

## Evapotranspiration seasonality across the Amazon basin

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**Abstract.** Evapotranspiration (ET) of Amazon forests is a main driver of regional climate patterns and an important indicator of ecosystem functioning. Despite its importance, the seasonal variability of ET over Amazon forests, and its relationship with environmental drivers, is still poorly understood. In this study, we carry out a water balance approach to analyze seasonal patterns in ET and their relationships with water and energy drivers over five sub-basins across the Amazon basin. We used *in-situ* measurements of river discharge, and remotely sensed estimates of terrestrial water storage, rainfall, and solar radiation. We show that the characteristics of ET seasonality in all sub-basins differ in timing and magnitude. The highest mean annual ET was found in the northern Rio Negro basin ( $\sim 1497$  mm year<sup>-1</sup>) and the lowest values in the Solimões River basin ( $\sim 986$  mm year<sup>-1</sup>). For the first time in a basin-scale study, using observational data, we show that factors limiting ET vary across climatic gradients in the Amazon, confirming local-scale eddy covariance studies. Both annual mean and seasonality in ET are driven by a combination of energy and water availability, as neither rainfall nor radiation alone could explain patterns in ET. In southern basins, despite seasonal rainfall deficits, deep root water uptake allows increasing rates of ET during the dry season, when radiation is usually higher than in the wet season. We demonstrate contrasting ET seasonality with satellite greenness across Amazon forests, with strong asynchronous relationships in ever-wet watersheds, and positive correlations observed in seasonally dry watersheds. Finally, we compared our results with estimates obtained by two ET models, and we conclude that neither of the two tested models could provide a consistent representation of ET seasonal patterns across the Amazon.

**Keywords:** Rainforest, water balance, hydrology, EVI.

37

## 1. Introduction

38 Evapotranspiration (ET) in the Amazon rainforest exerts large influences on regional and global  
39 climate patterns (Spracklen et al., 2012). Although exact figures vary, it is broadly known that the  
40 Amazon River basin transfers massive volumes of water from the land surface to the atmosphere every  
41 day, thereby having massive influence on the global energy budget (Aragão, 2012; Christoffersen et al.,  
42 2014; Hasler and Avissar, 2007; Restrepo-Coupe et al., 2016). ET is also an indicator of ecosystem  
43 functioning, given its intrinsic association with CO<sub>2</sub> fluxes during the transpiration process. Hence, any  
44 modification of ET over Amazon tropical forests would likely alter the global carbon cycle and further  
45 feedback to the rate of a changing climate.

46 Nonetheless, the spatial and temporal characteristics of ET across the Amazon basin, as well as the  
47 relative contribution of the multiple drivers to this process, are still uncertain. This may be attributed to  
48 the lack of high quality validation data over the full range of ecoregions across the basin, and the thus  
49 far unclear influence of climate on vegetation functioning. Recent studies suggested that vegetation  
50 phenology, as indicated by leaf demography (Lopes et al., 2016; Restrepo-Coupe et al., 2013; Wu et  
51 al., 2016), further increases the complexity of quantifying the relative importance of biotic and abiotic  
52 drivers of ecosystem functioning over the Amazon. These uncertainties are reflected in simulations by  
53 land surface models (LSMs) and global circulation models (GCMs), hindering the delineation of more  
54 reliable climate change scenarios (Karam and Bras, 2008; Restrepo-Coupe et al., 2013, 2016; Werth  
55 and Avissar, 2004).

56 Comprehensive assessments on ET have recently been carried out at local scales using eddy-covariance  
57 (EC) methods, which substantially contributed to the understanding of ET seasonality and its drivers in  
58 the Amazon (Christoffersen et al., 2014; Fisher et al., 2009; Hasler and Avissar, 2007). EC assessments  
59 are, however, limited to small areas. Due to the diversity of vegetation and climatic conditions across  
60 the Amazon basin, EC measurements cannot provide a broader overview of the spatial characteristics  
61 of ET across the region. The most comprehensive studies carried out so far are based on the data from  
62 five to seven flux towers (Christoffersen et al., 2014; Fisher et al., 2009), which although distributed in  
63 different ecoregions, cannot represent the full complexity of the Amazon basin. For instance, none of  
64 these towers is located in the western Amazon, or in the very wet Rio Negro basin. Furthermore, some  
65 sub-basins are characterized by a complex mosaic of land cover types and ecotones, making it  
66 impossible to describe the total ET based on unevenly distributed measurements.

67 Although hydrometeorological models have been implemented to provide spatially explicit assessments  
68 of ET in the Amazon, the poor understanding of drivers of ecosystem functioning hinder a more robust  
69 parameterization of models (Han et al., 2010). For instance, the spatio-temporal variation of ET is  
70 strongly linked to how vegetation assimilates available energy and water (Hasler and Avissar, 2007;  
71 Nepstad et al., 1994), a process which just recently started being elucidated (Restrepo-Coupe et al.,  
72 2013; Wu et al., 2016). Hence, generally ET models are shown to perform poorly in Amazon forest  
73 ecosystems (Karam and Bras, 2008; Restrepo-Coupe et al., 2016; Werth and Avissar, 2004).

