

Article

# Impact of Forest Cover and Aridity on the Interplay between Effective Rooting Depth and Annual Runoff in South-West Western Australia

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**Abstract:** Variants of the Budkyo hydrological model describe the effects of land use change on annual water yield. A recent modification using a simple process-based ecohydrological model provides insight into the partitioning of rainfall between runoff and evapotranspiration. In particular, the ‘effective vegetation rooting depth’ becomes the single free parameter in the model and can be related to land use and climate. We applied this approach to investigate the relations between mean annual runoff from 2000 to 2011, catchment average effective rooting depth and the proportion of forest cover across eleven catchments in South-west Western Australia. The proportion of forested and cleared land was partitioned using MODIS minimum annual average LAI values from 1 km<sup>2</sup> pixels over 2000–2011, with forest clearing ranging from 1% to 98% across the 11 catchments. Estimated mean annual runoff using catchment averaged effective rooting depths for forest and cleared (grassland) land obtained using an independent physiologically-based model gave better estimates than a widely used ‘default’ Budkyo-based model. If effective rooting depth declines with aridity, as described by the model then runoff decline may be considerably less (about 50%) than predicted if the effective root depth remains unchanged (70%–92% decline). This highlights the importance of understanding ecohydrological feedbacks between vegetation and climate in projecting scenarios of water yield response to climate change.

**Keywords:** ecohydrology; land use; climate change; effective root depth; Budkyo curve

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## 1. Introduction

Understanding the impacts of land use and climate change on the hydrologic cycle is of fundamental importance to catchment managers interested in optimizing trade-offs between water yield and ecosystem services [1].

At catchment scale there are strong feedbacks and co-evolution between biological and hydrological systems [2,3]. These linked feedbacks between climate, vegetation and soil are particularly pronounced in water-controlled dryland ecosystems [4,5] and in groundwater-dependent ecosystems [6].

Storage and transmission of water through the unsaturated zone exerts a strong control on the functioning of ecosystems and is highly dependent on climate, vegetation and soil properties. At large spatial scales ( $>1000 \text{ km}^2$ ) the long term annual average water balance can be described using only macro-climatic variables and this tends to dominate over vegetative and soil effects on the water balance [7].

At intermediate spatial scales ( $\approx 1000 \text{ km}^2$ ) when the dominant limitation to plant growth is water, the leaf area of native vegetation is tightly coupled to moisture availability and adjusts in a dynamic fluctuation with it [8]. This suggests a ‘steady-state’ or optimal condition in undisturbed ecosystems supporting perennial vegetation [9,10]. In effect, pristine vegetation reflects an integrated response to the processes that affect the availability of water [7].

Understanding the inter-relation between the water balance and biological processes is one of the central concerns of ecohydrology and can provide essential information to improve vegetation management for both water and environmental benefits. The Budkyo hydrological model [11] has been used to evaluate these feedbacks and estimate the effects of land use and climate change on actual annual evaporation and runoff [12–15]. It is essentially a ‘steady state’ model that can be used to determine the difference between two steady states under differing land use.

Until recently, the Budyko model has been used primarily to investigate the gross effects of climate and catchment properties on catchment hydrology, rather than on how changes to biotic factors may influence evapotranspiration and runoff. However, Donohue *et al.* [16] built on the work of Porporato *et al.* [17] to directly include the effects of plant-available soil water holding capacity, mean storm depth and effective rooting depth in the Budkyo framework.

In this paper we commence by overviewing the approach of Donohue *et al.* [16] and then apply it to obtain direct estimates of effective rooting depth from mean annual rainfall and runoff data across 11 catchments in South-west Western Australia that have been subject to varying degrees of forest clearing. We compare these estimates with values obtained from an independent physiologically-based model [18] in order to examine how the effective rooting depth of forest and pasture may be influenced by climate. We then partition out the area of cleared land in each catchment using MODIS-LAI data and use catchment average effective rooting depths from the physiologically-based model to obtain an un-calibrated estimate of mean annual runoff averaged over 2000–2011 for each catchment.

## 2. The Catchment Water Balance within the Budkyo Framework

On an annual timescale or longer, the mean annual flow from a catchment is often estimated using a simple water balance approach [19]. Assuming there are no water abstractions or transfers and that net deep drainage is negligible, the water balance may be expressed as:

$$dS_w/dt = P - E - R \quad (1)$$

where  $S_w$  is soil water,  $P$  is mean annual rainfall,  $E$  is mean actual evapotranspiration and  $R$  is runoff. All terms are expressed as units of depth. If the timescale of interest is greater than the fluctuations in  $S_w$  then the catchment is said to be at ‘steady state’ (left hand side of Equation (1) is 0) and evapotranspiration is given by:

$$E = P - R \quad (2)$$

Equation (2) is the foundation for many long term gauged catchment studies of land use effects on evapotranspiration.

Simple relations between  $E$  and climate have been proposed for steady-state conditions [20], with two general climatic constraints: Under very dry conditions, potential evapotranspiration ( $E_o$ ) exceeds precipitation and  $E$  equals precipitation. At the other extreme, under very wet conditions  $E_o$  is limited by available energy. Annual precipitation exceeds  $E_o$  and  $E$  equals  $E_o$ .

