b) using boundary condition Eq. (5) and the new condition Eq. (7) yields the following solution (details are provided in Appendix A):

\[
E = E_p + P - \left( 1 - y_0 \right)^{\kappa - 1} E_p^\kappa + P^\kappa \right)^{\frac{1}{2}}, \tag{8}
\]

with \( \kappa \) being a free model parameter. It follows that

\[
\frac{E}{P} = F(\Phi, \kappa, y_0) = 1 + \Phi - \left( 1 + (1 - y_0)^{\kappa - 1} (\Phi)^\kappa \right)^{\frac{1}{2}}, \tag{9}
\]

Similar to the traditional Budyko approach a free model parameter (named \( \kappa \) to avoid confusion with the traditional \( \omega \)) is obtained. The second parameter, \( y_0 \), as introduced in the previous section, is directly related to the new boundary condition. Hence, in contrast to \( \kappa \), which is a mathematical constant, \( y_0 \) has a physical interpretation as it accounts for additional water (i.e., storage water). However, similar to the \( \omega \) parameter in Fu’s equation, \( \kappa \) can be interpreted as an integrator of the variety of factors other than the aridity index that influence the partitioning of \( P \) into \( Q \) and \( E \).

The newly derived formulation given (Eq. 9) is similar to the classical solution (Eq. 1) but includes \( y_0 \) as a new parameter. For different values of \( y_0 \) and \( \kappa = 2.6 \) (corresponding to the best fit to the original Budyko function with \( \omega = 2.6 \) in Fu’s equation), Fig. 2 shows a set of curves providing insights on the basic characteristics of the modified equation.

If \( y_0 = 0 \) (being the original boundary condition), the obtained curve corresponds to the steady-state framework of Fu (1981) and Zhang et al. (2004). This shows that both model formulations are consistently transferable. If \( y_0 > 0 \), the supply limit is systematically exceeded. The exceedance of the supply limit increases with increasing \( y_0 \). If \( y_0 = 1 \), the curve follows the demand limit. All curves are further continuous and strictly increasing.

Taking a closer look at the underlying boundary conditions and definitions (see Sect. 2.1) reveals that \( y_0 \) explicitly accounts for the maximum amount of additional water (besides water supplied through \( P \)) at a certain location and within a certain time period that is available to \( E \). Since \( y_{\min} \) is defined to be the minimum of \( y = (P - E)/E_p \), the quantity \( y_0 = -y_{\min} \) physically represents the maximum fraction of \( E \) relative to \( E_p \) which does not originate from \( P \). A larger fraction consequently results in higher \( y_0 \) values and thus in a stronger exceedance of the original supply limit. Further details on \( y_0 \) are provided in Sect. 4.

The partial derivative \( \partial f(\Phi, \kappa, y_0)/\partial \Phi \) under varying \( \kappa \) and for three preselected values of \( y_0 \) is illustrated in Fig. 3. The sensitivity \( \partial f(\Phi, \kappa, y_0)/\partial \Phi \) for different values of \( y_0 \) and \( \kappa \) shows the effect of the parameter choice on changes in \( E/P \) relative to changes in \( \Phi \). In general, the sensitivity is largest for small \( \Phi \) (humid conditions), due to the fact that changes in \( E/P \) basically follow the demand limit (resulting in a sensitivity close to 1) regardless of parameter set \( (\kappa, y_0) \). For different parameter settings, the sensitivity generally decreases with increasing \( \Phi \). For small values of \( y_0 \) (close
to 0), sensitivity becomes smallest with increasing \( \Phi \), since small values of \( y_0 \) indicate conditions similar to the classical solution (Eq. 1). Further, the smallest sensitivity is reached for large values of \( \kappa \). Large values of \( y_0 \) (close to 1) indicate conditions mainly constrained by the demand limit, thus implying a sensitivity close to 1.

