



1 **The influence of water table depth on evapotranspiration in the Amazon arc of deforestation**

2 Authors: John O'Connor<sup>1</sup>, Maria J. Santos<sup>2</sup>, Karin T. Rebel<sup>1</sup> and Stefan C. Dekker<sup>1,3</sup>

3 <sup>1</sup>Copernicus Institute of Sustainable Development, Department Environmental Sciences, Utrecht University, The  
4 Netherlands

5 <sup>2</sup>University Research Priority Program in Global Change and Biodiversity and Department of Geography,  
6 University of Zürich, Switzerland

7 <sup>3</sup>Faculty of Management, Science and Technology, Open University, Heerlen, The Netherlands

8

9 *Correspondence to:* John C. O Connor (j.c.oconnor@uu.nl)

10

11 Abstract:

12

13 The Amazon rainforest evapotranspiration (ET) flux provides climate regulating and moisture provisioning  
14 ecosystem services through a moisture recycling system. The dense complex canopy and deep root system  
15 creates an optimum structure to provide large ET fluxes to the atmosphere forming the source for precipitation.  
16 Extensive land use and land cover change (LULCC) from forest to agriculture in the arc of deforestation breaks  
17 this moisture recycling system. Crops such as soybean are planted in large homogeneous monocultures and the  
18 maximum rooting depth of these crops is far shallower than forest. This difference in rooting depth is key as  
19 forests can access deep soil moisture and show no signs of water stress during the dry season while in contrast  
20 crops are highly seasonal with a growing season dependant on rainfall. As access to soil moisture is a limiting  
21 factor in vegetation growth, we hypothesised that if crops could access soil moisture they would undergo less  
22 water stress and therefore would have higher evapotranspiration rates than crops which could not access soil  
23 moisture.

24

25 We combined remote sensing data with modelled groundwater table depth (WTD) to assess whether vegetation  
26 in areas with a shallow WTD had higher ET than vegetation in deep WTD areas. We randomly selected areas of  
27 forest, savanna and crop with deep and shallow WTD and examined whether they differ on MODIS  
28 Evapotranspiration (ET), Land Surface Temperature (LST) and Enhanced Vegetation Index (EVI), from 2001 to  
29 2012, annually and during transition periods between the wet and dry season. As expected, we found no  
30 differences in ET, LST, and EVI for forest vegetation between deep and shallow WTD, which because of their  
31 deep roots could access water and maintain evapotranspiration for moisture recycling during the entire year. We  
32 found significantly higher ET and lower LST in shallow WTD crop areas than in deep WTD during the dry  
33 season transition, suggesting that crops in deep WTD undergo higher water stress than crops in shallow WTD  
34 areas.

35

36 The differences found between crop in deep and shallow WTD, however, are of low significance with regards  
37 the moisture recycling system as the difference resulting from conversion of forest to crop has an overwhelming  
38 influence (ET in forest is  $\approx 2 \text{ mm day}^{-1}$  higher than that in crops) and has the strongest impact on energy balance  
39 and ET. However, access to water during the transition between wet and dry seasons may positively influence  
40 growing season length in crop areas.

41

42



43 **Introduction**

44

45 The Amazon rainforest has been reduced to 80% of its original size due to deforestation over the past few decades  
46 (Davidson et al., 2012). Land use and land cover change (LULCC) from forest to agricultural land disrupts the  
47 Amazonian water cycle due to changes in evapotranspiration, infiltration, and runoff (Fearnside, 1997; Lawrence  
48 and Vandecar, 2014). Changes in evapotranspiration result in major changes to the water energy balance, as forest  
49 vegetation has high evapotranspiration rates and is replaced with agricultural vegetation with lower  
50 evapotranspiration which results in a lower latent heat flux and higher sensible heat flux (Swann et al., 2015). In  
51 addition changes in evapotranspiration reduce the available atmospheric moisture which can reduce rainfall.

52 Differences in vegetation structure are suggested to be the main drivers affecting the evapotranspiration rates.  
53 Forest vegetation has a high total leaf surface area while agricultural vegetation usually have a lower leaf area  
54 (Asner et al., 2003; Costa et al., 2007). This difference lowers the potential surface area for both interception  
55 evaporation and transpiration. In addition, the rooting depth of forest and agricultural vegetation differs greatly  
56 (Costa and Foley, 2000). Forest vegetation have deep roots which facilitate access to deep soil moisture  
57 maintaining their supply of water necessary for photosynthesis even during the dry season. Forests can maintain a  
58 high rate of evapotranspiration during the dry season, unaffected by low rainfall (Maeda et al., 2017; Staal et al.,  
59 2018a). Following LULCC the new vegetation cover may lack deep roots and therefore no longer accesses deeper  
60 soil moisture. However, little is known whether LULCC that occurred in areas with a shallow WTD facilitates  
61 access to water and leads to higher vegetation productivity and evapotranspiration compared to areas with a deep  
62 WTD. Over the past few decades, the developing agricultural industry driven by international demand encouraged  
63 extensive LULCC (Brando et al., 2014; Foley et al., 2007; Sampaio et al., 2007) concentrated along the southern  
64 and eastern edge of the Amazon in an area known as the arc of deforestation (Costa and Pires, 2010; Malhi et al.,  
65 2008). LULCC negatively impacts the ecosystem service provision of the Amazon including highly valuable  
66 services such as carbon storage and sequestration and moisture recycling and regulation. Understanding the effect  
67 that LULCC has on evapotranspiration is important as the loss of evapotranspiration impacts both climate and  
68 precipitation on local and regional scales.

69

70 Local climate can be impacted by LULCC due to changes in the energy balance as loss of evapotranspiration  
71 reduces latent heat and increases sensible heat. Studies in the Amazon have shown that temperatures increase on  
72 average 1.4 °C with a max of 7 °C following conversion to crop (Badger and Dirmeyer, 2015). The seasonal impact  
73 of LULCC is particularly strong during the dry season as crop evapotranspiration is at its lowest, latent heat flux  
74 can be reduced by 78% and the sensible heat flux can increase by 85% relative to forest (Ponte De Souza et al.,  
75 2011). The loss of evapotranspiration impacts rainfall both locally and on the continental scale. Evapotranspiration  
76 returns water to the atmosphere where it can precipitate again either in situ or be carried further downwind (Eltahir  
77 and Bras, 1994). Large forests like the Amazon, because of their density and extent create large evapotranspiration  
78 fluxes, leading to underpressure over land and the pressure differences draw moisture towards land (Makarieva  
79 and Gorshkov, 2007; Sheil, 2014). As high as 70% of rainfall in the Amazon and southern Brazil is a result of  
80 Amazonian evapotranspiration (van der Ent and Savenije, 2011). This evapotranspiration precipitation cycle is  
81 highly important in both maintaining the forest itself but also providing precipitation to non forested areas. LULCC  
82 reduces the evapotranspiration and breaks this moisture recycling system resulting in lower rainfall locally and  
83 downwind. The seasonal loss of evapotranspiration in crop areas during the dry season is of great significance,



84 evidence already suggests that LULCC has resulted in a lengthening of the dry season (Costa and Pires, 2010;  
85 Debortoli et al., 2017). Model simulations predict that if deforestation continues by 2050 the loss of  
86 evapotranspiration will result in a negative effect further reducing forest cover and evapotranspiration (Foley et  
87 al., 2007; Spracklen et al., 2012). The conversion of forest in Brazil is driven by an increasing demand for  
88 agricultural production which has almost doubled since 2000 (Zalles et al., 2019); however, losses in  
89 evapotranspiration could lead to subsequent losses in agricultural productivity as rainfall is reduced and the  
90 growing season is shortened (Oliveira et al., 2013).

91

92 Agricultural vegetation in the arc of deforestation experiences high seasonality during the dry season unseen in  
93 forest vegetation. Forest vegetation provides an optimum structure for evapotranspiration due to its tall complex,  
94 dense canopy and deep root systems which can access deep soil moisture stores and maintain high transpiration  
95 rates even during periods of low rainfall (Nepstad et al., 1994; Sheil, 2014). Agricultural crops, on the other hand,  
96 are known to contribute much less to evapotranspiration as a result of their shorter canopy and simpler structure  
97 (Fearnside, 1997). In addition, agricultural crops lack the deep root systems of forest which are credited for  
98 maintaining evapotranspiration throughout the dry season (Nepstad et al., 1994). In theory, if vegetation continues  
99 to access the water table within the root zone then this vegetation will continue to transpire during periods of  
100 reduced rainfall. Thus access to soil moisture is an important limiting factor for photosynthesis and transpiration.  
101 Shallow water table depths across South America are widely distributed and correspond to an area of  
102 approximately 36% of the Amazon (Fan and Miguez-Macho, 2010). We hypothesize that areas of shallow water  
103 table depth (WTD) allow vegetation to access soil moisture, with both shallow and deep rooted vegetation  
104 potentially facilitating vegetation productivity and higher evapotranspiration when compared to areas of deep  
105 WTD. Experimental manipulation of WTD using sub irrigation systems of soybean demonstrated that shallow  
106 WTD benefitted productivity and increased yield (Kahlow et al., 2005; Mejia et al., 2000). During the wet season,  
107 soybean can reach rates of evapotranspiration similar to that of forest (Costa and Foley, 2000). Some studies have  
108 suggested that the difference in annual ET between forest and agricultural crops is primarily due to access to water  
109 during the dry season (Costa et al., 2007).