74 Given these bottlenecks, a better understanding of ET seasonality, as well as its relationship with key  
75 climate forcings, are needed before model results can be reliably evaluated across the entire Amazon  
76 Basin. Water balance approaches are useful in these situations, as they do not necessarily rely on model

77 assumptions and calibration, and therefore can be applied when there is a lack of *in situ* ET data or  
78 when the drivers of the ET process are not fully understood.

79 ET assessments using water balance methods have also been undertaken in the Amazon basin, though  
80 generally these studies treated the Amazon basin as a whole (Karam and Bras, 2008; Ramillien et al.,  
81 2006; Werth and Avissar, 2004). Given the large scale of previous studies, assessments on the drivers  
82 of ET have in some cases been inconclusive (e.g. Werth and Avissar, 2004) or reached a single solution  
83 for the entire Amazon basin. For instance, Karam and Bras (2009) concluded that Amazonian ET is  
84 primarily limited by energy availability. These results provide important advances in our understanding  
85 of water and energy balance in the Amazon region, but more refined studies are necessary to resolve  
86 regional variations. Consequently, water balance assessments at smaller sub-basin scales are needed to  
87 evaluate ET limiting factors and their seasonality over a larger range of bioclimatic condition.

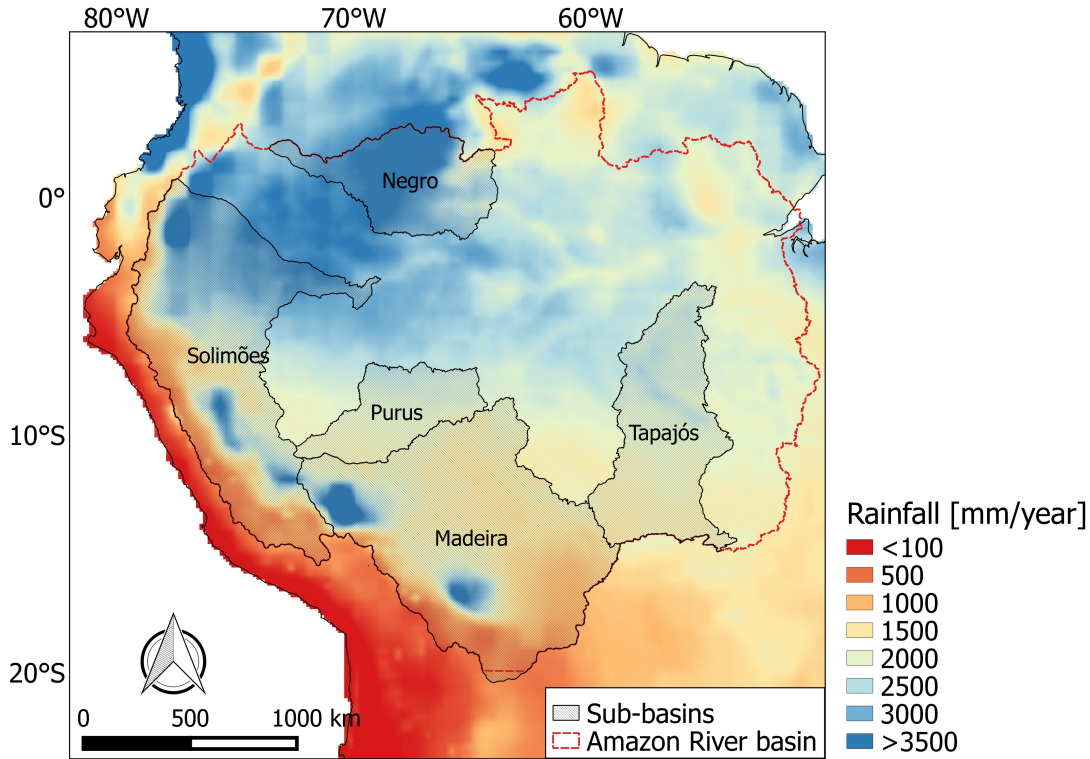
88 Given that plant transpiration is associated with CO<sub>2</sub> absorption through leaf stomata, ET is closely  
89 linked to ecosystem gross primary production (GPP). For this reason, remotely sensed proxies of  
90 photosynthetic activity, in particular vegetation indices (VIs), have often been incorporated into models  
91 of ET (e.g. Glenn et al., 2010; Yang et al., 2013). Assessing the relationships between ET and  
92 vegetation greenness measured by VIs can also lead to a better understanding of vegetation phenology  
93 determinants of ET and ecosystem functioning in general, fostering the improvement of model  
94 parameterization. However, studies have found contrasting results on the relationship between canopy  
95 greenness measured by VIs and GPP patterns in Amazon forests (Huete et al., 2006; Jones et al., 2014;  
96 Maeda et al., 2014). Recent assessments helped clarify this discrepancy, showing that in some parts of  
97 the Amazon GPP is driven by the synchronization of new leaf growth with dry season litterfall,  
98 increasing the proportion of younger and more light-use efficient leaves, highlighting the importance of  
99 leaf phenology (Wu et al., 2016).