A similar analysis is performed for varying values of \( \kappa \) under three preselected levels of \( y_0 \) (see Fig. 4). For \( y_0 = 0 \) (steady-state conditions), the sensitivity \( \partial f / \partial \Phi \) is rather large under humid conditions (\( \Phi < 1 \)), since changes in \( E / P \) are mainly constrained by demand limit. This especially applies for large values of \( \kappa \). Under more arid conditions (\( \Phi > 1 \)), the Budyko curve slowly converges towards the (horizontal) supply limit, resulting in a near-zero sensitivity. For \( y_0 = 0.2 \), denoting conditions relatively similar to steady-state conditions, the decrease in sensitivity with increasing \( \Phi \) is weaker, whereas for \( y_0 = 0.8 \), denoting conditions where \( E \) is mainly constrained by the demand limit, sensitivity is large for large \( \kappa \) values and decreases rather slowly with increasing \( \Phi \).

### 4 Interpreting the new parameter \( y_0 \)

The new parameter \( y_0 \) is, in contrast to \( \kappa \), physically well defined. The combination of Eqs. (4b) and (7) shows that \( y_0 \) is explicitly related to the amount of additional water (besides water supplied through \( P \)) which is available to \( E \). If we rewrite Eq. (4b) with respect to \( y_0 \),

\[
y_0 = -y_{\text{min}} = -\left( \frac{P - E}{E_p} \right)_{\text{min}} \leq -\frac{P_{\text{min}} - E_{\text{max}}}{E_p},
\]

if \( P_{\text{min}} - E_{\text{max}} < 0 \),

(10)

where \( P_{\text{min}} \) and \( E_{\text{max}} \) are chosen in order to minimize \( y_{\text{min}} \) for a given \( E_p \), we obtain a linear equation in terms of aridity index

\[
\left( \frac{E}{P} \right)_{\text{max}} = y_0 \left( \frac{E_p}{P_{\text{min}}} \right) + 1,
\]

(11)

which constitutes the mathematical interpretation of \( y_0 \) within the modified framework; i.e., \( y_0 \) determines the maximum slope of the upper limit with which the obtained curve
from Eq. (9) asymptotically converges if $\kappa \to \infty$ (see Fig. 5). Physically, $y_0$ determines the maximum $E/P$ that is reached in relation to $\Phi$ within a certain time period and spatial domain. It thus represents an estimate of the maximum amount of additional water that contributes to $E$ and originates from other sources than $P$. Technically speaking, $y_0$ determines the slope of the upper limit such that all possible pairs $(\Phi, E/P)$ are just below the line $y_0 \Phi + 1$. It is further important to note that, for mean annual conditions ($P - E \geq 0$), $y_0 = 0$ is considered, which results in a zero slope and thus determines the original supply limit of Eq. (1). Please also note that this approach is not valid if $P_{\text{min}} = 0$.

However, the actual slope $m$ of the upper limit is smaller than $y_0$ but directly related to both $y_0$ and $\kappa$ as follows (see Appendix B for more information):

$$m = 1 - (1 - y_0)^{1/\kappa}.$$  

(12)

The relative difference between the maximum slope $y_0$ and the actual slope $m$ of the upper limit (being the ratio of $y_0/m$) is thus determined following the relationship

$$\frac{y_0}{m} = (1 - y_0)^{1/\kappa}.$$  

(13)

The ratio $y_0/m$ as a function of both $y_0$ and $\kappa$ is illustrated in Fig. 6. For small $\kappa$ and large $y_0$ (close to 1), the difference between the actual slope $m$ and the maximum slope $y_0$ is large, whereas for large $\kappa$ the actual slope $m$ converges towards $y_0$. However, in any case, $y_0$ determines the maximum overshoot allowed with respect to the original supply limit at $y_0 = 0$.

5 Example application: seasonal carryover effects in terrestrial water storage

At monthly timescales, changes in terrestrial water storage (due to changes in water storage components such as soil moisture, snow or groundwater) potentially play an important role in $E$ and $Q$ and are by no means negligible. Such changes can provide a significant source of additional water that is (besides $P$) available to $E$. Here we analyze the climatological mean seasonal cycle of $E/P$ by using gridded, monthly data estimates of $P$, $E$ and $E_p$. This allows us to evaluate the capability of the obtained framework (given by Eq. 9) to represent additional water sources at such timescales.