Using these relations, Budyko [11,20] estimated the mean annual evapotranspiration ( $E$ ) for large ( $>1000 \text{ km}^2$ ) catchments and regional scales using:

$$E = P \sqrt{\frac{E_o}{P}} \tanh\left(\frac{P}{E_o}\right) \left[1 - \exp\left(-\frac{E_o}{P}\right)\right] \quad (3)$$

Equation (3) is widely referred to as the ‘Budkyo curve’.

In order to avoid problems with definition of ‘potential evaporation’ [21], Donohue *et al.* [7] preferred the definition of ‘available energy’ as  $\frac{\bar{R}_n}{\lambda}$ , the average net radiation ( $\bar{R}_n$ ) given in  $\text{Js}^{-1}$  and the latent heat of vaporisation,  $\lambda$ , ( $\text{Jkg}^{-1}$ ). Averaged over a year or longer, the net input from sensible heat can be neglected, so ( $\bar{R}_n$ ) is a good approximation of available energy and can be obtained as a combined observation and model based product from satellite data.

Earlier, Choudhury [22] also used net radiation when fitting a Budyko-type curve to field data, but more recently it has become common to use potential evaporation,  $E_o$  [16,23]. The resulting empirical equation is:

$$E = \frac{PE_o}{(P^n + E_o^n)^{1/n}} \quad (4)$$

where  $n$  is a fitting parameter that represents how any processes other than  $P$  and  $E_o$  (and associated errors) affect runoff. [15,23]. Because runoff is the difference between  $P$  and  $E$ , Equation (4) can be combined with Equation (2) and rearranged to give:

$$R = P - \frac{PE_o}{(P^n + E_o^n)^{1/n}} \quad (5)$$

Equation (4) matches the Budkyo curve when  $n = 1.9$  for  $E_o/P = 1.0$  [24] and this is generally referred to as the Choudhury ‘default’ model. Zhang *et al.* [15] used regression analysis to develop a similar ‘default’ model of mean annual runoff for forest or grass cover under steady-state conditions.

Donohue *et al.* [7] point out that the Budyko framework is most applicable to large ( $>10,000 \text{ km}^2$ ) catchments on an annual time step. They argued for the inclusion of vegetation dynamics into the framework in order for it to be extended to smaller catchments and timescales and also for use in operational activities relating to vegetation and water management.

Yang *et al.* [25] used the Budyko framework to estimate the regional water balance using data from 99 catchments in the non-humid region of China. They found that the regional long term water balance forms a group of Budyko curves due to interactions between vegetation, climate and the water cycle.

The difference in the Budyko curve groups reported by Yang *et al.* [25] is captured in the ‘ $n$ ’ parameter of Equation (4). The physical meaning of this parameter remained comparatively obscure until Donohue *et al.* [23] proposed an ecohydrological framework that included the effects of plant available soil water holding capacity, mean storm depth and effective rooting depth.

#### *The Ecohydrological Basis of the Choudhury ‘ $n$ ’ Parameter*

Porporato *et al.* [17] showed that the terrestrial water balance is governed by the ratio between the soil storage capacity and the mean rainfall input per event and either the dryness index (the ratio between the maximum evapotranspiration and the mean rainfall rate) or the ratio between the rate of occurrence of rainfall events and the maximum evapotranspiration rate. The first ratio is given by

$$\gamma = \frac{\kappa Z_e}{\alpha} \quad (6)$$

where the dimensionless ratio,  $\gamma$ , is a function of ‘effective’ rooting depth,  $Z_e$  (mm), the fractional plant available water holding capacity,  $\kappa$ , (0–1 dimensionless) and the mean depth per storm event,  $\alpha$  (mm).

Donohue *et al.* [16] then developed an approximate empirical linear relation between  $\gamma$  and the Choudhury ‘ $n$ ’ parameter, which in combination with Equation (6) is expressed as:

$$n = \frac{0.21\kappa Z_e}{\alpha} + 0.6 \quad (7)$$

If values for  $\alpha$  can be estimated from pluviograph data and  $\kappa$  obtained from soil databases then the average annual  $Z_e$  can be obtained by fitting  $n$  to mean annual runoff data using Equation (5) and inverting Equation (7) to give

$$Z_e = \frac{\alpha(n-0.6)}{0.21\kappa} \quad (8)$$

An alternative ‘forward’ (uncalibrated) approach to estimating effective rooting depth was given by Guswa [18]. Ignoring groundwater-dependency, the Guswa model is given by

$$Z = \frac{\alpha}{\kappa(1-W)} \ln X \quad (9)$$

where  $W$  is given by Donohue *et al.* [16] as  $\left(\frac{E_o}{P}\right)^{-1}$ . In the definition of  $W$ , Guswa [18] refers to potential transpiration, not  $E_o$  but since the Penman  $E_o$  essentially defines the atmospheric demand for water we follow the notation of Donohue *et al.* [16].