100 The objective of this study was to utilize a water-balance approach to describe seasonal patterns of  
101 watershed scale ET across Amazon forests, and relate seasonal patterns with climatic drivers and  
102 vegetation greenness. The research questions addressed were: (1) How do seasonal patterns of ET vary  
103 across five sub-basins of the Amazon basin? (2) Are the environmental controls of ET similar among  
104 sub-basins and across time? (3) How does ET seasonality relate with greenness seasonality? Finally,  
105 we compare our ET results with those estimated by a LSM and remote sensing based ET retrievals.

106

## 107 **2. Material and methods**

108



109

110 **Figure 1.** Amazon River sub-basins assessed in this study. The background map shows the mean  
 111 annual rainfall 2001-2014, measured by the Tropical Rainfall Measuring Mission (TRMM). The  
 112 extents of five sub-basins analyzed here are indicated on the map with solid black lines and shading.  
 113 The solid red line indicates the boundary of the entire Amazon River basin.

114

115 **2.1. Evapotranspiration calculation using water-balance approach**

116 This analyses were carried out at the watershed level, considering the drainage area of the five major  
 117 rivers inside the Amazon basin: the Negro, Solimões, Purus, Madeira and Tapajós Rivers (Figure 1).  
 118 These basins are distributed within different ecoregions inside the Amazon basin. The size and number  
 119 of sub-basins were, however, limited by the availability of reliable river discharge data, which is a  
 120 critical element for the water balance calculation. The ET in each watershed was calculated using the  
 121 following water budget equation:

122 
$$ET = P - R - \frac{dS}{dT} \quad (1)$$

123 where ET is the monthly evapotranspiration, P is the monthly rainfall, R is the river discharge and  
 124  $dS/dT$  is the change in terrestrial water storage.

125 Monthly river discharge measurements were obtained from the Environmental Research Observatory  
 126 (ORE) HYBAM (Geodynamical, hydrological and biogeochemical control of erosion/alteration and  
 127 material transport in the Amazon basin). Changes in water storage ( $dS$ ) were calculated using Total  
 128 Water Storage Anomalies (TWSA) estimated from NASA's Gravity Recovery and Climate Experiment

129 (GRACE) satellites (Landerer and Swenson, 2012; Tapley et al., 2004) using the following equation  
130 (Swenson and Wahr, 2006):

$$131 \quad dS_n = (TWSA_{n+1} - TWSA_{n-1}) \quad (2)$$

132 where  $TWSA_{n-1}$  and  $TWSA_{n+1}$  are the TWSA values for the months preceding and succeeding month  $n$ ,  
133 respectively.

134 Three monthly GRACE solutions, from different processing centers, were used to compile monthly  
135 TWSA: the GFZ (GeoforschungsZentrum Potsdam), CSR (Center for Space Research at University of  
136 Texas, Austin), and JPL (Jet Propulsion Laboratory) (Landerer and Swenson, 2012). The three  
137 solutions were combined by simple arithmetic mean of the gravity fields, which according to recent  
138 studies is the most effective approach for reducing the noise in the gravity field solutions (Sakumura et  
139 al., 2014). Given that these products provide observations for the middle of each month, with varying  
140 dates, TWSA values were adjusted for the first day of each month using linear interpolation.

141 Rainfall data were obtained from the TRMM 3B43 V7 product. The 3B43 V7 product consists of  
142 monthly average precipitation rate ( $\text{mm hr}^{-1}$ ), at  $0.25^\circ \times 0.25^\circ$  spatial resolution, which combines the  
143 estimates generated by sensors on board of the TRMM, geostationary satellites and ground data  
144 (Huffman et al., 2007). The ground data were obtained from NOAA's Climate Anomaly Monitoring  
145 System (CAMS), and the global rain gauge product produced by the Global Precipitation Climatology  
146 Center (GPCC) (Huffman et al., 2007).

147 To facilitate the visualization of ET seasonal patterns, ET for each month was calculated using a three-  
148 month sliding window. Hence, the changes in water storage for a certain month were assessed by  
149 evaluating the changes in TWSA between the previous and following month (equation 2). The rainfall  
150 and river discharge were then calculated accordingly, providing the average volumes inside the  
151 averaged window period.

152

## 153 **2.2. Climate drivers of ET**

154 We evaluate the influence of energy and water input on ET seasonal patterns across all sub-basins.  
155 Monthly incident shortwave radiation flux data were obtained from CERES SYN1deg product, version  
156 3A (Kato et al., 2011). Shortwave radiation refers to radiant energy with wavelengths in the visible,  
157 near-ultraviolet, and near-infrared spectra. The SYN1deg product provides radiation variables  
158 calculated for all-sky, clear-sky, pristine (clear-sky without aerosols), and all-sky without aerosol  
159 conditions. In this study, we used the product made for all-sky. The incident radiation flux from  
160 SYN1deg product was shown to have a good relationship with photosynthetically active radiation  
161 (PAR) measured at flux towers in central Amazon (Maeda et al., 2014). Monthly rainfall values were  
162 obtained from the TRMM 3B43 product, as described in the previous section.