We employ the following well-established, gridded data products: (i) the Global Precipitation Climatology Project (GPCP) $P$ dataset (Adler et al., 2003), (ii) an $E_p$ estimate (Sheffield et al., 2006, 2012) based on the Penman–Monteith $E_p$ algorithm (Monteith, 1965; Sheffield et al., 2012) with the stomatal conductance set to 0 and aerodynamic resistance defined after Maidment (1992) and (iii) the LandFlux-Eval $E$ dataset (Mueller et al., 2013). All data are bilinearly interpolated to a unified 1° grid, and the mean seasonal cycle for the 1990–2000 period is calculated at grid point scale. Please note that the combination of datasets used here is arbitrary and only used to illustrate the capability of the newly developed framework to represent the climatological mean annual cycle of $E/P$.

We estimate the parameter set $(\kappa, y_0)$ from Eq. (9) by minimizing the residual sum of squares (see Fig. 7). This means that at every grid point 12-monthly climatologies of $E/P$ (representing the mean seasonal cycle of $E/P$) are used to determine one specific parameter set (for all months).

To evaluate the modified framework, the derived parameter sets at each grid point are used in Eq. (9) to compute mean seasonal cycles of $E/P$. The correlation between the computed and the observed seasonal cycle is shown in Fig. 8a. The correlations are relatively large in most regions. Largest correlations ($> 0.9$) are found in most mid- to high-latitude and tropical areas, clearly showing the capability of the modified formulation to represent the seasonal cycle in $E/P$. Correlations are generally somewhat lower in drier regions, especially in parts of Africa and central Asia, probably oc-
Figure 7. Estimated values of $\kappa$ (a) and $\gamma_0$ (b) estimated in a least squares fitting using standard monthly datasets of $P$, $E$ and $E_p$ within the 1990–2000 period.

Figure 8. Correlation between the mean seasonal cycle of $E/P$ computed from Eq. (9) and observed $E/P$ for (a) a grid-point-specific parameter set ($\kappa$, $\gamma_0$) and (b) ($\kappa$, 0) (Fu’s equation).

Figure 9. Data cloud of monthly climatologies within the Budyko space for a grid points in (a) central Europe (51.5° N, 12° E) and (b) central Africa (5.5° N, 20° E). The black solid line denotes the demand limit; the dashed line denotes the original supply limit. The blue line depicts the obtained curve using the modified formulation of Fu’s equation, whereas the red line shows the original Fu curve. Numbers within the dots denote the particular month of the year. (c, d) Observed (grey) and computed mean seasonal cycles at both grid points. The blue line depicts the obtained seasonal cycle using the modified formulation of Fu’s equation, whereas the red line shows the seasonal cycle obtained using Fu’s equation. Please note that axes are different in each plot.

...currying due to more complex seasonal patterns in $E/P$ and phenology, which are not considered here. Using instead Fu’s original equation (or setting $\gamma_0 = 0$) to estimate the mean seasonal cycle of $E/P$ shows overall lower correlations, especially in semi-arid regions (Fig. 8b).

Taking a closer look at the mean seasonal cycle for example grid points in (i) central Europe (humid climate) and (ii) Africa (semi-arid climate) clearly shows the improvement gained through the use of the modified formulation (Fig. 9). In central Europe, additional water is available in the early summer months due to, e.g., depletion of soil moisture or snow melt, resulting in values of $E/P$ exceeding the original supply limit. The modified formulation has the ability to represent this exceedance, whereas the original formulation...
is naturally bounded to 1. This is even more evident for the example grid point in Africa, showing a large overshoot of the original supply limit under dry-season conditions.

6 Conclusions

In conclusion we present an extension to the Budyko framework that explicitly accounts for conditions under which $E$ is also driven by other water sources than $P$ (i.e., changes in water storage). The original Budyko framework is limited to mean annual catchment scales at which $P$ and $E_p$ constitute natural constraints of $E$. Here we assume that the boundary condition constituted by $E_p$ remains overall valid, whereas the boundary condition constituted by $P$ is also subject to additional water stemming from other sources. Such additional water could, e.g., originate from changes in the terrestrial water storage, landscape changes and human interventions.