For  $W < 1$ ,  $X$  is obtained from

$$X = W \left( 1 + \frac{\kappa (1-W)^2}{\alpha \frac{1}{2A}} \right) + \sqrt{\frac{\kappa (1-W)^2}{\alpha \frac{1}{2A}} + \left( \frac{\kappa (1-W)^2}{\alpha \frac{1}{2A}} \right)^2} \quad (10)$$

If  $W \geq 1$ , the square root term in Equation (10) is subtracted, rather than added. The parameter  $A$  ( $\text{mm}^{-1}$ ) represents the physiological cost-benefit of additional deeper roots for a given vegetation type [18] and is defined as

$$A = \frac{\gamma_r D_r}{L_r W_{ph} T_p f_s} \quad (11)$$

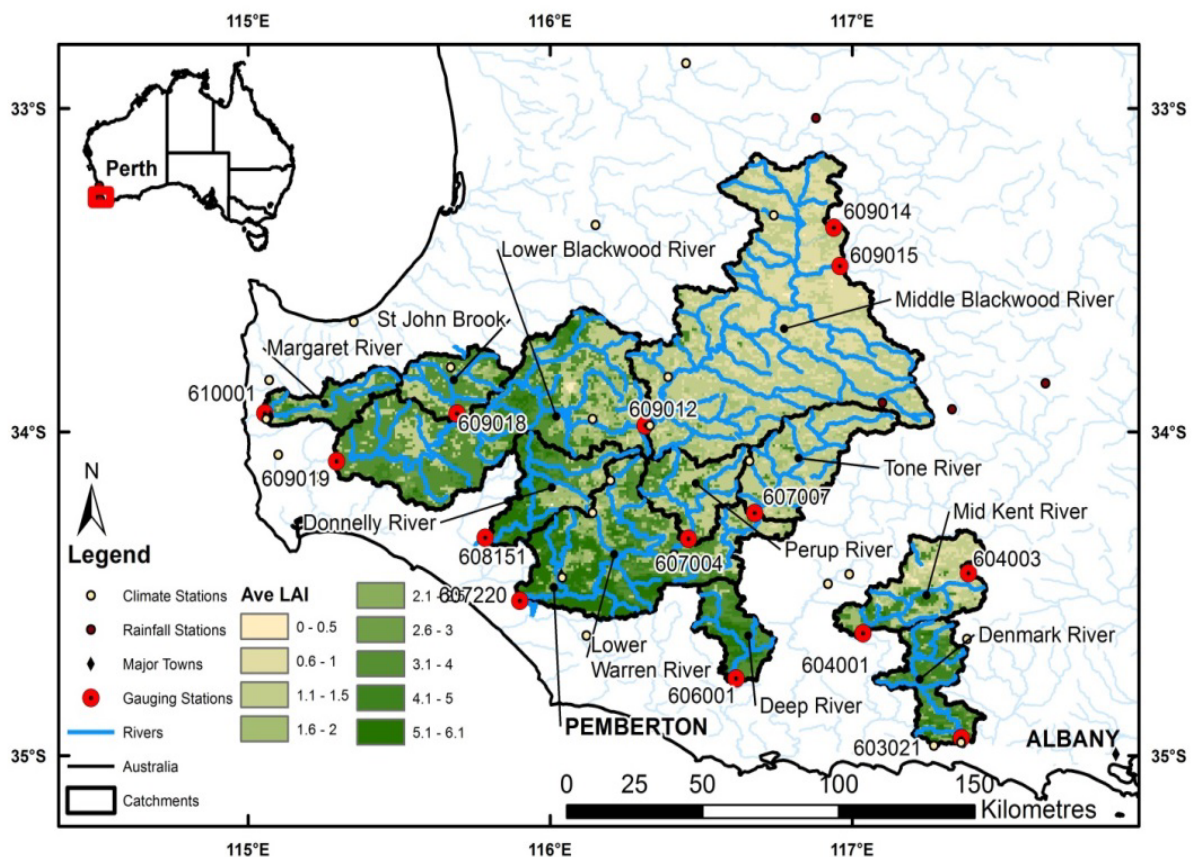
where  $\gamma_r$  is the root respiration rate ( $\text{mmol CO}_2 \text{ g}^{-1} \text{ d}^{-1}$ ),  $D_r$  is the root length density ( $\text{cm cm}^{-3}$ )  $L_r$  is the specific root length ( $\text{cm g}^{-1}$ ),  $W_{ph}$  is the water use efficiency of photosynthesis ( $\text{mmol CO}_2 \text{ cm}^{-3} \text{ H}_2\text{O}$ ),  $T_p$  is the daily potential evapotranspiration rate ( $\text{mm d}^{-1}$ ) and  $f_s$  is the growing season length (fraction of year).

Note that in Equation (9)  $Z$  is clearly related to  $W$  (the reciprocal of the climate index), modified by  $\kappa$ ,  $\alpha$ , and  $A$ . Since  $A$  generally depends upon regional parameters for forest and grassland, the parameters  $\kappa$  and  $\alpha$  account for variations in  $Z$  for a given  $W$ .

### 3. Materials and Methods

The study was undertaken on eleven gauged catchments with areas exceeding  $440 \text{ km}^2$  in South-west Western Australia, spanning the 540–1200 mm rainfall zone (Figure 1).

**Figure 1.** Location of study catchments in South-west Western Australia.



Study catchment characteristics are summarized in Table 1. Percent clearing (Table 1) was calculated using mean annual minimum Leaf Area Index (LAI) on a 1 km<sup>2</sup> grid derived from the improved (2011) MOD16 Global Terrestrial Data Set [26,27]. Average annual minimum LAI was calculated by taking the minimum LAI value in each water year (April to March) and averaging across the period April 2000 to March 2011. Use of minimum annual LAI gives a clear distinction between annual (predominantly grassland) and perennial (forest) vegetation, as illustrated in Figure 2. Note that in this region cleared land is mostly converted to grassland pasture in the >600 mm rainfall zone.