163 The influence of climate forcings on ET seasonal patterns was assessed using a modified Budyko  
164 analysis (Chen et al., 2013; Du et al., 2016). The original Budyko framework (Budyko, 1958) was  
165 created to describe the links between climate and catchment hydrological components, resulting in  
166 what is known as the "Budyko curve". In this framework, ET is limited by the supply of either water or

167 energy. The type and degree of limitation is determined by the dryness index, which is the ratio of  
168 potential ET (PET) to rainfall (P). The PET provides a proxy of the available energy, and represents the  
169 maximum possible value of evapotranspiration under given conditions. Hence, dryness indices lower  
170 than 1 represent energy-limited environments, while values higher than 1, water-limited (Budyko,  
171 1958; Donohue et al., 2007). Monthly PET estimates were obtained from the MODIS MOD16A2  
172 (collection 5) product (Mu et al., 2007). In MOD16 product, PET is calculated using the Penman-  
173 Monteith equation driven by surface and remote sensing derived input (Cleugh et al., 2007; Mu et al.,  
174 2007).

175 The other component of the Budyko framework is the evaporative index (ET/P), which describes the  
176 partitioning of P into ET and R. In this case, R is proportional to the distance between the curve and a  
177 water limit line (i.e. evaporative index=1) and sensible heat is proportional to the distance between the  
178 curve and an energy limit line (i.e. when evaporative index=dryness index) (Budyko, 1958; Donohue et  
179 al., 2007).

180 However, these approximations can only be used at steady-state conditions, assuming  $dS \sim 0$ . Hence, the  
181 original Budyko framework is usually recommended for annual or longer time-scales. For shorter time-  
182 scales, studies have shown that inter-annual water storage change should be considered to properly  
183 represent the ratio between ET and R (Wang et al., 2009; Zhang et al., 2008). The difference between  
184 rainfall and storage change was shown to be a good approach for representing effective precipitation in  
185 seasonal models (Chen et al., 2013; Du et al., 2016). Here, we follow this modified Budyko framework,  
186 in which the effective precipitation is represented by  $P-dS$ , so that the evaporative index is  $ET/(P-dS)$   
187 and the dryness index is  $PET/(P-dS)$ .

188

### 189 **2.3. Vegetation greenness proxy**

190 Seasonal patterns of vegetation greenness were assessed using the enhanced vegetation index (EVI)  
191 obtained from the Moderate Resolution Imaging Spectroradiometer (MODIS) (Huete et al., 2002). For  
192 this study we used the MODIS MAIAC product, which is processed using MODIS Collection 6 Level  
193 1B (calibrated and geometrically corrected) observations. MAIAC uses an adaptive time series analysis  
194 and processing of groups of pixels for advanced cloud detection, aerosol retrievals and atmospheric  
195 correction (Lyapustin et al., 2012). This dataset provides geometrically-normalized spectral  
196 reflectances (BRF<sub>n</sub>), which were used in this study. EVI was calculated considering a fixed sun-sensor  
197 geometry, with sun zenith angle of 45 degrees and nadir view angle. We used observations from the  
198 Terra and Aqua satellites collected between 2001 and 2012, and data were obtained from the  
199 Atmosphere Archive and Distribution System (LAADS Web: <ftp://ladsweb.nascom.nasa.gov/MAIAC>).

200

### 201 **2.4. Comparison with modelled ET**

202 We compare our ET estimates with two model-based estimates. The first modelled ET dataset was  
203 obtained from the NOAA 2.7.1 Land Surface Model (LSM) in the Global Land Data Assimilation  
204 System (GLDAS) (Rodell et al., 2004). The data have a 0.25° spatial resolution and the temporal  
205 resolution is monthly. The NOAA LSM comprises three components of latent heat: bare soil

206 evaporation, transpiration, and wet surface evaporation (Chen et al., 1996). These components are then  
 207 summed after constraints on PET have been computed (Mahrt and Ek, 1984). PET in the NOAH model  
 208 is estimated using a modified version of Penman (1948) (Mahrt and Ek, 1984).

209 The second modeled ET dataset was obtained from the MODIS MOD16A2 product (Mu et al., 2007).  
 210 The MOD16 ET is calculated by a modified Penman–Monteith ET method, which uses ground-based  
 211 meteorological observations and remote sensing data from MODIS to provide global estimates of ET.  
 212 For both modeled ET datasets, NOAH and MOD16, data were obtained from January 2001 to  
 213 December 2014.

214

### 215 3. Results

216

#### 217 3.1. Spatial and seasonal variations in ET across five Amazon sub-basins

218 A summary of the components used for the water balance equation (eq 1), for the period between 2001  
 219 and 2014, are presented in Table 1. The largest river discharge and rainfall volumes were observed in  
 220 the Rio Negro basin, with an annual mean of 1692 mm year<sup>-1</sup> and 3285 mm year<sup>-1</sup>, respectively. The  
 221 lowest values were observed in the Madeira River, where mean discharge was 584 mm year<sup>-1</sup> and mean  
 222 rainfall 1716 mm year<sup>-1</sup> (Table 1). Seasonal variations in total water storage are larger in the Tapajós  
 223 River basin, where the mean maximum was 132 mm month<sup>-1</sup> (i.e. increasing water storage) and mean  
 224 minimum was -123 mm month<sup>-1</sup> (i.e. decreasing water storage) (Table 1).