110

111 In this study, we investigate if naturally occurring shallow water table depth could increase evapotranspiration  
112 compared to deep water table depth. We use a number of freely available remote sensing products in combination  
113 with modelled WTD to assess whether differences in evapotranspiration occur in areas of deep and shallow WTD,  
114 especially crop vegetation. We expect that the influence of WTD will be greater in the transition periods between  
115 wet and dry seasons as rainfall as a water source is limited. In areas of shallow WTD, the saturated zone is closer  
116 to the root zone of the vegetation. In these locations we, therefore, expect vegetation to be buffered against the  
117 reduction in rainfall during the dry season transition and experience drought conditions later, thus delaying the  
118 effect of the dry season. Similarly, during the wet season transition (WST), we expect that areas of shallow WTD  
119 will have higher productivity as vegetation may access the shallow WTD to supplement their demand when rainfall  
120 is low, therefore growing sooner than areas with deep WTD, effectively shortening the dry season. Finally, we  
121 discuss whether differences found in ET between deep and shallow WTD are important for moisture recycling,  
122 vegetation productivity and what are the implications for future LULCC.



123 **2 Methods**

124 **2.1 Study Area**

125 The study area is located in the southern Amazon region, mostly in the northern region of Mato Grosso and  
126 incorporating the border area with Pará (Figure 1). Mato Grosso is classified into three major biomes with  
127 rainforest in the North, cerrado in the central region and wetlands in the southwest (Kastens et al., 2017;  
128 Lathuillière et al., 2012). The climate has two seasons, the wet season in the austral winter and the dry season in  
129 austral summer, the dry season lasts around 5 months with an annual average rainfall of 2000 mm and annual  
130 average temperatures ranging between 22 - 26 °C (Arvor et al., 2014). This region is well-known as a dynamic  
131 agricultural frontier – the arc of deforestation – with high rates of LULCC, where forest was converted for  
132 extensive agriculture, mostly cattle ranching and soy production (Kastens et al., 2017). Mato Grosso is the leading  
133 producer of agricultural crops such as soybean in Brazil (Gusso et al., 2014). We chose a study area which is  
134 centrally located in the arc of deforestation and has both large areas of primary forest and agricultural frontier  
135 regions, covering an area of approximately 750 km x 750 km. Within the selected study area, the dominant land  
136 cover is forest (73 %), followed by savanna (19 %), with cropland accounting for 3 %.

137  
138  
139



140  
141  
142  
143  
144  
145

**Figure 1: Study area on the arc of deforestation the Amazon, in Northern Mato Grosso. Inlayed image shows MODIS land cover classification map (2001) for the three land cover classes analysed. Forest – Green, Savanna – Beige, Crop – Yellow and Other - Grey. Due to the sinusoidal projection of MODIS satellite data, the study area looks distorted.**



## 146 2.2 Datasets

### 147 2.2.1 Remote sensing data

148 Remote sensing offers excellent tools for monitoring changes in vegetation over large regions as it provides full  
149 geographic coverage, high temporal frequency at spatial scales relevant to most Earth system processes (Chambers  
150 et al., 2007). Here we use three separate products from the Moderate Resolution Imaging Spectrometer (MODIS),  
151 namely MODIS Evapotranspiration (MOD16A2), MODIS Land Surface Temperature (MOD11A2), and MODIS  
152 Enhanced Vegetation Index (MOD13A2), to assess the influence of WTD on evapotranspiration. MODIS remote  
153 sensing products were used as they offer a moderate spatial resolution and a high temporal resolution which is  
154 ideal for examination of seasonal processes. We chose to perform the analysis for the currently available MODIS  
155 land cover archive using data from 2001 to 2012. In addition, this period represents a time with high variability of  
156 precipitation extremes in which the Amazon experienced droughts, floods and could depict the variability the  
157 system experiences (Nobre et al., 2016). Data was downloaded from the NASA data sharing portal  
158 ([earthdata.nasa.gov](http://earthdata.nasa.gov)). Data was rescaled to 1 km resolution, no additional post-processing was conducted.

159  
160 MODIS Evapotranspiration (hereafter ET) data (Mu et al., 2011) provides 8 day accumulated evapotranspiration  
161 at 500 m resolution (rescaled to 1 km). The ET dataset is one of the best available datasets due to its relatively  
162 high spatial and temporal resolution as such it has been widely used to investigate the effect of land use change on  
163 evapotranspiration in the Amazon (Loarie et al., 2011; Neill et al., 2013; Vergopolan and Fisher, 2016). The  
164 baseline algorithm to the MODIS ET product is based on the Penman-Monteith equation, and combines parameters  
165 such as land cover, leaf area index (LAI), Albedo and Fraction of Photosynthetically Active Radiation (FPAR)  
166 directly observed with or modelled from MODIS data, with reanalysis data on Radiation, Air Temperature and  
167 Humidity from the Global Modelling and Assimilation Office (Mu et al., 2011). The MODIS ET products were  
168 previously tested over the Amazon by comparing its outputs with eddy covariance tower data, showing that the  
169 product is more accurate over longer temporal scales and larger areas (Ruhoff et al., 2013). Therefore, the ET  
170 product may not be able to represent expected local differences in ET created by difference in WTD at the scale  
171 we want to answer our research questions, and we chose to analyse additional remote sensing products in which  
172 the differences might be detectable, and potentially show a signal to the effects of WTD on the water cycle.

173 MODIS Land Surface Temperature (hereafter LST) provides an 8 day mean day time land surface temperature in  
174 degrees Kelvin at 1 km resolution. LST data are produced by detection of thermal infrared radiation between 3 –  
175 15  $\mu\text{m}$  spread across 15 bands of the thermal sensor on board the MODIS satellite system and temperatures are  
176 modelled based on land cover classification with a clear sky accuracy of 1 degree K (Wan, 2014). MODIS LST  
177 data was converted to degrees Celsius. Despite low albedo in the Amazon and high net radiation, the strong  
178 evapotranspiration results in a net cooling effect (Bonan, 2008). We expected that areas with lower LST will be  
179 observed in areas of higher ET (Eltahir and Bras, 1994).

180 MODIS Enhanced Vegetation Index (hereafter EVI) provides a 16 day repeat observation on vegetation greenness  
181 at 500 m resolution (rescaled to 1 km). EVI is a vegetation index that measures greenness as a proxy for  
182 productivity (Huete et al., 2002). It was developed to improve upon the normalized difference vegetation index  
183 (NDVI), as it is less sensitive to saturation in highly dense canopies as those in the Amazon, and EVI also corrects  
184 for canopy background effects and atmospheric aerosol effects (Huete et al., 2002). This MODIS product offers  
185 an observation of vegetation productivity as it measures “greenness” and is correlated to



186 photosynthesis/evapotranspiration (Sims et al., 2006). Thus we expect that vegetation with adequate access to  
187 water near their root zone will have a comparatively higher EVI than vegetation which is water stressed. This  
188 higher EVI, in turn, would correspond to areas of higher ET.

189 In addition, we also used the MODIS land cover product for selection of our analysis sites (see below). MODIS  
190 land cover (hereafter land cover) provides a classification of global land cover at 1 km resolution, and it is annually  
191 updated and used as input for other MODIS datasets utilised in this study (Friedl et al., 2010). For this study, we  
192 only used pixels that were classified as the same land cover type during the entire study period 2001 - 2012. The  
193 study area chosen provides a sufficient number of representative pixels for random selection of each land cover  
194 type. The use of stable land cover classes was necessary to determine and describe the patterns of ET, LST, and  
195 EVI over time and assess the effects of WTD on such trends without the confounding effect of land cover change.  
196 Further, this choice avoids potential circularity in using land cover classification to detect an effect on a parameter  
197 that uses land cover classification to produce its modelled value.

198 Finally, Tropical Rainfall Measuring Mission (hereafter TRMM) 3B42 provides daily precipitation at a resolution  
199 of  $0.25^{\circ}$  (downloaded from [earthdata.nasa.gov](http://earthdata.nasa.gov)). The TRMM data was used to calculate the seasonality of rainfall,  
200 i.e. start of the dry season and the wet season (see below for further details).

## 201 2.2.2 Water table depth

202 Water table depth (WTD) values were extracted from the Fan et al. (2010) equilibrium WTD model of South  
203 America at 30 arc seconds (~ 1 km). The model was created using a combination of literature reported depths and  
204 national databases of groundwater table depth most of which are from drinking water wells mostly from areas of  
205 high population. This data is interpolated using a groundwater model forced by present day climate, terrain, and  
206 sea level. Very few of these observation points are located in or near our study area. We used the output of the  
207 model to obtain WTD data, which was projected to the same sinusoidal projection of the MODIS data. The  
208 equilibrium WTD model is intended for use in dynamic simulations, and although our study is not the intended  
209 use of the WTD model, it is the best currently available. The authors compared their WTD calculations with values  
210 reported in the literature and found good agreement and shallower WTD however, the model over estimated deep  
211 WTD. We selected two broad WTD classes in order to further reduce some of the uncertainty around this key  
212 parameter: Shallow <2 m and Deep >8 m (and we will refer to these as such from hereafter). Figure 2. shows a  
213 theoretical graphical representation of the difference between forest (deep rooting depth) and crop (shallow rooting  
214 depth) land cover classes. These depths were selected as they represented rooting depth values for crop and forest  
215 vegetation from literature (Fan et al., 2016; Moreira et al., 2000; Nepstad et al., 1994; Setiyono et al., 2008).