**Table 1.** Characteristics of the study catchments.

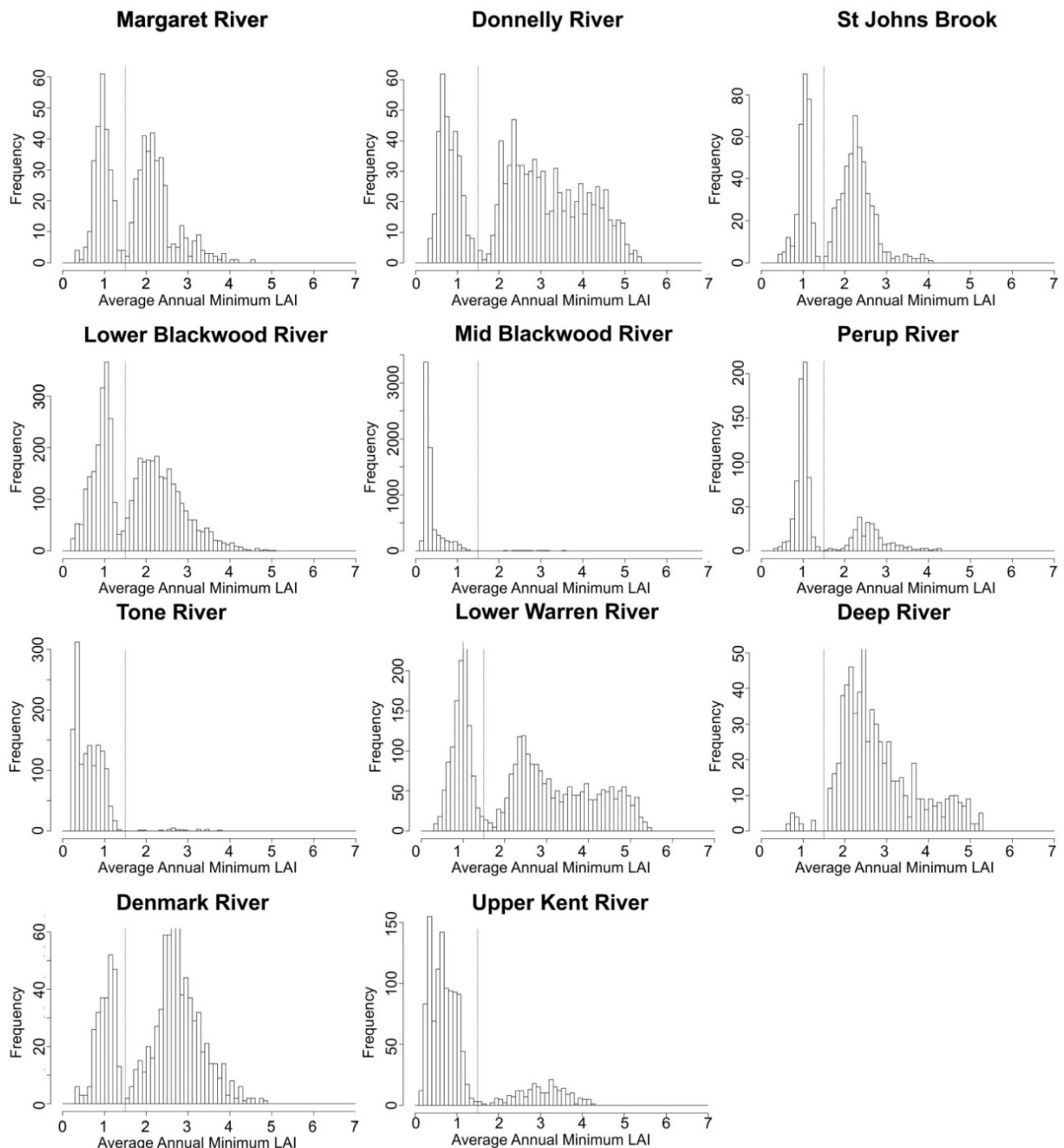
Study Catchment	Area (km <sup>2</sup> )	Percent Forest	Mean Annual Rain (mm) (2000–2011)	Mean Annual Runoff (mm) (2000–2011)	Penman E <sub>o</sub> (mm)
1 Margaret River	443	58.0	790	114.5	1575
2 Donnelly River	787	70.0	862	77.6	1610
3 St Johns Brook	575	62.0	800	44.9	1575
4 Lower Blackwood River	3066	58.0	730	45.0	1591
5 Middle Blackwood River	5043	01.0	535	31.7	1610
6 Perup River	666	29.0	628	09.0	1591
7 Tone River	1015	02.0	593	32.5	1610
8 Lower Warren River	2187	64.0	867	54.7	1591
9 Deep River	467	98.0	994	51.2	1511
10 Denmark River	671	72.0	891	50.1	1526
11 Mid Kent River	880	18.0	591	17.8	1610

For three of the catchments (Warren River, Blackwood River and Kent River), we excised reach data (Figure 1). By using three gauging stations on the Warren River (two cleared upstream tributaries (West Australian Department of Water (DoW) Stations: 607004 and 607007) and one downstream (DoW Station: 607220)), we calculated mean annual runoff (mm) from available flow data [28] for the more heavily forested downstream area only. Similarly, for the Blackwood River we deducted the mid-catchment flow (DoW Station: 609012) and St John's Brook (DoW Station: 609018) from the downstream gauge at (DoW Station: 609019) to isolate the more heavily forested lower section of this catchment. The mid Kent River reach was also excised between (Dow Station: 604001) and (Dow Station: 604003). This provided a dataset of eleven discharge records across a range of land clearing from 2% to 99%.