225

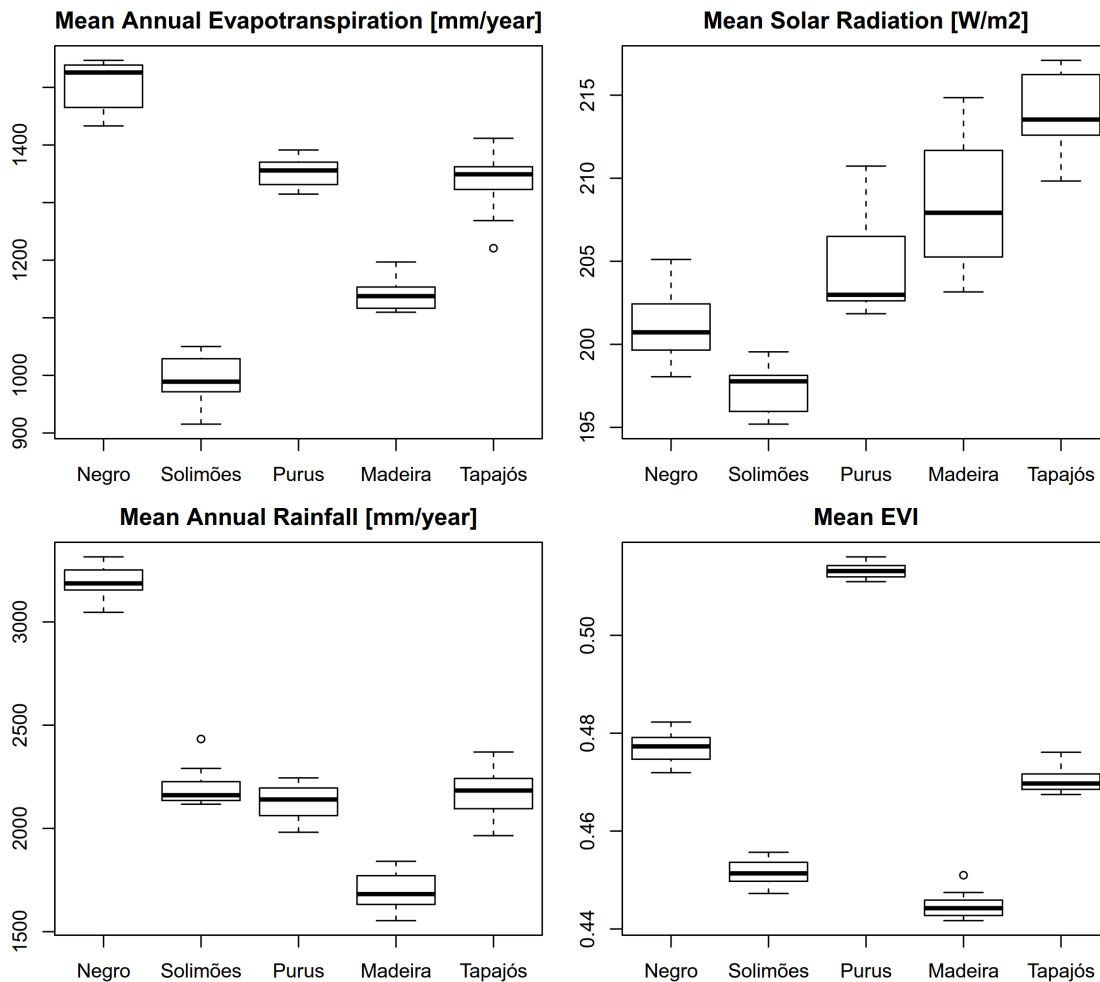
226 **Table 1.** Summary of the river discharge, rainfall and dS/dT in the five sub-basins analyzed in this  
 227 study. For each variable, the monthly average maximum and minimum, as well as the annual mean, are  
 228 presented. All values are averages for the period between 2001 and 2014. Long-term annual averages  
 229 of dS/dT are generally close to zero, and therefore not presented.

|               | Mean values (2001-2014)                   | Negro       | Solimões    | Purus       | Madeira     | Tapajós     |
|---------------|---|-------------|-------------|-------------|-------------|-------------|
| Discharge (R) | Monthly Max [mm month <sup>-1</sup> ]     | 213         | 138         | 123         | 84          | 117         |
|               | Monthly Min [mm month <sup>-1</sup> ]     | 96          | 63          | 15          | 12          | 24          |
|               | <b>Mean annual [mm year<sup>-1</sup>]</b> | <b>1692</b> | <b>1241</b> | <b>767</b>  | <b>584</b>  | <b>767</b>  |
| Rainfall (P)  | Monthly Max [mm month <sup>-1</sup> ]     | 360         | 234         | 294         | 252         | 327         |
|               | Monthly Min [mm month <sup>-1</sup> ]     | 213         | 123         | 45          | 39          | 21          |
|               | <b>Mean annual [mm year<sup>-1</sup>]</b> | <b>3285</b> | <b>2227</b> | <b>2154</b> | <b>1716</b> | <b>2154</b> |
| dS/dT         | Monthly Max [mm month <sup>-1</sup> ]     | 48          | 54          | 99          | 87          | 132         |
|               | Monthly Min [mm month <sup>-1</sup> ]     | -45         | -72         | -96         | -75         | -123        |
| ET            | Monthly Max [mm month <sup>-1</sup> ]     | 132         | 105         | 138         | 114         | 123         |
|               | Monthly Min [mm month <sup>-1</sup> ]     | 108         | 63          | 90          | 78          | 99          |
|               | <b>Mean annual [mm year<sup>-1</sup>]</b> | <b>1497</b> | <b>986</b>  | <b>1351</b> | <b>1132</b> | <b>1314</b> |