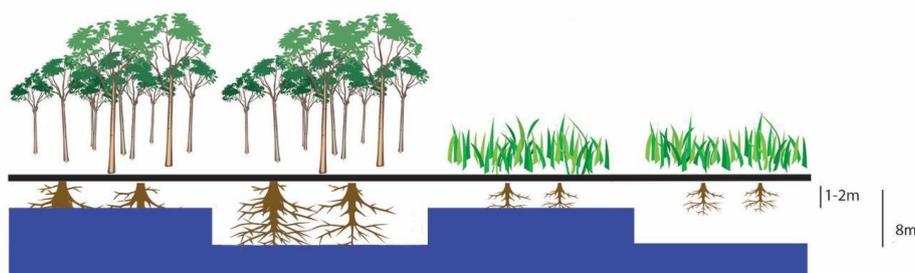
## 216 2.3 Sampling design

### 217 2.3.1 Spatial sampling

218 We chose to avoid pixels which experienced LULCC during the study period as we wanted to use the full time  
219 series for each pixel. We used MODIS land cover to identify pixels of each land cover class which remained  
220 unchanged between years and used these for analysis. We combined three land cover classes with the two water  
221 table depths and analysed the following classes: Forest Deep, Forest Shallow, Savanna Deep, Savanna Shallow,  
222 Crop Deep, and Crop Shallow.



223 For each class, we randomly selected 1000 pixels and performed this random selection 20 times to account for the  
224 effect of the randomization process in the results. This random selection method increased computational  
225 efficiency by limiting the number of total pixels examined and producing even group sizes for statistical analysis.  
226



227  
228 **Figure 2: Diagram showing that forest root depth can infiltrate until the saturated zone in both shallow (< 2 m) and**  
229 **deep (> 8 m) WTD, while other vegetation has a lower maximum rooting depth. These roots may infiltrate soil until the**  
230 **saturated zone in shallow WTD (< 2 m); however, does not penetrate further in deep WTD (> 8 m).**

### 231 2.3.2 Data analysis

232 The Amazon arc of deforestation is located in a region that has two major seasons defined by the difference in  
233 rainfall, the wet season from October to March (approximately 1500 mm) and the dry season from April to  
234 September (approximately 400 mm). The difference in rainfall can have significant impacts as the area can be  
235 prone to both seasonal flooding and droughts. In recent years the Amazon arc of deforestation has undergone an  
236 increased frequency of extreme weather events with drought in 2005, 2010 and flooding in 2009, 2012 (Nobre et  
237 al., 2016). These extreme climatic conditions can have a large influence on ET, investigation into the drivers of  
238 these extreme variations and how each land cover class is influenced is however beyond the scope of this study.

239  
240 Analysis of the data was conducted using two primary time periods. At first, we compared mean daily values of  
241 ET, EVI and LST between deep and shallow WTD as this gives an indication of the influence of WTD on our land  
242 cover classes without considering the seasonal variation. Secondly, we compared ET, EVI and LST of our land  
243 cover classes during the dry and wet season transition periods.

244  
245 For each year we calculated the dry and wet season transitions using daily precipitation data from the TRMM  
246 mission with the anomalous accumulation method of Liebmann et al., (2007). This method uses the following  
247 equation:

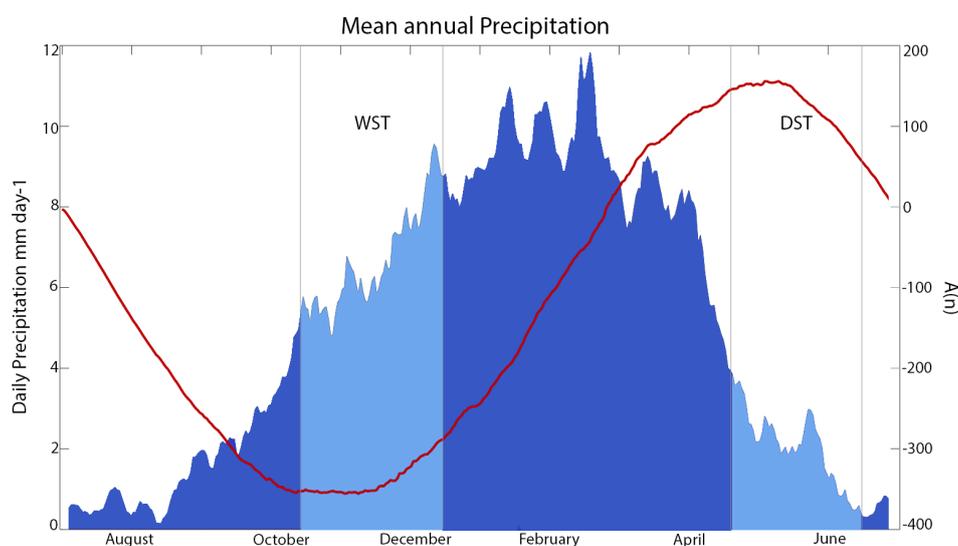
$$248 \quad A(n) = \sum_{n=1}^{day} [R(n) - \bar{R}]$$

249 Where  $R(n)$  is daily precipitation and  $\bar{R}$  is the average daily precipitation. Calculation of the anomalous  
250 accumulation begins at the driest month of the year, when the difference between daily precipitation and annual



251 average is summed to a running total of the anomalous accumulation ( $A$ ). The wet season onset is defined as the  
252 beginning of the longest period where the anomalous accumulation remains positive while the dry season onset is  
253 defined as the day after this anomalous accumulation reaches its maximum (Figure 3). These onset points of the  
254 dry and wet seasons were applied to find the closest time stamp from each MODIS product in the time series. We  
255 then considered the dry season transition to last on average 8 repeats in the MODIS record (5 for EVI due to the  
256 lower frequency of the product) and the wet season transition 7 repeats (4 for EVI). We used an average value  
257 over these transition periods to assess the difference between shallow and deep WTD on evapotranspiration.

258



259

260 **Figure 3** Mean annual precipitation of the study area calculated from TRMM with 7 day average for graphical  
261 smoothing. wet season transition (WST) and dry season transition (DST) periods are represented in lighter blue. The  
262 vertical lines represent the average start and end dates, however exact dates were calculated per year between July 2001  
263 and July 2012. Red line represents anomalous accumulation method  $A(n)$  from Liebmann et al. (2007).

264 These seasonal periods were selected as we wanted to examine the difference between deep and shallow WTD  
265 during a period when access to moisture is limiting. We hypothesized that differences between deep and shallow  
266 WTD will be most discernible during the dry season transition as rainfall is limited and vegetation rely on deep  
267 soil moisture to survive this season, while during the wet season it might not be necessary as the higher rainfall  
268 provides more near surface water. Thus, vegetation in shallow WTD areas may have higher access to water as  
269 their root zone is closer to the water table this will likely produce higher ET, EVI and lower LST during the dry  
270 season transition than vegetation in deep WTD areas. This is because the WTD is much deeper and further from  
271 the vegetation rooting zone, which leads to a lack of access to water and the vegetation will likely be stressed.  
272 Similarly, during the wet season transition, we expect vegetation in shallow WTD to exhibit higher ET, EVI and  
273 lower LST than that in deep WTD because vegetation cannot yet be sustained by precipitation alone. In shallow  
274 WTD areas, vegetation will have greater access to the WTD areas and show accelerated growth when compared  
275 to deep WTD areas.



276 We tested whether ET, LST, and EVI followed a normal distribution using the Kolmogorov–Smirnov test. This  
277 test served two purposes, to assess whether parametric statistics could be used and also indicate whether the WTD  
278 influences the frequency distribution of ET, LST, and EVI. Since a large number of response variables were not  
279 normally distributed, we chose to use non-parametric methods. Therefore, Wilcoxon rank sum test was used to  
280 test whether there was a significant difference in median ET, LST, and EVI due to the deep and shallow water  
281 table.

282 We further examined the frequency distribution of deep and shallow WTD of each of the datasets using the  
283 methodology of Wilcox (2012) where the lower and upper quantiles of the distribution are compared. Wilcox’s  
284 method utilises bootstrapping in order to compare the distribution of the 10<sup>th</sup> and 90<sup>th</sup> quantile using the Wilcoxon–  
285 Mann–Whitney test. Due to our large sample size, 100 bootstrapped datasets were used.

286 Statistical analysis between each deep and shallow land cover pair was performed separately each year for all 20  
287 randomisations e.g. differences in forest ET was tested for significance 12 years \* 20 randomisations. A year was  
288 considered statistically significant when more than 66.7% of randomisations were significant and an overall  
289 significance was determined if the majority (>50 %) of the years were significant. Statistical analysis was  
290 performed using Matlab R2018a (The MathWorks Inc., Natick, USA) statistical toolbox and Wilcox (2012)  
291 quantile distribution tool.