Annual rainfall data for open stations in this region were converted into a spatial dataset via regularised splines, produced at the same 1 km<sup>2</sup> grid as the MODIS LAI product. Locations of climate and rainfall stations within the study region are shown on Figure 1.

For each catchment the average annual rainfall over the study period 2000–2011 was analysed by sampling and then “binning” and averaging of point raster values using automated Python scripts utilizing the ArcGIS10 GeoProcessor Toolbox. Penman potential ET was averaged for each catchment using data from Donohue *et al.* [16]. We chose to average annual (water year runoff) from January 2000–November 2010 as it coincided with the available MODIS record and because there was no significant decline in rainfall across the study catchments over this period [29].

**Figure 2.** Separation of land use in each catchment into ‘forest’ (LAI > 1.5) and ‘grassland’ (LAI < 1.5) using MODIS minimum annual average LAI values (2000–2011).



#### 4. Results and Discussion

The optimal value of Choudhury's  $n$  parameter was estimated for each of the eleven runoff records using Equation (5). Catchment average  $Z_e$  values were then calculated from Equation (8), with  $\kappa$  (range 0.10–0.13) estimated from regional soil hydrology groups documented by Moore *et al.* [30] and  $\alpha$  (range 4.5–8.0) from [16].

For this region, the average catchment value of  $Z_e$  is influenced by both the fraction of uncleared land and the aridity. In Table 2, catchments are ranked in order of percent forest cover (highest to lowest) and then average  $Z_e$  values and climate indices are presented for each catchment, together with the  $n$ ,  $\kappa$  and  $\alpha$  values used in the calculation of  $Z_e$ .

**Table 2.** Physical and model parameters for the study catchments.

Study Catchment	Percent Forest	Climate Index, $\phi$ (E <sub>o</sub> /P)	$n^1$	$\kappa$ (cm <sup>3</sup> /cm <sup>-3</sup> )	$\alpha$ (mm)	Average $Z_e^2$ (mm)
Deep River	98.0	1.52	3.68	0.13	6.0	676
Denmark River	72.0	1.71	3.10	0.12	6.0	595
Donnelly River	70.0	1.87	2.28	0.12	7.8	520
Lower Warren River	64.0	1.83	2.71	0.13	6.2	479
St Johns Brook	62.0	1.97	2.65	0.12	6.0	489
Lower Blackwood River	58.0	2.18	2.35	0.12	8.0	556
Margaret River	58.0	2.00	1.71	0.10	7.5	396
Perup River	29.0	2.53	3.25	0.12	5.2	547
Mid Kent River	18.0	2.72	2.52	0.13	4.5	316
Tone River	02.0	2.72	2.08	0.12	4.6	270
Middle Blackwood River	01.0	3.00	1.90	0.13	6.0	287

Notes: <sup>1</sup> Equation (5); <sup>2</sup> Equation (8).

It is clear from Table 2 that forest clearing has been greatest in the drier catchments. The three catchments with less than 20% forest cover are in a zone with climate indices between 2.72 and 3.0. The mean  $Z_e$  of 291 mm for these catchments could reasonably be taken as a value for ‘grassland’ in this aridity zone. Conversely, the three catchments with 70% or more forest cover are much less arid and have a mean  $Z_e$  of 597 mm. However, the highest  $Z_e$  value (676 mm) is obtained for the catchment with the least (2%) clearing.

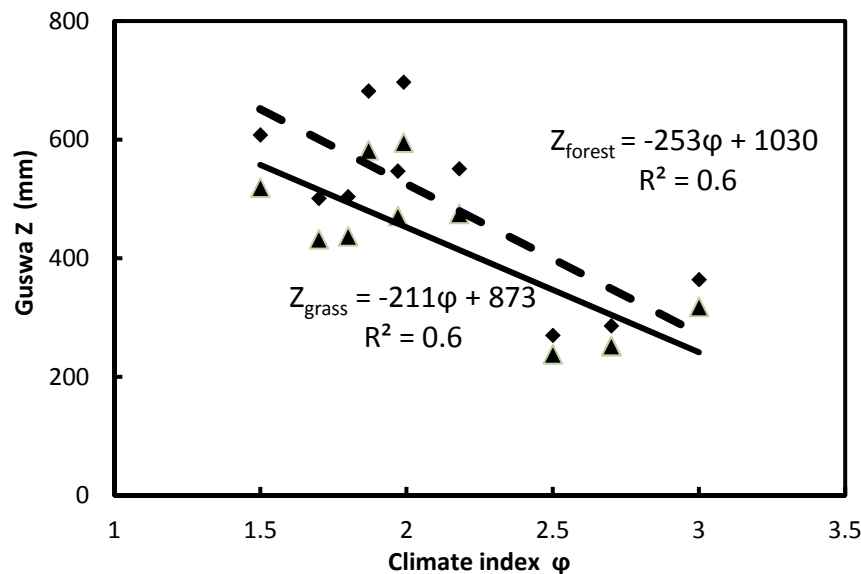
To investigate in more detail the how the effective root depths of grassland and forest might change with increasing aridity we used the physiologically-based model of Guswa [18].

Guswa  $Z$  values were obtained for grassland and forest using Equation (9), with the parameter values required to estimate  $A$  in Equation (11) taken from Donohue *et al.* [16]. The following values were common for grassland and forests:  $\gamma_r$  0.5;  $D_r$  0.1;  $L_r$  1500. Values for  $W_{ph}$  and  $f_s$  were 0.22 and 0.8 for grassland and 0.33 and 1.0 for forest.