230

231 Annual mean ET values varied among five sub-basins (Table 1; Figure 2). The largest mean annual ET  
 232 was observed in the Rio Negro basin (~1497 mm year<sup>-1</sup>), while the lowest value was observed in the

233 Solimões River basin ( $\sim 986 \text{ mm year}^{-1}$ ) (Table 1; Figure 2). The relative magnitude of mean ET among  
 234 the Negro, Purus, Madeira and Tapajós basins are consistent with rainfall variation within these  
 235 regions, i.e., the highest mean annual ET corresponds to the highest mean annual rainfall, and vice  
 236 versa (Figure 2). The Solimões basin, however, is an exception. Despite having annual average rainfall  
 237 similar to what was observed in Purus, its mean ET rates were significantly smaller (Figure 2). This  
 238 may be explained by the lower average solar radiation inside the Solimões basin, with an annual  
 239 average of  $197 \text{ W m}^{-2}$ , while the average in the Purus basin was  $204 \text{ W m}^{-2}$  (Figure 2). Furthermore,  
 240 portions of the Solimões basin are located in the Andes region, which is characterized by higher  
 241 altitudes, lower rainfall and sparse vegetation (Figure 1).



242  
 243 **Figure 2.** Boxplots with mean annual evapotranspiration, solar radiation, rainfall and EVI for the five  
 244 sub-basins analyzed in the study for the period 2001 – 2014 inclusive.

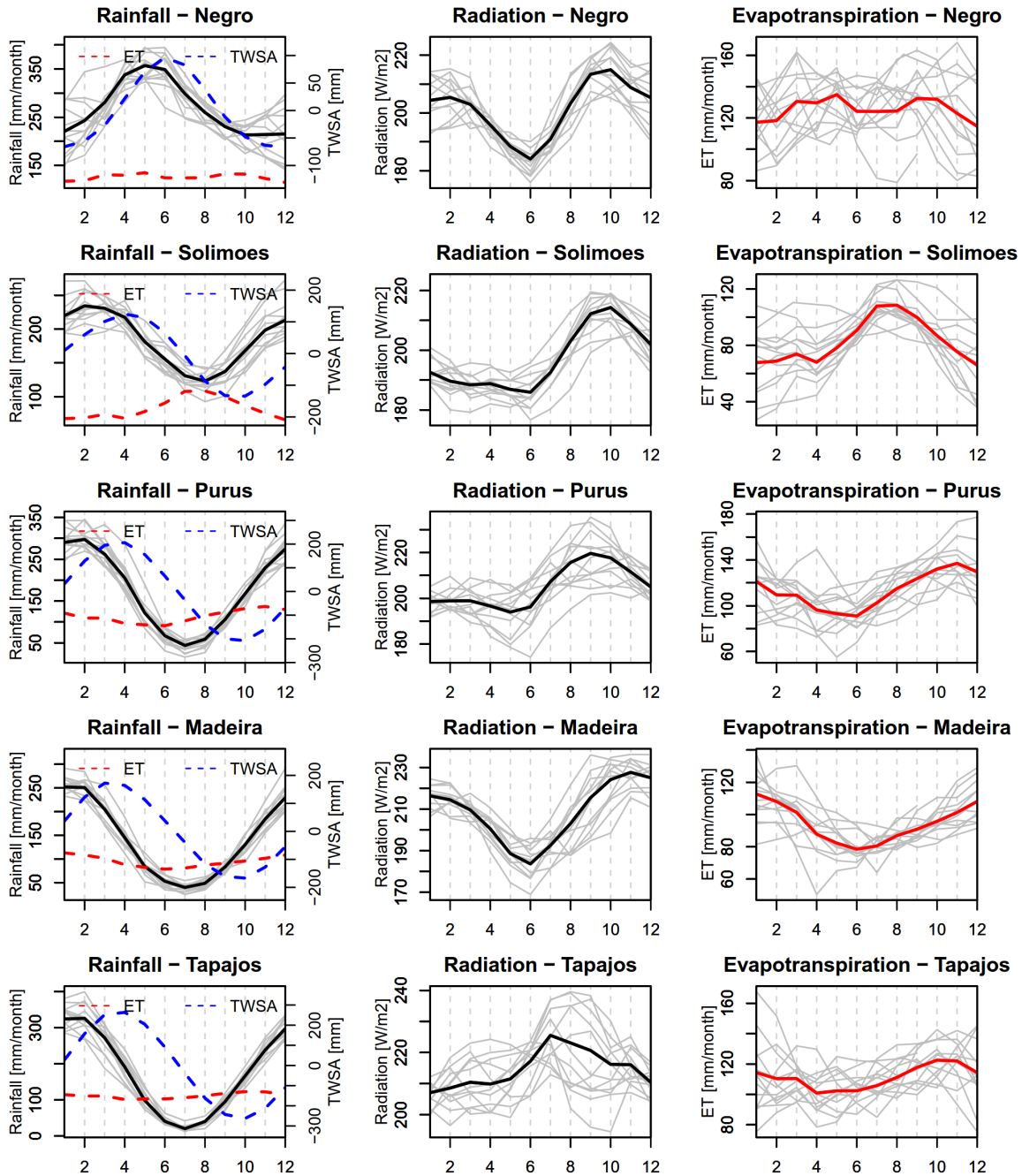
245  
 246 The seasonal patterns of rainfall, radiation and ET are presented in Figure 3. Seasonal variation of ET  
 247 is clearly observed in Solimões, Purus, Madeira and Tapajós, but less evident in the Rio Negro basin.  
 248 In the Solimões basin, ET was highest in September and October, while the lowest values were