In order to account for such additional water, we modified the set of equations underlying the derivation of Fu’s equation (Fu, 1981; Zhang et al., 2004) and obtained a similar formulation including an additional parameter. The additional parameter is physically well defined and technically rotates the original supply limit upwards. Similar to the original Budyko framework, the derived two-parameter Budyko model represents the influence of first-order controls (namely $P$ and $E_p$) on water availability. The integrated influence of second-order controls (like vegetation, topography, etc.) is, comparable to Fu’s equation, represented by the first parameter. Analysis of such controls in Fu’s formula has been undertaken in numerous studies, but no conclusive assessment has been conducted until present. Assessing the combined influence of climatic and catchment controls is hence clearly beyond the scope of this study. However, the additional second parameter of the modified formulation $y_0$ does have a clear physical interpretation as it represents a measure of the maximum amount of additional water (besides $P$) available to $E$ at a certain location and within a particular time period.

Besides this study, a limited number of previous studies have assessed the Budyko hypothesis under conditions of $E$ exceeding $P$, especially at seasonal timescales. In a top-down approach, Zhang et al. (2008) showed that the Budyko model has to be extended in order to model the water balance on shorter than mean annual timescales. Their extended Budyko model (which was also based on Fu, 1981) showed good performance in modeling monthly $Q$ but includes four additional parameters that require extensive calibration. Chen et al. (2013) further introduced an approach (referring to Wang, 2012) that is based on replacing $P$ by effective precipitation, which is the difference between $P$ and soil water storage change. This allows the framework to be extended to seasonal timescales but requires explicit knowledge of changes in the soil water storage. In our approach, however, we provide an analytical derivation of an extension to Fu’s equation that is able to account for conditions under which $E$ exceeds $P$ by including only one additional parameter. However, the framework is also subject to some limitations. The estimation of the parameter $y_0$ is, similar to the estimation of the $\omega$ in Fu’s equation (Fu, 1981), nontrivial, and the parameter apparently varies in space and potentially also in time, therefore questioning steady-state assumptions. The framework is further not capable of directly estimating $Q$. Since in contrast to the original Budyko framework changes in terrestrial water storage are not negligible, the runoff ratio $Q/P$ can not be assessed through $1 - E/P$. Hence, explicit knowledge of changes in the terrestrial water storage is required, therefore aggravating assessments of $Q$.

The new framework was validated for the special case of average seasonal changes in water storage by using monthly climatologies of global, gridded standard estimates of $P$, $E$ and $E_p$. The computed grid-point-specific seasonal cycle of $E/P$ using the modified framework did adequately represent mean seasonal storage changes for many parts of the world. However, the application of the modified framework is by no means limited to this case and could be extended to a variety of climatic conditions under which additional water besides $P$ is available to $E$.
Appendix A: Complete solution

Equations (3a), (3b), (5) and (7) form a system of differential equations. A necessary condition to solve this system is

\[
\frac{\partial f(x)}{\partial E_p} + \frac{\partial f(x)}{\partial E} g(y) = \frac{\partial g(y)}{\partial P} + \frac{\partial g(y)}{\partial E} f(x).
\]  
(A1)

Combining Eq. (A1) with Eqs. (4a) and (4b) yields

\[
\frac{\partial f(x)}{\partial E_p} = \frac{\partial f(x)}{\partial E_p} \frac{\partial x}{\partial x} = \frac{1}{P} \left( 1 - \frac{\partial E}{\partial E_p} \right) \frac{\partial f(x)}{\partial x},
\]

(A2a)

\[
\frac{\partial f(x)}{\partial E} = \frac{\partial f(x)}{\partial E} \frac{\partial x}{\partial x} = \frac{1}{P} \left( \frac{\partial E_p}{\partial E} - 1 \right) \frac{\partial f(x)}{\partial x},
\]

(A2b)

\[
\frac{\partial g(y)}{\partial P} = \frac{\partial g(y)}{\partial P} \frac{\partial y}{\partial y} = \frac{1}{E_p} \left( 1 - \frac{\partial E}{\partial P} \right) \frac{\partial g(y)}{\partial y},
\]

(A2c)

\[
\frac{\partial g(y)}{\partial E} = \frac{\partial g(y)}{\partial E} \frac{\partial y}{\partial y} = \frac{1}{E_p} \left( \frac{\partial P}{\partial E} - 1 \right) \frac{\partial g(y)}{\partial y}.
\]