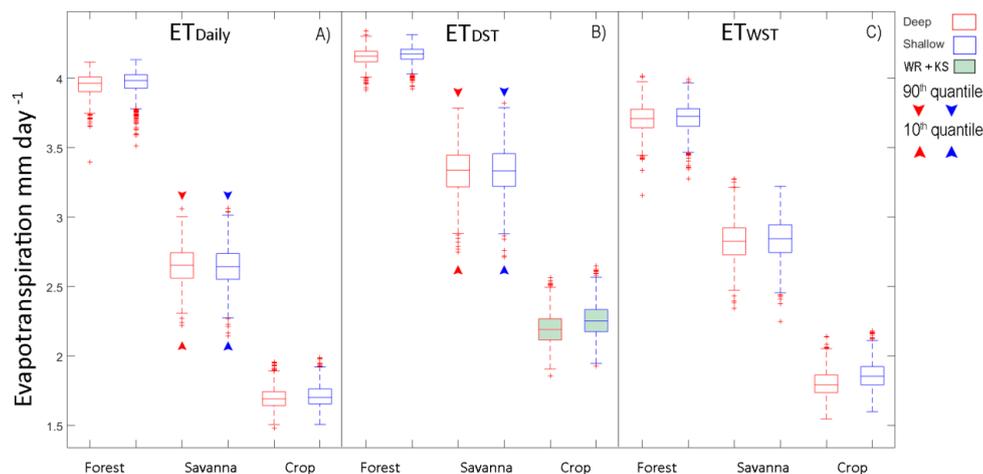
### 292 3. Results

293 The following results section is split into three subsections, one for each of the MODIS products used in the  
294 analysis. Each of the subsections and accompanying figures follows the same structure. Each figure uses three  
295 panels for the three time periods on the analysis A) annual daily mean, B) daily mean during DST, C) daily mean  
296 during WST. Each panel has three pairs of box plots which represent the deep and shallow WTD data for forest,  
297 savanna and crop.

#### 298 3.1 Effect of ground water depth on Evapotranspiration

299 None of the three land cover classes had significant differences in the average daily evapotranspiration ( $ET_{\text{daily}}$ )  
300 between deep and shallow WTD areas (Figure 4A). However, while we did not find consistent significant  
301 differences in both forest and crop  $ET_{\text{daily}}$  we do see a trend towards higher  $ET_{\text{daily}}$  in shallow WTD areas for both  
302 crop and forest (average  $\pm$  standard deviation: Forest Deep =  $3.953 \pm 0.08 \text{ mm day}^{-1}$ , Forest Shallow  $3.967 \pm 0.09$   
303  $\text{mm day}^{-1}$ ; Crop Deep =  $1.697 \pm 0.07 \text{ mm day}^{-1}$ , Crop Shallow =  $1.713 \pm 0.08 \text{ mm day}^{-1}$ ). Interestingly, we found  
304 significant differences for Savanna at the extremes of the distributions, depicted by the arrows in Figure 4A. Both  
305 the 10<sup>th</sup> and 90<sup>th</sup> quantiles of  $ET_{\text{daily}}$  were significantly higher in deep WTD areas than in shallow (difference of  
306 10<sup>th</sup> =  $0.017 \text{ mm day}^{-1}$ , difference of 90<sup>th</sup> =  $0.02 \text{ mm day}^{-1}$ , see supplemental information table S.2.4 for all the  
307 quantile analyses).

308



309

310 **Figure 4. (A) Average daily evapotranspiration (ET) annually  $ET_{daily}$ , (B) during the dry season transition period  $ET_{DST}$ , (C) during the wet season transition  $ET_{WST}$ . Red boxes represent deep WTD Blue boxes represent shallow WTD. Significant results are shown by the green filled boxes if significance was found with both Wilcoxon Rank (WR) and Kolmogorov-Smirnov (KS). Significant differences in 10<sup>th</sup> and 90<sup>th</sup> quantile are depicted by the arrows.**

314

315 We find that land cover types show different seasonal behaviour (see supplemental information figures S.3.1, S.3.2  
 316 and S.3.3). As hypothesized, crop ET during the dry season transition (hereafter  $ET_{DST}$ ) was significantly higher  
 317 in shallow than deep WTD areas (average  $\pm$  standard deviation ET: Deep =  $2.196 \pm 0.11$  mm day<sup>-1</sup>, Shallow =  $2.26$   
 318  $\pm 0.12$  mm day<sup>-1</sup>, see the green filled boxes in Fig 4B). Again we observed significant differences at the extremes  
 319 of the distribution for savanna, on average the 10<sup>th</sup> quantile of  $ET_{DST}$  was higher in shallow (average difference =  
 320  $0.003$  mm day<sup>-1</sup>) and on average the 90<sup>th</sup> quantile of  $ET_{DST}$  was higher in shallow (average difference =  
 321  $0.005$  mm day<sup>-1</sup>).

322 During the wet season transition (WST), while on average  $ET_{WST}$  was higher in shallow WTD areas than in deep  
 323 WTD areas (average difference: Forest =  $0.01$  mm day<sup>-1</sup>; Savanna =  $0.01$  mm day<sup>-1</sup>; Crop =  $0.06$  mm day<sup>-1</sup>) this  
 324 difference was not significant (Figure 4C).

325

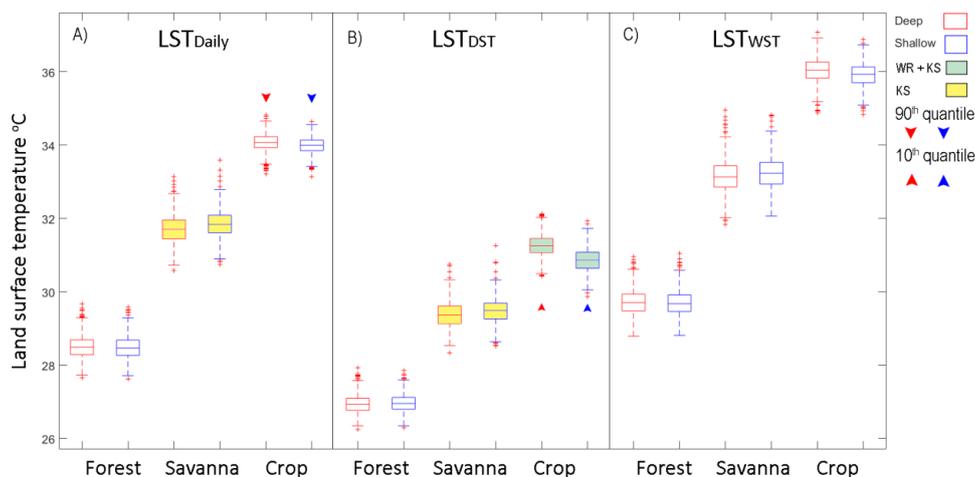
### 326 3.2 Effect of ground water depth in Land Surface Temperature

327 We found that the distribution of the average land surface temperature ( $LST_{daily}$ ) was significantly different only  
 328 for savanna and the 90<sup>th</sup> quantile of crop. Deep WTD areas of savanna showed a distribution skewed towards lower  
 329 temperatures (average  $\pm$  standard deviation LST: Deep =  $31.705$  °C  $\pm 0.38$ , Shallow =  $31.848$  °C  $\pm 0.37$ ), see yellow  
 330 filled boxes in Figure 5A. The 90<sup>th</sup> quantile of crop  $LST_{daily}$  deep WTD areas was on average  $0.1$  °C higher than in  
 331 shallow WTD areas. Although this is only part of the distribution, it indicates that the warmest crop areas are found  
 332 in deep WTD.



333

334



335

336 **Figure 5. (A) Average daily land surface temperature (LST) annually LST<sub>daily</sub>, (B) during the dry season transition**  
 337 **period LST<sub>DST</sub>, (C) during the wet season transition LST<sub>WST</sub>. Red boxes represent deep WTD Blue boxes represent**  
 338 **shallow WTD. Yellow filled boxes represent a statistical difference in skewness, calculated by Kolmogorov–Smirnov,**  
 339 **and green filled boxes represent statistical differences by both Wilcoxon-rank and Kolmogorov–Smirnov. Significant**  
 340 **differences in 10<sup>th</sup> and 90<sup>th</sup> quantile are depicted by the arrows.**

341

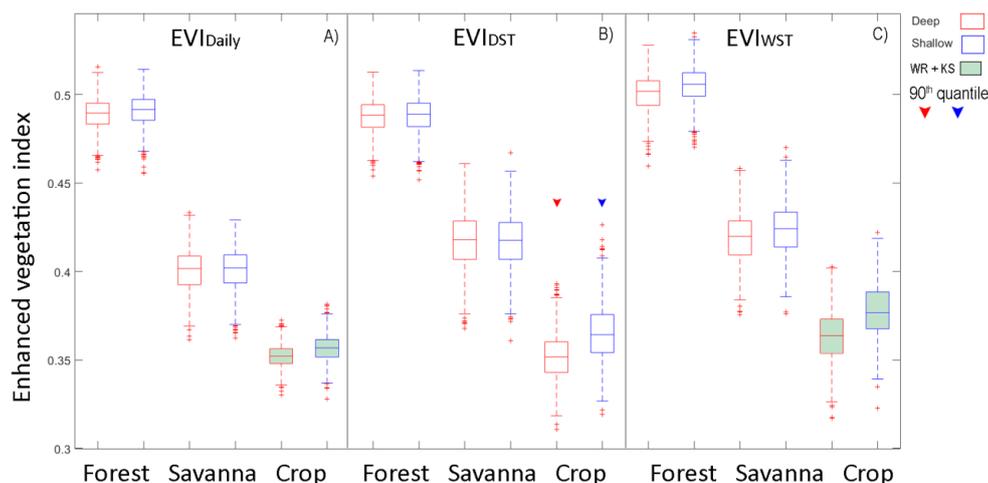
342 LST also shows clear seasonal patterns (Supplemental information figure S.5.1, S.5.2 and S.5.3). During the dry  
 343 season transition, we found that crop in deep WTD areas had a significantly higher LST than in shallow WTD  
 344 areas (average  $\pm$  standard deviation LST: Deep =  $31.256 \pm 0.29$  °C, Shallow =  $30.864 \pm 0.31$  °C, green filled boxes  
 345 in Figure 5B). In addition, the 10<sup>th</sup> quantile of the crop distributions was significantly higher by 0.42 °C in deep  
 346 WTD areas than in shallow. During these periods we found again a significant difference in the distribution of  
 347 savanna, where deep savanna distribution was skewed towards lower LST values. No significant differences were  
 348 found during the wet season transition (Figure 5C).