249 observed in December and January (Fig. 3). In the Purus, Madeira and Tapajós basins, ET peaks  
250 around November, February and November, respectively (Fig. 3).

251 In terms of long-term average values, ET did not exceed rainfall in any season of the year, in the Negro  
252 and Solimões basin sites. This indicates that, under average conditions, ET is not limited by water  
253 availability, even in the driest season. In the Purus, Madeira and Tapajós sites, rainfall deficit (i.e.  
254  $ET > \text{rainfall}$ ) was observed between June and August. Water availability is, therefore, a limitation for  
255 ET during the dry season. In fact, in these three basins, the smallest rate of ET was observed in May-  
256 June, period in which rainfall volumes are in steady decline.

257



258

259 **Figure 3.** Seasonal variations of rainfall, radiation and evapotranspiration inside each sub-basin. Gray  
 260 lines represent the values for each year from 2002-2014, and solid dark lines represent the average  
 261 values for each month. Months are represented from 1 (January) to 12 (December). The dashed blue  
 262 line in the first column shows the mean seasonal variation of GRACE terrestrial water storage  
 263 anomalies (TWSA), and the dashed red line is the mean seasonal variation of water-balance ET, for  
 264 each sub-basin.

265

266 **3.2. Climatic drivers of Amazon ET seasonality**

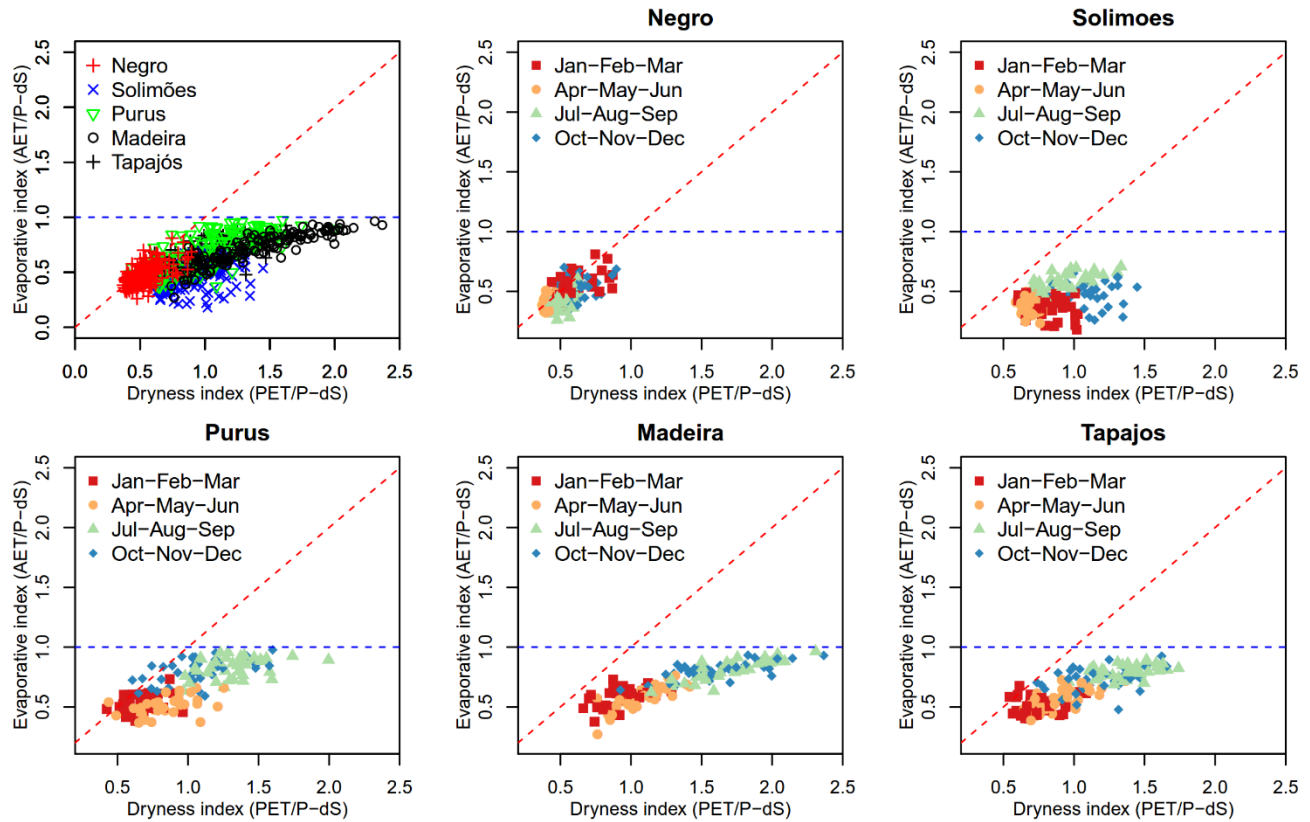
267 The modified Budyko analysis of monthly ET values are presented in Figure 4. The dryness index in  
268 the Negro basin was consistently below the water limit threshold ( $<1$ ). For this sub-basin, the water  
269 balance analyses show the basin to consistently follow the energy limited line (red dashed line),  
270 indicating some degree of energy limitation. However, our results show small seasonal variation of ET  
271 in the Negro basin, despite clear intra-annual variation in solar radiation (mean annual amplitude of  $30$   
272  $\text{W}\cdot\text{m}^{-2}$ ) and rainfall (mean annual amplitude of  $140 \text{ mm}\cdot\text{month}^{-1}$ ). These contrasting results are likely  
273 explained by the very high ET rates at the Negro basin (Table 1), which could represent an upper limit  
274 in forest water use capacity.