Estimates of the Guswa  $Z$  values for forest ( $Z_{\text{forest}}$ ) and grassland ( $Z_{\text{grass}}$ ), across each of the eleven catchments, show a clear trend of declining  $Z$  values with increasing aridity (Figure 3). The average ‘grassland’ Guswa  $Z$  value for the three driest catchments is 269 mm, with is remarkably close to the mean  $Z_e$  value of 291 mm. Using the regression relations in Figure 3 we calculate that at  $\phi$  of 1.5, the Guswa  $Z$  for forest is 650 mm, which again compares favourably to the  $Z_e$  value for Deep river in Table 2 (676 mm). If we consider a ‘mix’ of 70% forest and 30% grassland for catchments with an aridity index of 1.8, then from the regression relations in Figure 3 we obtain a catchment average Guswa  $Z$  of  $(0.7 \times 575) + (0.3 \times 493) = 550$  mm. Again, this compares well with the average  $Z_e$  from Table 2 for the Denmark, Donnelly and Lower Warren rivers (531 mm).



**Figure 3.**  $Z_{\text{forest}}$  (◆) and  $Z_{\text{grass}}$  (▲) calculated from the Guswa model for each catchment plotted against the climate index ( $\phi = E_o/P$ ).



The decline in  $Z_{\text{forest}}$  with increasing aridity shown in Figure 3 also has implications for forest response to forecasted increased aridity across this region in the next 50 years. It is also evident from Figure 3 that the difference between Guswa  $Z$  values for forest and grassland diminishes with increasing aridity. At a climate index of about 1.5–2.0 the difference is 80 to 100 mm but from climate indices of 2.5–3.0 the difference has diminished to 30 to 35 mm.

#### 4.1. Estimation of Runoff Using Guswa-Based and Choudhury ‘Default’ Models

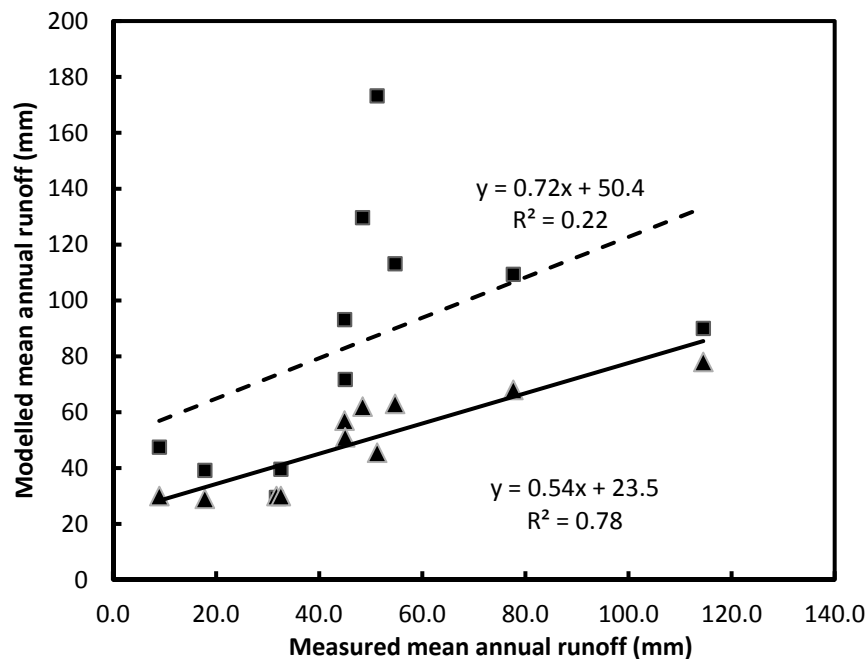
Catchment average Guswa  $Z$  ( $Z_{\text{ave}}$ ) values can be obtained from

$$Z_{\text{ave}} = (Z_{\text{forest}}F_{\text{forest}} + Z_{\text{grass}}F_{\text{grass}}) \quad (12)$$

where  $F_{\text{grass}}$  and  $F_{\text{forest}}$  are the fractions of the catchment under grass and forest.

The resulting average Guswa  $Z$  values can then be used to independently estimate the Choudhury ‘ $n$ ’ parameter for each catchment from Equation (7) and provide an estimate of catchment runoff without any calibration using Equation (5). This approach is shown in Figure 4 and is compared to the Choudhury ‘default’ model ( $n = 1.9$ ). The runoff estimates that include an average rooting depth from the Guswa model are generally much closer to the observed mean annual runoff (2000–2011) than the Choudhury ‘default’ model. Note that catchments with less than 40 mm measured mean annual runoff are predominantly cleared (<30% forest cover) but have the most arid climate indices. For catchments with 40 mm to 80 mm measured mean annual runoff the forest cover ranges from 58% to 90% and the estimation of Guswa  $Z_{\text{forest}}$  to give catchment  $Z_{\text{ave}}$  values (Equation (12)) results in ‘ $n$ ’ values that more accurately describe the runoff than the ‘default’ case. Root Mean Square Error is reduced from 21.0 (Choudhury ‘default’ model) to 7.5 (Guswa-based model). Bias is also reduced from 563.0 to 58.9. Weighted by catchment area over all eleven catchments the Guswa-based model overestimates regional mean annual runoff by 12% whereas the Choudhury ‘default’ model overestimates by 82%.