### 349 3.3 Effect of ground water depth on Enhanced Vegetation Index

350 We found significant differences in daily average EVI (EVI<sub>daily</sub>) between deep and shallow only in crop (average  
 351  $\pm$  standard deviation EVI: Deep =  $0.352 \pm 0.01$ ; Shallow =  $0.357 \pm 0.01$ ), with shallow WTD areas EVI being  
 352 higher than that of deep WTD areas (Figure 6A green filled boxes).



353



354

355 **Figure 6.** (A) Average enhanced vegetation index (EVI) annually  $EVI_{daily}$ , (B) during the dry season transition period  
356  $EVI_{DST}$ , (C) during the wet season transition  $EVI_{WST}$ . Red boxes represent deep WTD Blue boxes represent shallow  
357 WTD. Green filled boxes represent statistical differences by both Wilcoxon-rank and Kolmogorov-Smirnov tests.  
358 Significant differences in 90<sup>th</sup> quantile are depicted by the arrows.

359

360 Seasonality in EVI is shown in Supplemental information figure S.7.1, S.7.2 and S.7.3. Mean EVI during the dry  
361 season transition ( $EVI_{DST}$ ) for crop showed a trend towards higher EVI in shallow WTD areas; however, this  
362 difference was only significant in 5 of the 11 years and therefore is not considered consistent enough to be  
363 statistically significant (average  $\pm$  standard deviation EVI: Deep =  $0.352 \pm 0.01$ , Shallow =  $0.3656 \pm 0.01$ . Figure  
364 6B, Table S.6.8). The 90<sup>th</sup> quantile EVI of crop was significantly higher in shallow WTD areas than deep. During  
365 the wet season transition ( $EVI_{WST}$ ), crop was the only different class where EVI was significantly higher in shallow  
366 WTD areas than in deep WTD areas (average  $\pm$  standard deviation EVI: Deep =  $0.364 \pm 0.01$ , Shallow =  $0.378 \pm$   
367  $0.02$ , green filled boxes in Figure 6C).

#### 368 4 Discussion

369 In this study, we tested the hypothesis that areas of shallow water table depth (WTD) would have higher  
370 evapotranspiration when compared to areas of deep WTD, primarily in shallow rooted crop vegetation. The annual  
371 daily mean ET, however, was not different between crop in deep and shallow WTD areas. Since crop experiences  
372 high seasonality, this annual variability may override differences between deep and shallow WTD areas in the  
373 daily average values of the MODIS products. For example, average crop ET reaches a high of  $3.5 \text{ mm day}^{-1}$  in the  
374 wet season while the dry season ET reaches a low of  $0.4 \text{ mm day}^{-1}$ . Interestingly we found significant differences  
375 in annual mean LST and EVI for crop. For LST, we found that the upper 90<sup>th</sup> quantile was  $0.11 \text{ }^\circ\text{C}$  higher in deep  
376 than in shallow WTD areas. While this difference is only found in the 90<sup>th</sup> quantile of the distribution it does  
377 indicate that local conditions can be much warmer in deep WTD areas than shallow. In addition, we found that  
378 crop in shallow WTD areas had a significantly higher EVI than in deep WTD. Crop EVI in shallow WTD areas is



379 1.2 % higher than in deep WTD. This provides support to our hypothesis that crop would have higher EVI in  
380 shallow WTD compared to deep WTD areas. The rooting depth of crop vegetation only penetrates to a maximum  
381 of 2 m, in shallow WTD areas this means the root zone is close to the WTD and would have access to water while  
382 in deep WTD the roots are far from the saturated zone. This access to water in shallow WTD areas could also lead  
383 to higher ET and therefore evaporative cooling explaining the cooler temperatures in the 90<sup>th</sup> quantile.

384

385 The second part of our hypothesis was that the effect of WTD would be most evident during the transition periods  
386 between wet and dry seasons when rainfall is reduced and vegetation activity is limited by access to soil moisture.  
387 We found support for this hypothesis during the dry season transition. In the DST, crop ET was significantly higher  
388 in shallow WTD areas and crop LST was significantly lower in shallow WTD areas, while in crop EVI we saw a  
389 trend towards higher EVI in shallow WTD areas (significant differences were only found in 5 of the 11 years).  
390 While the difference in crop ET is not large (0.063 mm day<sup>-1</sup>, 2.9 % higher in shallow); during the dry season  
391 transition, the results are important as they indicate that crops in the shallow WTD areas have a delayed response  
392 to lower rainfall and have a relatively longer growing season. Crop LST further supports our hypothesis as LST in  
393 deep WTD areas was 0.39 °C higher than in shallow WTD areas, while no significant effects were found in EVI.  
394 Therefore cooler temperatures in shallow WTD areas are expected to be the result of higher evaporative cooling  
395 from ET. If this extra warming above the canopy is caused by a change in ET, then better estimates of ET should  
396 be possible, however, this is not trivial (Glenn et al., 2007). These relatively low differences in ET as measured  
397 with MODIS data might also be due to the ET product itself. The MODIS ET model is not optimised for  
398 comparison over relatively small spatial extents and short temporal scales (Ruhoff et al., 2013). In addition, the  
399 model does not take into account water storage and ET is based largely on atmospheric forcing and global land  
400 cover parameterisation. Therefore, the modelled data was expected to underperform, making the differences we  
401 found for the dry season even more important.

402

403 Ponte De Souza et al. (2011) highlighted that one of the strongest impacts of LULCC from forest to crop was due  
404 to the simultaneous 85% increase in sensible heat flux and 78% reduction in latent heat recorded during the dry  
405 season. Studies examining the change in LST due to LULCC found that LST increased by 6 °C from forest to crop  
406 (Silvério et al., 2015) and 1.5 °C from savanna to crop (Loarie et al., 2011). Further global models estimated an  
407 increase of 5 °C during the summer season for the Amazon, due to a shift from forest to grass (Brovkin et al., 2009;  
408 Dekker et al., 2010). This could mean that in deep WTD temperature could even be higher following LULCC,  
409 however, WTD were not used as input for these modelling studies. Our results show a maximum temperature of  
410 30 °C in forest compared to a maximum temperature of 38 °C in crops.

411

412 We also expected that the influence of WTD would also be important during the wet season transition, as in this  
413 period rainfall is increasing. In areas of shallow WTD vegetation may still access water to supplement if rainfall  
414 is not sufficient as their root zone is close to the water table. Therefore, vegetation growth may be accelerated in  
415 comparison to areas of deep WTD which rely more directly on precipitation. Crop EVI was significantly higher in  
416 shallow WTD than deep by about 3.8 %, and this was the only data for which we found a significant difference.  
417 EVI measures vegetation greenness and could be an indication of more rapid growth in shallow WTD areas. As  
418 EVI data is directly observed and not modelled the differences are solely reliant on differences in reflected



419 radiation. It may be that smaller differences between deep and shallow WTD areas are more easily detectable using  
420 this data. Along the arc of deforestation observations of a lengthening dry season since the 1970s, are linked to a  
421 delay in the WST (Butt et al., 2011; Fu et al., 2013). This delay correlates with LULCC and the large reduction  
422 this has on ET (Debortoli et al., 2017). Although the difference in WTD seen in crops does not have a strong  
423 influence on ET when compared to the difference in ET between the land cover classes, evidence of earlier or  
424 faster growth due to the shallow WTD could be beneficial on a local scale.

425

426 These results are even more relevant when comparing the effects of WTD in crop and forest. As forest has been  
427 shown to maintain ET throughout the seasons (Kunert et al., 2017) as its deep roots access deeper groundwater  
428 (Gash and Nobre, 1997; Nepstad et al., 1994), we hypothesised that no change should be observed in ET, LST,  
429 and EVI. Indeed, we found no significant differences across the three MODIS products, both annually or during  
430 the dry and wet season transitions. While this does not directly support our hypothesis about the role of WTD for  
431 shallow rooted vegetation, this does help validate that our approach reflects our knowledge of the system for  
432 vegetation with deep roots.

433

434 Savanna is a complex land cover type both in terms of its natural structure (Staal et al., 2018b) and because it is  
435 challenging to classify with remote sensing data (Gibbes et al., 2010). MODIS classification accuracy of savanna  
436 is about 40 %, about half of that of forest and crop (90 % and 80 % respectively) (Friedl et al., 2010). Savanna  
437 includes both trees and grasses, which through the deep roots of trees may access moisture directly and facilitate  
438 moisture uptake via hydraulic redistribution (Oliveira et al., 2005) and large areas of shallow root grasses without  
439 trees would be negatively affected by water stress. A number of the findings for savanna were not in line with our  
440 proposed hypothesis. The distribution between shallow and deep LST was significantly different, with deep WTD  
441 areas having a skewed distribution towards lower temperatures. In our hypothesis, we expected to find lower  
442 temperature where shallow WTD occurs or no differences in temperature. A similar trend was found in ET where  
443 the 10<sup>th</sup> and 90<sup>th</sup> quantiles of the distribution were significantly higher in deep WTD areas. The difference in ET  
444 was very small < 1 % difference between deep and shallow.