(A2d)

Substituting the factors in Eq. (A1) with those given in Eq. (A1) gives

\[
\frac{\partial f(x)}{\partial x} \left[ 1 - g(y) \right] + \left( \frac{1}{g(y)} - 1 \right) g(y) = \frac{P \frac{\partial g(y)}{\partial y}}{E_p} \left[ 1 - f(x) \right] + \left( \frac{1}{f(x)} - 1 \right) f(x).
\]

Expanding \( P/E_p \) with consideration given to Eqs. (4a) and (4b) yields

\[
\frac{P}{E_p} = \frac{E_p + P - E}{E_p} = \frac{1 + \frac{P - E}{E_p}}{1 + \frac{E_p - E}{P}} = \frac{1 + y}{1 + x}.
\]

(A4)

From Eqs. (A3) and (A4) follows

\[
(1 - g(y)) \frac{\partial f(x)}{\partial x} = \frac{1 + y}{1 + x} \left( 1 - f(x) \right) \frac{\partial g(y)}{\partial y} + \frac{1 + y}{1 + x} \frac{\partial g(y)}{\partial E} f(x),
\]

(A5)

where each side is a function of \( x \) or \( y \) only. Assuming the result of each side is \( \alpha \), it follows that

\[
\frac{1 + x}{1 - f(x)} \frac{\partial f(x)}{\partial x} = \alpha,
\]

\[
\frac{1 + y}{1 - g(y)} \frac{\partial g(y)}{\partial y} = \alpha.
\]

(A6a)

(A6b)

Integrating Eq. (A6a) with consideration given to the boundary condition given by Eq. (7) leads to the following expression for \( f(x) \):

\[
\int_0^x \frac{1}{1 - f(t)} \frac{\partial f(t)}{\partial t} dt = \alpha \int_0^x \frac{1}{1 - t} dt
\]

\[
[- \ln (1 - f(t))]_0^x = \alpha [\ln (1 + t)]_0^x
\]

\[
\ln (1 - f(x)) = -\alpha \ln (1 + x) 1 - f(x) = (1 + x)^{-\alpha}
\]

\[
f(x) = 1 - (1 + x)^{-\alpha}.
\]

(A7)

Integrating Eq. (A6b) is different from the traditional solution given in Zhang et al. (2004), as we are using the new boundary condition given by Eq. (7).

\[
\int_{-y_0}^{y} \frac{1}{1 - g(t)} \frac{\partial g(t)}{\partial t} dt \alpha \int_{-y_0}^{y} \frac{1}{1 - t} dt
\]

\[
[- \ln (1 - g(t))]_{-y_0}^{y} = \alpha [\ln (1 + t)]_{-y_0}^{y}
\]

\[
l \ln (1 - g(y)) = \alpha \ln \left( \frac{1 - y_0}{1 + y} \right)
\]

\[
g(y) = \frac{1 - y_0}{1 + y}
\]

\[
\alpha.
\]

(A8)

Considering the expansion from Eq. (A4) finally gives

\[
\frac{\partial E}{\partial P} = 1 - (1 + x)^{-\alpha} = 1 - \left( \frac{P}{E_p + P - E} \right)^{\alpha},
\]

(A9)

\[
\frac{\partial E}{\partial E_0} = 1 - (1 - y_0)^{\alpha} (1 + y)^{-\alpha}
\]

\[
= 1 - (1 - y_0)^{\alpha} \left( \frac{E_0}{E_0 + P - E} \right)^{\alpha}.
\]

(A10)

In the next step, Eq. (A9) is integrated over \( P \). As Eq. (A9) is identical to those in Zhang et al. (2004), we follow their substitution approach. It follows that

\[
E = E_0 + P = \left( k + P^{\alpha+1} \right) \frac{1}{\alpha+1},
\]

(A11)
where \( k \) is a function of \( E_0 \) only. Differentiating Eq. (A11) with respect to \( E_0 \) gives an estimate of \( \partial E / \partial E_0 \), which when used with Eq. (A10) determines \( k \):

\[
\frac{\partial E}{\partial E_0} = 1 - \frac{1}{\alpha + 1} \left( k + P^{\alpha + 1} \right)^{\frac{\alpha}{\alpha + 1}} \frac{\partial k}{\partial E_0} = 1 - (1 - y_0)^\alpha \left( \frac{E_0}{E_0 + P - E} \right)^{\alpha}.
\]