**Figure 4.** Comparison of runoff estimated from the Guswa-based model (▲) and Choudhury ‘default’ model (■) for eleven catchments in south-west Western Australia.



#### 4.2. Implications for Climate Change

Within 400 km of the coast of south-west Western Australia the temperature is projected to increase in the range of 0.4 to 5.4 °C by 2080 and rainfall is projected to decrease by 0% to 80% [31], with a likely doubling of drought frequency [32]. Increasing temperature should lead to an increase in potential evapotranspiration but this has been countered by a decrease in global windspeed [33].

If there is no feedback between effective rooting depth and climate (and assuming  $\kappa$  and  $\alpha$  remain constant), then for the model used here, the only effect on runoff will be rainfall decline and the ‘ $n$ ’ parameter in Equation (5) remains unchanged. However, if effective rooting depth decreases with increasing aridity then ‘ $n$ ’ will be altered (Equation (7)).

As an example, if we consider a 40% decline in rainfall over the forested Deep river catchment then holding ‘ $n$ ’ constant leads to a decline in runoff from 51.2 mm (2000–2011 annual mean) to 5.2 mm. However, noting that a 40% decline in rainfall increases the aridity index to 2.53 allows ‘ $n$ ’ to be adjusted in equation 7 by altering  $Z_e$  for the new aridity (climate index of 2.53) using the regression relation in Figure 3. The adjusted model  $Z_e$  is now 390 mm, giving an ‘ $n$ ’ value of 2.37 and a reduction in the projected runoff decline to 25.8 mm (50% decline as opposed to 90% decline without feedback).

A decline in effective rooting depth with increasing aridity was demonstrated by Donohue *et al.* [16] for the Murray-Darling Basin in eastern Australia. If this climate-dependency follows the Guswa model projections then runoff decline is mitigated considerably when compared to the constant effective root depth case.

This simple example does not of course factor in any possible changes to the occurrence-intensity distribution of rainfall but illustrates the need for greater consideration of vegetation feedbacks when considering future runoff scenarios that may result from climate change.

## 5. Conclusions

We have shown that Choudhury's 'default' model with  $n = 1.9$  requires modification in order to describe mean annual runoff from catchments in south-west Western Australia. By using Donohue *et al.*'s [16] approach of incorporating three process-based ecohydrologic variables (effective rooting depth, mean storm depth and soil water storage capacity) into the description of  $n$  we were able to improve runoff prediction and gain insight into the effect of land use on runoff generation. The average effective rooting depth in a catchment was strongly correlated with the fraction of uncleared land identified using MODIS minimum annual average LAI values (2000–2011). Using the Guswa model to obtain values of  $Z$  for forest and grasslands that varied with catchment aridity gave an entirely uncalibrated model of mean annual runoff that overestimated the regional mean annual runoff by only 12%.

If effective rooting depth declines with increasing aridity then ' $n$ ' will decrease and water yields will be higher than if ' $n$ ' remains constant. However, ' $n$ ' is also affected by mean storm depth and so an understanding of how this is likely to change with declining rainfall is also important for improved understanding of how climate change could impact on water yields.

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## Author Contributions

Both authors contributed to the conception and development of this manuscript.

## Conflicts of Interest

The authors declare no conflict of interest.

## References

1. Townsend, P.V.; Harper, R.J.; Brennan, P.D.; Dean, C.; Wu, S.; Smettem, K.R.J.; Cook, S. Multiple environmental services as an opportunity for watershed restoration. *For. Policy Econ.* **2012**, *17*, 45–58.
2. Montaldo, N.; Rondena, R.; Albertson, J.D.; Mancini, M. Parsimonious modelling of vegetation dynamics for ecohydrologic studies of water-limited ecosystems. *Water Resour. Res.* **2005**, *41*, doi:10.1029/2005WR.
3. Berry, S.L.; Farquhar, G.D.; Roderick, M.L. Co-evolution of climate, soil and vegetation. In *Encyclopaedia of Hydrological Sciences*; Anderson, M., Ed.; John Wiley and Sons: Indianapolis, IN, USA, 2005.
4. Rodriguez-Iturbe, I.; Porporato, A.; Laio, F.; Ridolfi, L. Plants in water-controlled ecosystems: Active role in response to hydrologic processes and response to water stress—I. Scope and general outline. *Adv. Water Resour.* **2001**, *24*, 695–705.