445

446 The differences found for crop support our hypothesis that shallow WTD areas may facilitate water uptake  
447 compared with areas of deep WTD during the transition between wet and dry seasons. Previous crop production  
448 studies have shown that artificially maintaining a shallow WTD through sub irrigation systems can increase the  
449 productivity of crops such as soy (Kahlow et al., 2005; Mejia et al., 2000) but this has not been previously shown  
450 in the naturally occurring shallow WTD areas of the arc of deforestation in the Amazon. In deep WTD areas, crop  
451 vegetation undergoes more severe water stress compared with shallow WTD further reducing evapotranspiration  
452 and its potential impact on the moisture recycling system. At the regional scale, the difference between deep and  
453 shallow WTD is not that important. The most significant differences in ET are driven by deforestation and strong  
454 annual variations in rainfall. Although not analysed specifically in this study, the remote sensing data clearly shows  
455 these distinctions between different land cover classes and high seasonal and inter-annual variability. On a local  
456 scale, the difference between deep and shallow WTD on crop may be of great importance. During the dry season  
457 transition crop areas in shallow WTD maintained higher ET. This difference may be important for overall  
458 productivity as the dry season influence is delayed and as a result, is increasing the growing season length. This



459 could facilitate natural double cropping systems without the need for investment in irrigation which is still an  
460 uncommon practice in the Amazon arc of deforestation (Lathuilière et al., 2012). Agricultural intensification is a  
461 pathway to increasing the sustainability of agriculture in the arc of deforestation if it prevents or reduces  
462 deforestation or facilitates reforestation (Oliveira et al., 2013). If agricultural productivity can be increased by  
463 focusing on already cleared shallow WTD areas, areas of deep WTD could be reforested or returned to secondary  
464 forest. Reforestation of previously degraded or logged forest has been shown to return to near natural levels of ET  
465 within a few years (Davidson et al., 2012; Hölscher et al., 1997).

466

467 This study is a first approach into gaining a better understanding on the influence of shallow WTD on shallow  
468 rooted vegetation and it heavily relies on models and remote sensing data which are most appropriate for analyses  
469 at larger spatial and temporal scales. The results presented here are limited by the inherent uncertainty of the data  
470 used, both in the WTD model and the remote sensing data. Although we believe that the WTD model used here is  
471 the best currently available, due to limited data availability it was created using data located mostly in the coastal  
472 regions of the continent with very few observations from near our study site (Fan and Miguez-Macho, 2010). In  
473 this study, the authors note that there is an overestimation of deep WTD areas when validated against literature  
474 reported values. We believe that by the use of a conservative definition of deep WTD >8 m the model outputs are  
475 appropriate for our purposes. As discussed above, the remote sensing data has obvious limitations but does provide  
476 some insights into how depth of the water table at a local scale might affect water transfer and evaporative  
477 processes. Nonetheless, the second main source of uncertainty is in the MODIS land cover classification. We chose  
478 to use this land cover classification as the ET and LST products use this classification in their algorithm. Although  
479 the classes used are broad and do not reflect the full complexity and heterogeneity of the arc of deforestation, they  
480 are robust enough for our purposes. As the influence of WTD on ET is most relevant on smaller scales, further  
481 research in these areas could focus on the smaller spatial scales and validate sites with accurate classification and  
482 WTD measurements.

483

## 484 **5. Conclusions**

485 The results of this study indicate that crop vegetation in areas of shallow WTD undergoes less water stress than  
486 crop vegetation in deep WTD areas, especially during the dry season transition. The influence this difference has  
487 on regional evapotranspiration or on the moisture recycling service might be low and most likely not influential,  
488 because most of the effects are in the conversion from forest to crop. However, on a local level, this buffering  
489 may essentially shorten the total dry season length by delaying the water stress response. This difference could  
490 have implications for agricultural productivity in the Amazon arc of deforestation. Utilising areas of shallow WTD  
491 to increase agricultural efficiency could mean a reduction in the total agricultural area allowing for areas of deep  
492 WTD to be reforested or prevented from being deforested. As evapotranspiration rates of secondary forests have  
493 been shown to reach near natural levels, reforestation of these areas could directly benefit the moisture recycling  
494 ecosystem services. These ecosystem services provided by the Amazon are highly valuable and ideally should be  
495 prioritised. The possible introduction of payment for ecosystem services could target deep WTD areas as they may  
496 be of lower agricultural value, very inefficient in terms of their water use, but still, provide the desired ecosystem  
497 services.



498 An unknown factor will be whether the role of groundwater table depth will become more important in the future  
499 due to changes in climate. Many modelling studies predict that the dry season will lengthen and that the arc of  
500 deforestation will move to a warmer drier climate (Bonan, 2008). Under this scenario, where possible it would be  
501 wisest to conserve the natural forest ecosystems of the Amazon, especially where groundwater is deepest in order  
502 to maintain the moisture recycling services as well as the local climate cooling effect.



503 **References**

- 504 Arvor, D., Dubreuil, V., Ronchail, J., Simões, M. and Funatsu, B. M.: Spatial patterns of rainfall regimes related  
505 to levels of double cropping agriculture systems in Mato Grosso (Brazil), *Int. J. Climatol.*, 34(8), 2622–2633,  
506 doi:10.1002/joc.3863, 2014.
- 507 Asner, G. P., Scurlock, J. M. O. and A. Hicke, J.: Global synthesis of leaf area index observations: implications  
508 for ecological and remote sensing studies, *Glob. Ecol. Biogeogr.*, 12(3), 191–205, doi:10.1046/j.1466-  
509 822X.2003.00026.x, 2003.
- 510 Badger, A. M. and Dirmeyer, P. A.: Climate response to Amazon forest replacement by heterogeneous crop cover,  
511 *Hydrol. Earth Syst. Sci.*, 19(11), 4547–4557, doi:10.5194/hess-19-4547-2015, 2015.
- 512 Bonan, G. B.: Forests and climate change: forcings, feedbacks, and the climate benefits of forests., *Science*,  
513 320(5882), 1444–1449, doi:10.1126/science.1155121, 2008.
- 514 Brando, P. M., Balch, J. K., Nepstad, D. C., Morton, D. C., Putz, F. E., Coe, M. T., Silverio, D., Macedo, M. N.,  
515 Davidson, E. A., Nobrega, C. C., Alencar, A. and Soares-Filho, B. S.: Abrupt increases in Amazonian tree  
516 mortality due to drought-fire interactions, *Proc. Natl. Acad. Sci.*, 111(17), 6347–6352,  
517 doi:10.1073/pnas.1305499111, 2014.
- 518 Brovkin, V., Raddatz, T., Reick, C. H., Claussen, M. and Gayler, V.: Global biogeophysical interactions between  
519 forest and climate, *Geophys. Res. Lett.*, 36(7), 1–5, doi:10.1029/2009GL037543, 2009.
- 520 Butt, N., de Oliveira, P. A. and Costa, M. H.: Evidence that deforestation affects the onset of the rainy season in  
521 Rondonia, Brazil, *J. Geophys. Res.*, 116(D11), D11120, doi:10.1029/2010JD015174, 2011.
- 522 Chambers, J. Q., Asner, G. P., Morton, D. C., Anderson, L. O., Saatchi, S. S., Espírito-Santo, F. D. B., Palace, M.  
523 and Souza, C.: Regional ecosystem structure and function: ecological insights from remote sensing of tropical  
524 forests, *Trends Ecol. Evol.*, 22(8), 414–423, doi:10.1016/j.tree.2007.05.001, 2007.
- 525 Costa, M. H. and Foley, J. A.: Combined effects of deforestation and doubled atmospheric CO<sub>2</sub> concentrations on  
526 the climate of Amazonia, *J. Clim.*, 13(1), 18–34, doi:10.1175/1520-0442(2000)013<0018:CEODAD>2.0.CO;2,  
527 2000.
- 528 Costa, M. H. and Pires, G. F.: Effects of Amazon and Central Brazil deforestation scenarios on the duration of the  
529 dry season in the arc of deforestation, *Int. J. Climatol.*, 30(13), 1970–1979, doi:10.1002/joc.2048, 2010.
- 530 Costa, M. H., Yanagi, S. N. M., Souza, P. J. O. P., Ribeiro, A. and Rocha, E. J. P.: Climate change in Amazonia  
531 caused by soybean cropland expansion, as compared to caused by pastureland expansion, *Geophys. Res. Lett.*,  
532 34(7), 2–5, doi:10.1029/2007GL029271, 2007.
- 533 Davidson, E. a., de Araújo, A. C., Artaxo, P., Balch, J. K., Brown, I. F., C. Bustamante, M. M., Coe, M. T., DeFries,  
534 R. S., Keller, M., Longo, M., Munger, J. W., Schroeder, W., Soares-Filho, B. S., Souza, C. M. and Wofsy, S. C.:  
535 The Amazon basin in transition, *Nature*, 481(7381), 321–328, doi:10.1038/nature10717, 2012.