(A12)

This leads, with consideration given to Eq. (A11), to the following expression:

\[
\frac{\partial k}{\partial E_0} = (\alpha + 1)(1 - y_0)^\alpha \left( \frac{E_0}{E_0 + P - E} \right)^{\alpha} \left( k + P^{\alpha + 1} \right)^{\frac{\alpha}{\alpha + 1}}
\]

\[
= (\alpha + 1)(1 - y_0)^\alpha \int E_0^\alpha dE_0
\]

\[
k = (1 - y_0)^\alpha E_0^\alpha + C,
\]

(A13)

with \( C \) being an integration constant. By substituting Eq. (A13) back into Eq. (A11), one obtains the following expression:

\[
E = E_0 + P - \left( (1 - y_0)^\alpha E_0^{\alpha + 1} + C + P^{\alpha + 1} \right)^{\frac{1}{\alpha + 1}}.
\]

(A14)

As \( \lim_{P \to 0} E = 0 \), it follows that \( C = 0 \). Substituting \( \kappa = \alpha + 1 \) finally gives

\[
E = E_p + P - \left( (1 - y_0)^{\kappa - 1} E_p^\kappa + P^\kappa \right)^{\frac{1}{\kappa}}
\]

(A15)

and further provides by writing \( \Phi = E_p / P \)

\[
\frac{E}{P} = 1 + \Phi - \left( 1 + (1 - y_0)^{\kappa - 1} (\Phi)^\kappa \right)^{\frac{1}{\kappa}}
\]

(A16)

\[
F \left( \frac{E}{E_p}, \kappa, y_0 \right) = \frac{E}{E_p} = 1 + \frac{P}{E_p}
\]

\[
- \left( (1 - y_0)^{\kappa - 1} + \left( \frac{P}{E_p} \right)^\kappa \right)^{\frac{1}{\kappa}}.
\]

(A17)

**Appendix B: Solution of the actual slope**

The actual slope \( m \) of the upper limit with which the obtained Budyko curve converges is smaller than \( y_0 \). We introduced Eq. (12) to calculate \( m \), and in the following we provide the complete solution in order to obtain Eq. (12).

The value of \( m \) is the slope of the linear function \( m \Phi + 1 \) that forms the asymptote to \( F(\Phi, \kappa, y_0) \) given by Eq. (9). Hence,

\[
\lim_{\Phi \to \infty} \left[ F(\Phi, \kappa, y_0) - (m \Phi + 1) \right] = 0.
\]

(B1)

Using Eq. (9) and dividing by \( \Phi \) yields

\[
\lim_{\Phi \to \infty} \left[ \frac{1 + (1 - y_0)^{\kappa - 1} (\Phi)^\kappa}{\Phi} + 1 - m \right] = 0.
\]

(B2)

By raising the term in brackets to the power of \( \kappa \), one obtains

\[
\lim_{\Phi \to \infty} \left[ (1 - m)^\kappa - (1 - y_0)^{\kappa - 1} - \Phi^{-\kappa} \right] = 0,
\]

(B3)

and it follows that

\[
\lim_{\Phi \to \infty} \left[ (1 - m)^\kappa - (1 - y_0)^{\kappa - 1} - \Phi^{-\kappa} \right] = 0.
\]

(B4)

Since \( \Phi^{-\kappa} \to 0 \) for \( \Phi \to \infty \), we obtain

\[
(1 - m)^\kappa = (1 - y_0)^{\kappa - 1}.
\]

(B5)

Solving for \( m \) yields

\[
m = (1 - y_0)^{1 - \frac{1}{\kappa}}.
\]

(B6)
Acknowledgements. The Center for Climate Systems Modeling (C2SM) at ETH Zurich is acknowledged for providing technical and scientific support. We acknowledge partial support from the ETH Research Grant CH2-01 11-1, EU FP7 EMBRACE and the ERC DROUGHT-HEAT Project.

Edited by: R. Moussa

References


P. Greve et al.: Two-parameter Budyko curve