5. Rodriguez-Iturbe, I.; Porporato, A. *Ecohydrology of Water-Controlled Ecosystems: Soil Moisture and Plant Dynamics*; Cambridge University Press: Cambridge, UK, 2004.
6. Eamus, D.; Froend, R.; Loomes, R.; Hose, G.; Murray, B. A functional methodology for determining the groundwater regime needed to maintain the health of groundwater-dependent vegetation. *Austr. J. Bot.* **2006**, *54*, 97–114.
7. Donohue, R.J.; Roderick, M.L.; McVicar, T.R. On the importance of including vegetation dynamics in Budyko's hydrological model. *Hydrol. Earth Syst. Sci.* **2007**, *11*, 983–995.
8. Nemani, R.R.; Running, S.W. Testing a theoretical climate soil leaf-area hydrologic equilibrium of forests using satellite data and ecosystem simulation. *Agric. For. Meteorol.* **1989**, *44*, 245–260.
9. Eagleson, P.S. Ecological optimality in water-limited natural soil vegetation systems 1, theory and hypothesis. *Water Resour. Res.* **1982**, *18*, 325–340.
10. Eagleson, P.S. *Ecohydrology: Darwinian Expression of Vegetation Form and Function*; Cambridge University Press: New York, NY, USA, 2002.
11. Budyko, M.I. *Climate and Life. International Geophysics Series*; Academic: New York, NY, USA, 1974; Volume 18, p. 508.
12. Gerrits, A.M.J.; Savenije, H.H.J.; Veling, E.J.M.; Pfister, L. Analytical derivation of the Budyko curve based on rainfall characteristics and a simple evaporation model. *Water Resour. Res.* **2009**, *45*, W04403.
13. Milly, P.C.D. Climate, soil water storage, and the average annual water balance. *Water Resour. Res.* **1994**, *30*, 2143–2156.
14. Roderick, M.L.; Farquhar, G.D. A simple framework for relating variations in runoff to variations in climatic conditions and catchment properties. *Water Resour. Res.* **2011**, *47*, W00G07.
15. Zhang, L.; Hickel, K.; Dawes, W.R.; Chiew, F.H.S.; Western, A.W.; Briggs, P.R. A rational function approach for estimating mean annual evapotranspiration. *Water Resour. Res.* **2004**, *40*, W02502.
16. Donohue, R.J.; Roderick, M.L.; McVicar, T.R. Roots, storms and soil pores: Incorporating key ecohydrological processes into Budyko's hydrological model. *J. Hydrol.* **2012**, *436–437*, 35–50.
17. Porporato, A.; Daly, E.; Rodriguez-Iturbe, I. Soil water balance and ecosystem response to climate change. *Am. Natur.* **2004**, *164*, 625–632.
18. Guswa, A.J. The influence of climate on root depth: A carbon cost-benefit analysis. *Water Resour. Res.* **2008**, *44*, W04403.
19. Bari, M.A.; Smettem, K.R.J.; Sivapalan, M. Understanding changes in annual runoff following land use changes: A systematic data-based approach. *Hydrol. Process.* **2005**, *19*, 2463–2479.
20. Budyko, M.I. *The Heat Balance of the Earth's Surface*; U.S. Department of Commerce: Washington, WA, USA, 1958.
21. Granger, R.J. An examination of the concept of 'potential evaporation'. *J. Hydrol.* **1989**, *111*, 9–19.
22. Choudhury, B.J. Evaluation of an empirical equation for annual evaporation using field observations and results from a biophysical model. *J. Hydrol.* **1989**, *216*, 99–100.
23. Yang, H.; Yang, D.; Lei, Z.; Sun, F. New analytical derivation of the mean annual water-energy balance equation. *Water Resour. Res.* **2008**, *44*, W04426.

24. Donohue, R.L.; Roderick, M.L.; McVicar, T.R. Assessing the differences in sensitivities of runoff to changes in climatic conditions across a large basin. *J. Hydrol.* **2011**, *406*, 234–244.
25. Yang, D.; Shao, W.; Yeh, P.J.-F.; Yang, H.; Kanae, S.; Oki, T. Impact of vegetation coverage on regional water balance in the nonhumid regions of China. *Water Resour. Res.* **2009**, *45*, W02502.
26. Mu, Q.; Heinsch, F.A.; Zhao, M.; Running, S.W. Development of a global evapotranspiration algorithm based on MODIS and global meteorology data. *Remote Sens. Environ.* **2007**, *111*, 519–536.
27. Mu, Q.; Zhao, M.; Running, S.W. Improvements to a MODIS global terrestrial evapotranspiration algorithm. *Remote Sens. Environ.* **2011**, *115*, 1781–1800.
28. River Monitoring Stations. Department of Water, Government of Western Australia. Available online: <http://kumina.water.wa.gov.au/waterinformation/telem/stage.cfm> (accessed on 18 January 2013).
29. Smettem, K.R.J.; Waring, R.H.; Callow, N.; Wilson, M.; Mu, Q. Satellite-derived estimates of forest leaf area index in South-West Western Australia are not tightly coupled to inter-annual variations in rainfall: Implications for groundwater decline in a drying climate. *Glob. Change Biol.* **2013**, *19*, 2401–2412.
30. Moore, G.; Hall, D.; Russell, J. Soil Water. In *Soilguide. A Handbook for Understanding and Managing Agricultural Soils*; Moore, G., Ed.; Agriculture Western Australia Bulletin No. 4343; Sands Print Group: Perth, Australia, 1998; pp. 80–93.
31. Hennessy, K.J.; Fitzharris, B.; Bates, B.C.; Harvey, N.; Howden, M.; Hughes, L.; Salinger, J.; Warrick, R. Australia and New Zealand: Climate change 2007, impacts, adaptation and vulnerability: Contribution of Working Group II to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change. In *IPCC (2007) Climate Change*; Cambridge University Press: Cambridge, UK, 2007; Volume 4, Chapter 11, pp. 507–540.
32. Kirono, D.G.C.; Kent, D.M.; Hennessy, K.J.; Mpelasoka, F. Characteristics of Australian droughts under enhanced greenhouse conditions: Results from 14 global climate models. *J. Arid Environ.* **2011**, *75*, 566–575.
33. McVicar, T.R.; Roderick, M.L.; Donohue, R.J.; van Neil, T.G. Less bluster ahead? Ecohydrological implications of global trends of terrestrial near-surface wind speeds. *Ecohydrology* **2012**, *5*, 381–388.