- 536 Debortoli, N. S., Dubreuil, V., Hirota, M., Filho, S. R., Lindoso, D. P. and Nabucet, J.: Detecting deforestation  
537 impacts in Southern Amazonia rainfall using rain gauges, *Int. J. Climatol.*, 37(6), 2889–2900,  
538 doi:10.1002/joc.4886, 2017.
- 539 Dekker, S. C., de Boer, H. J., Brovkin, V., Fraedrich, K., Wassen, M. J. and Rietkerk, M.: Biogeophysical  
540 feedbacks trigger shifts in the modelled vegetation-atmosphere system at multiple scales, *Biogeosciences*, 7(4),  
541 1237–1245, doi:10.5194/bg-7-1237-2010, 2010.
- 542 Eltahir, E. A. B. and Bras, R. L.: Precipitation recycling in the Amazon basin, *Q. J. R. Meteorol. Soc.*, 120(518),  
543 861–880, doi:10.1002/qj.49712051806, 1994a.
- 544 Eltahir, E. A. B. and Bras, R. L.: Sensitivity of regional climate to deforestation in the Amazon basin, *Adv. Water*  
545 *Resour.*, 17(1–2), 101–115, doi:10.1016/0309-1708(94)90027-2, 1994b.
- 546 van der Ent, R. J. and Savenije, H. H. G.: Length and time scales of atmospheric moisture recycling, *Atmos. Chem.*  
547 *Phys.*, 11(5), 1853–1863, doi:10.5194/acp-11-1853-2011, 2011.
- 548 Fan, J., McConkey, B., Wang, H. and Janzen, H.: Root distribution by depth for temperate agricultural crops, *F.*  
549 *Crop. Res.*, 189, 68–74, doi:10.1016/j.fcr.2016.02.013, 2016.
- 550 Fan, Y. and Miguez-Macho, G.: Potential groundwater contribution to Amazon evapotranspiration, *Hydrol. Earth*  
551 *Syst. Sci.*, 14(10), 2039–2056, doi:10.5194/hess-14-2039-2010, 2010.
- 552 Fearnside, P. M.: Environmental services as a strategy for sustainable development in rural Amazonia, *Ecol. Econ.*,  
553 20(1), 53–70, doi:10.1016/S0921-8009(96)00066-3, 1997.
- 554 Foley, J. A., Asner, G. P., Costa, M. H., Coe, M. T., DeFries, R., Gibbs, H. K., Howard, E. A., Olson, S., Patz, J.,  
555 Ramankutty, N. and Snyder, P.: Amazonia revealed: forest degradation and loss of ecosystem goods and services  
556 in the Amazon Basin, *Front. Ecol. Environ.*, 5(1), 25–32, doi:10.1890/1540-9295(2007)5[25:ARFDAL]2.0.CO;2,  
557 2007a.
- 558 Foley, J. A., Asner, G. P., Heil Costa, M., Coe, M. T., DeFries, R., Gibbs, H. K., Howard, E. A., Olson, S., Patz,  
559 J., Ramankutty, N. and Snyder, P.: Amazonia revealed: forest degradation and loss of ecosystem goods and  
560 services in the Amazon Basin, *Front Ecol Env.*, 5(1), 25–32, doi:http://doi.wiley.com/10.1890/1540-  
561 9295(2007)5[25:ARFDAL]2.0.CO;2, 2007b.
- 562 Friedl, M. A., Sulla-Menashe, D., Tan, B., Schneider, A., Ramankutty, N., Sibley, A. and Huang, X.: MODIS  
563 Collection 5 global land cover: Algorithm refinements and characterization of new datasets, *Remote Sens.*  
564 *Environ.*, 114(1), 168–182, doi:10.1016/j.rse.2009.08.016, 2010.
- 565 Fu, R., Yin, L., Li, W., Arias, P. A., Dickinson, R. E., Huang, L., Chakraborty, S., Fernandes, K., Liebmann, B.,  
566 Fisher, R. and Myneni, R. B.: Increased dry-season length over southern Amazonia in recent decades and its  
567 implication for future climate projection, *Proc. Natl. Acad. Sci.*, doi:10.1073/pnas.1302584110, 2013.
- 568 Gash, J. H. C. and Nobre, C. A.: Climatic Effects of Amazonian Deforestation: Some Results from ABRACOS,  
569 *Bull. Am. Meteorol. Soc.*, 78(5), 823–830, doi:10.1175/1520-0477(1997)078<0823:CEOADS>2.0.CO;2, 1997.



- 570 Gibbes, C., Adhikari, S., Rostant, L., Southworth, J. and Qiu, Y.: Application of Object Based Classification and  
571 High Resolution Satellite Imagery for Savanna Ecosystem Analysis, *Remote Sens.*, 2(12), 2748–2772,  
572 doi:10.3390/rs2122748, 2010.
- 573 Glenn, E. P., Huete, A. R., Nagler, P. L., Hirschboeck, K. K. and Brown, P.: Integrating remote sensing and ground  
574 methods to estimate evapotranspiration, *CRC. Crit. Rev. Plant Sci.*, 26(3), 139–168,  
575 doi:10.1080/07352680701402503, 2007.
- 576 Gusso, A., Arvor, D., Ricardo Ducati, J., Veronez, M. R. and Da Silveira, L. G.: Assessing the modis crop detection  
577 algorithm for soybean crop area mapping and expansion in the Mato Grosso state, Brazil, *Sci. World J.*, 2014(1),  
578 doi:10.1155/2014/863141, 2014.
- 579 Hölscher, D., de A. Sá, T. ., Bastos, T. ., Denich, M. and Fölster, H.: Evaporation from young secondary vegetation  
580 in eastern Amazonia, *J. Hydrol.*, 193(1–4), 293–305, doi:10.1016/S0022-1694(96)03145-9, 1997.
- 581 Huete, A., Didan, K., Miura, H., Rodriguez, E. P., Gao, X. and Ferreira, L. F.: Overview of the radiometric and  
582 biophysical performance of the MODIS vegetation indices, *Remote Sens. Environ.*, 83, 195–213 [online]  
583 Available from: [https://ac-els-cdn-com.sire.ub.edu/S0034425702000962/1-s2.0-S0034425702000962-](https://ac-els-cdn-com.sire.ub.edu/S0034425702000962/1-s2.0-S0034425702000962-main.pdf?_tid=420e9dda-1821-11e8-b53d-00000aacb35d&acdnat=1519339356_ef83dc686b96e75a110fbad8d8dc950a)  
584 [main.pdf?\\_tid=420e9dda-1821-11e8-b53d-](https://ac-els-cdn-com.sire.ub.edu/S0034425702000962/1-s2.0-S0034425702000962-main.pdf?_tid=420e9dda-1821-11e8-b53d-00000aacb35d&acdnat=1519339356_ef83dc686b96e75a110fbad8d8dc950a)  
585 [00000aacb35d&acdnat=1519339356\\_ef83dc686b96e75a110fbad8d8dc950a](https://ac-els-cdn-com.sire.ub.edu/S0034425702000962/1-s2.0-S0034425702000962-main.pdf?_tid=420e9dda-1821-11e8-b53d-00000aacb35d&acdnat=1519339356_ef83dc686b96e75a110fbad8d8dc950a), 2002.
- 586 Kahlow, M. A., Ashraf, M. and Zia-Ul-Haq: Effect of shallow groundwater table on crop water requirements and  
587 crop yields, *Agric. Water Manag.*, 76(1), 24–35, doi:10.1016/j.agwat.2005.01.005, 2005.
- 588 Kastens, J. H., Brown, J. C., Coutinho, A. C., Bishop, C. R. and Esquerdo, J. C. D. M.: Soy moratorium impacts  
589 on soybean and deforestation dynamics in Mato Grosso, Brazil, *PLoS One*, 12(4), 1–21,  
590 doi:10.1371/journal.pone.0176168, 2017.
- 591 Kunert, N., Maria Aparecido, L. T., Wolff, S., Higuchi, N., dos, J., Carioca de Araujo, A. and Trumbore, S.: A  
592 revised hydrological model for the Central Amazon: The importance of emergent canopy trees in the forest water  
593 budget, *Agric. For. Meteorol.*, 239, 47–57, doi:10.1016/j.agrformet.2017.03.002, 2017.
- 594 Lathuillière, M. J., Johnson, M. S. and Donner, S. D.: Water use by terrestrial ecosystems: temporal variability in  
595 rainforest and agricultural contributions to evapotranspiration in Mato Grosso, Brazil, *Environ. Res. Lett.*, 7(2),  
596 024024, doi:10.1088/1748-9326/7/2/024024, 2012.
- 597 Lawrence, D. and Vandecar, K.: Effects of tropical deforestation on climate and agriculture, *Nat. Clim. Chang.*,  
598 5(1), 27–36, doi:10.1038/nclimate2430, 2014.
- 599 Liebmann, B., Camargo, S. J., Seth, A., Marengo, J. A., Carvalho, L. M. V., Allured, D., Fu, R. and Vera, C. S.:  
600 Onset and end of the rainy season in South America in observations and the ECHAM 4.5 atmospheric general  
601 circulation model, *J. Clim.*, 20(10), 2037–2050, doi:10.1175/JCLI4122.1, 2007.
- 602 Loarie, S. R., Lobell, D. B., Asner, G. P., Mu, Q. and Field, C. B.: Direct impacts on local climate of sugar-cane  
603 expansion in Brazil, *Nat. Clim. Chang.*, 1(2), 105–109, doi:10.1038/nclimate1067, 2011.
- 604 Maeda, E. E., Ma, X., Wagner, F., Kim, H., Oki, T., Eamus, D. and Huete, A.: Evapotranspiration seasonality



- 605 across the Amazon basin, *Earth Syst. Dyn. Discuss.*, (January), 1–28, doi:10.5194/esd-2016-75, 2017.
- 606 Makarieva, a M. and Gorshkov, V. G.: Biotic pump of atmospheric moisture as driver of the hydrological cycle  
607 on land, *Hydrol. Earth Syst. Sci.*, 11(2), 1013–1033, doi:10.5194/hess-11-1013-2007, 2007.
- 608 Malhi, Y., Roberts, J. T., Betts, R. a, Killeen, T. J., Li, W. and Nobre, C. a: Climate Change, Deforestation, and  
609 the Fate of the Amazon, *Science* (80-. ), 319(5860), 169–172, doi:10.1126/science.1146961, 2008.
- 610 Mejia, M. N., Madramootoo, C. A. and Broughton, R. S.: Influence of water table management on corn and  
611 soybean yields, *Agric. Water Manag.*, 46(1), 73–89, doi:10.1016/S0378-3774(99)00109-2, 2000.
- 612 Moreira, M. Z., Sternberg, L. D. L. and Nepstad, D. C.: Vertical patterns of soil water uptake by plants in a primary  
613 forest and an abandoned pasture in the eastern Amazon: an isotopic approach, *Plant Soil*, 222(1–2), 95–107,  
614 doi:10.1023/A:1004773217189, 2000.
- 615 Mu, Q., Zhao, M. and Running, S. W.: Improvements to a MODIS global terrestrial evapotranspiration algorithm,  
616 *Remote Sens. Environ.*, 115(8), 1781–1800, doi:10.1016/j.rse.2011.02.019, 2011.
- 617 Neill, C., Coe, M. T., Riskin, S. H., Krusche, A. V., Elsenbeer, H., Macedo, M. N., McHorney, R., Lefebvre, P.,  
618 Davidson, E. A., Scheffler, R., Figueira, A. M. A. S., Porder, S. and Deegan, L. A.: Watershed responses to  
619 Amazon soya bean cropland expansion and intensification, *Philos. Trans. R. Soc. Lond. B. Biol. Sci.*, 368(1619),  
620 2–7, doi:10.1098/rstb.2012.0425, 2013.
- 621 Nepstad, D. C., de Carvalho, C. R., Davidson, E. A., Jipp, P. H., Lefebvre, P. A., Negreiros, G. H., da Silva, E. D.,  
622 Stone, T. a., Trumbore, S. E. and Vieira, S.: The role of deep roots in the hydrological and carbon cycles of  
623 Amazonian forests and pastures, *Nature*, 372(6507), 666–669, doi:10.1038/372666a0, 1994.
- 624 Nobre, C. A., Sampaio, G., Borma, L. S., Castilla-Rubio, J. C., Silva, J. S. and Cardoso, M.: Land-use and climate  
625 change risks in the Amazon and the need of a novel sustainable development paradigm, *Proc. Natl. Acad. Sci.*,  
626 201605516, doi:10.1073/pnas.1605516113, 2016a.
- 627 Nobre, C. A., Marengo, J. A., Seluchi, M. E., Cuartas, A. and Alves, L. M.: Some Characteristics and Impacts of  
628 the Drought and Water Crisis in Southeastern Brazil during 2014 and 2015 Some Characteristics and Impacts of  
629 the Drought and Water Crisis in Southeastern Brazil during 2014 and 2015, *J. od Water Resour. Prot.*, 8(August),  
630 252–262, doi:10.4236/jwarp.2016.82022, 2016b.
- 631 Oliveira, L. J. C., Costa, M. H., Soares-Filho, B. S. and Coe, M. T.: Large-scale expansion of agriculture in  
632 Amazonia may be a no-win scenario, *Environ. Res. Lett.*, 8(2), 024021, doi:10.1088/1748-9326/8/2/024021, 2013.
- 633 Oliveira, R. S., Dawson, T. E., Burgess, S. S. O. and Nepstad, D. C.: Hydraulic redistribution in three Amazonian  
634 trees, *Oecologia*, 145(3), 354–363, doi:10.1007/s00442-005-0108-2, 2005.
- 635 Ponte De Souza, P. J. D. O., Ribeiro, A., Da Rocha, E. J. P., Botelho, M. D. N., De Sousa, A. M. L., De Souza, E.  
636 B. and Bouças Farias, J. R.: Imapcts of soybean expansion on the Amazon energy balance: A case study, *Exp.*  
637 *Agric.*, 47(03), 553–567, doi:10.1017/S0014479711000391, 2011.



- 638 Ruhoff, A. L., Paz, A. R., Aragao, L. E. O. C., Mu, Q., Malhi, Y., Collischonn, W., Rocha, H. R. and Running, S.  
639 W.: Assessment of the MODIS global evapotranspiration algorithm using eddy covariance measurements and  
640 hydrological modelling in the Rio Grande basin, *Hydrol. Sci. J.*, 58(8), 1658–1676,  
641 doi:10.1080/02626667.2013.837578, 2013.
- 642 Sampaio, G., Nobre, C., Costa, M. H., Satyamurty, P., Soares-Filho, B. S. and Cardoso, M.: Regional climate  
643 change over eastern Amazonia caused by pasture and soybean cropland expansion, *Geophys. Res. Lett.*, 34(17),  
644 L17709, doi:10.1029/2007GL030612, 2007.
- 645 Setiyono, T. D., Weiss, A., Specht, J. E., Cassman, K. G. and Dobermann, A.: Leaf area index simulation in  
646 soybean grown under near-optimal conditions, *F. Crop. Res.*, 108(1), 82–92, doi:10.1016/j.fcr.2008.03.005, 2008.
- 647 Sheil, D.: How plants water our planet: advances and imperatives, *Trends Plant Sci.*, 19(4), 209–211,  
648 doi:10.1016/j.tplants.2014.01.002, 2014.
- 649 Silvério, D. V., Brando, P. M., Macedo, M. N., Beck, P. S. A., Bustamante, M. and Coe, M. T.: Agricultural  
650 expansion dominates climate changes in southeastern Amazonia: The overlooked non-GHG forcing, *Environ. Res.*  
651 *Lett.*, 10(10), doi:10.1088/1748-9326/10/10/104015, 2015.
- 652 Sims, D. A., Rahman, A. F., Cordova, V. D., El-Masri, B. Z., Baldocchi, D. D., Flanagan, L. B., Goldstein, A. H.,  
653 Hollinger, D. Y., Misson, L., Monson, R. K., Oechel, W. C., Schmid, H. P., Wofsy, S. C. and Xu, L.: On the use  
654 of MODIS EVI to assess gross primary productivity of North American ecosystems, *J. Geophys. Res.*  
655 *Biogeosciences*, 111(4), 1–16, doi:10.1029/2006JG000162, 2006.
- 656 Spracklen, D. V., Arnold, S. R. and Taylor, C. M.: Observations of increased tropical rainfall preceded by air  
657 passage over forests, *Nature*, 489(7415), 282–285, doi:10.1038/nature11390, 2012.
- 658 Staal, A., Tuinenburg, O. A., Bosmans, J. H. C., Holmgren, M., van Nes, E. H., Scheffer, M., Zemp, D. C. and  
659 Dekker, S. C.: Forest-rainfall cascades buffer against drought across the Amazon, *Nat. Clim. Chang.*, 1,  
660 doi:10.1038/s41558-018-0177-y, 2018a.
- 661 Staal, A., van Nes, E. H., Hantson, S., Holmgren, M., Dekker, S. C., Pueyo, S., Xu, C. and Scheffer, M.: Resilience  
662 of tropical tree cover: The roles of climate, fire, and herbivory, *Glob. Chang. Biol.*, (June), 5096–5109,  
663 doi:10.1111/gcb.14408, 2018b.
- 664 Swann, A. L. S., Longo, M., Knox, R. G., Lee, E. and Moorcroft, P. R.: Future deforestation in the Amazon and  
665 consequences for South American climate, *Agric. For. Meteorol.*, 214–215, 12–24,  
666 doi:10.1016/j.agrformet.2015.07.006, 2015.
- 667 Vergopolan, N. and Fisher, J. B.: The impact of deforestation on the hydrological cycle in Amazonia as observed  
668 from remote sensing, *Int. J. Remote Sens.*, 37(22), 5412–5430, doi:10.1080/01431161.2016.1232874, 2016.
- 669 Wan, Z.: New refinements and validation of the collection-6 MODIS land-surface temperature/emissivity product,  
670 *Remote Sens. Environ.*, 140, 36–45, doi:10.1016/j.rse.2013.08.027, 2014.
- 671 Wilcox, R. R.: Comparing two independent groups via a quantile generalization of the wilcoxon-mann-whitney  
672 test, *J. Mod. Appl. Stat. Methods*, 11(2), doi:10.22237/jmasm/1351742460, 2012.



673 Zalles, V., Hansen, M. C., Potapov, P. V., Stehman, S. V., Tyukavina, A., Pickens, A., Song, X., Adusei, B., Okpa,  
674 C., Aguilar, R., John, N. and Chavez, S.: Near doubling of Brazil's intensive row crop area since 2000  
675 [Sustainability Science], Proc. Natl. Acad. Sci. U. S. A., 116(2), doi:10.1073/pnas.1810301115, 2019.

676