275 In the three southern basins, Purus, Madeira and Tapajós, water limitation was consistently observed  
276 during July, August and September (Figure 4). This is consistent with the observation of seasonal  
277 rainfall deficits in these regions, but it contrasts with the ET seasonal patterns in these basins (Figure  
278 3). In all southern basins, ET reached the lowest values before the period of minimum rainfall. These  
279 results suggest that in the southern Amazon ecotone, deep root water intake plays a key role in  
280 maintaining ecosystem productivity during the dry season. In the Purus and Tapajós basins, the Budyko  
281 curves are particularly close to the energy limit threshold during January, February and March. This  
282 shows that ET in these regions can experience some degree of energy limitation during the wet season.

283 The Solimões basin is shown to be located in a transition region, where water limitation can occur in  
284 drier years. The energy constraint in the Solimões basin was also lower than that observed in the Negro  
285 basin. Given these characteristics, the Solimões basin is the only site where ET was shown to maximize  
286 the use of both solar radiation and water. In other words, ET reaches its peak when the ratio between  
287 radiation and rainfall is maximum (Figure 5).

288

289



290

291 **Figure 4.** Modified Budyko analysis for monthly water balance values. The red dashed line represents  
 292 the energy limitation threshold, above which ET is limited by solar radiation. The blue dashed line  
 293 represents the water limitation threshold.

294

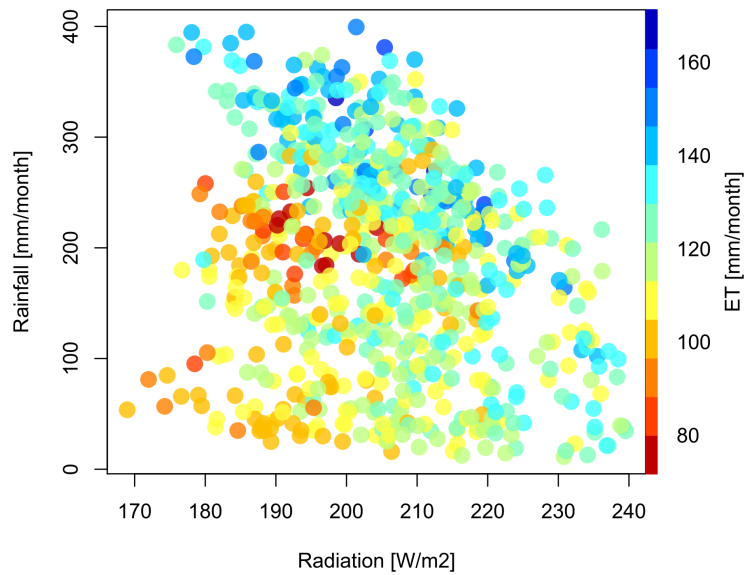
295 Figure 6 shows a scatterplot of monthly radiation *versus* rainfall, with data points labeled by their  
 296 corresponding monthly average ET values. This figure reveals a general pattern on the relationships  
 297 among monthly rainfall, radiation and ET. As expected, lower monthly ET values are consistently  
 298 observed when both radiation and rainfall are low. Interestingly, the highest ET values are not observed  
 299 when radiation was highest, providing more evidence that water availability is also a limiting factor of  
 300 ET, in combination with radiation.



301

302 **Figure 5.** Monthly values of the ratio between solar radiation [ $\text{W m}^{-2}$ ] and rainfall [ $\text{mm month}^{-1}$ ]  
 303 (Radiation/Rain) (solid black and gray lines), and mean seasonal variation in evapotranspiration (ET)  
 304 (dashed red line) at the Solimões river basin.

305



306

307 **Figure 6.** Scatterplot of monthly radiation and rainfall for the five sub-basins. Colour gradient indicates  
 308 the monthly ET value, from high (blue) to low (red).

309

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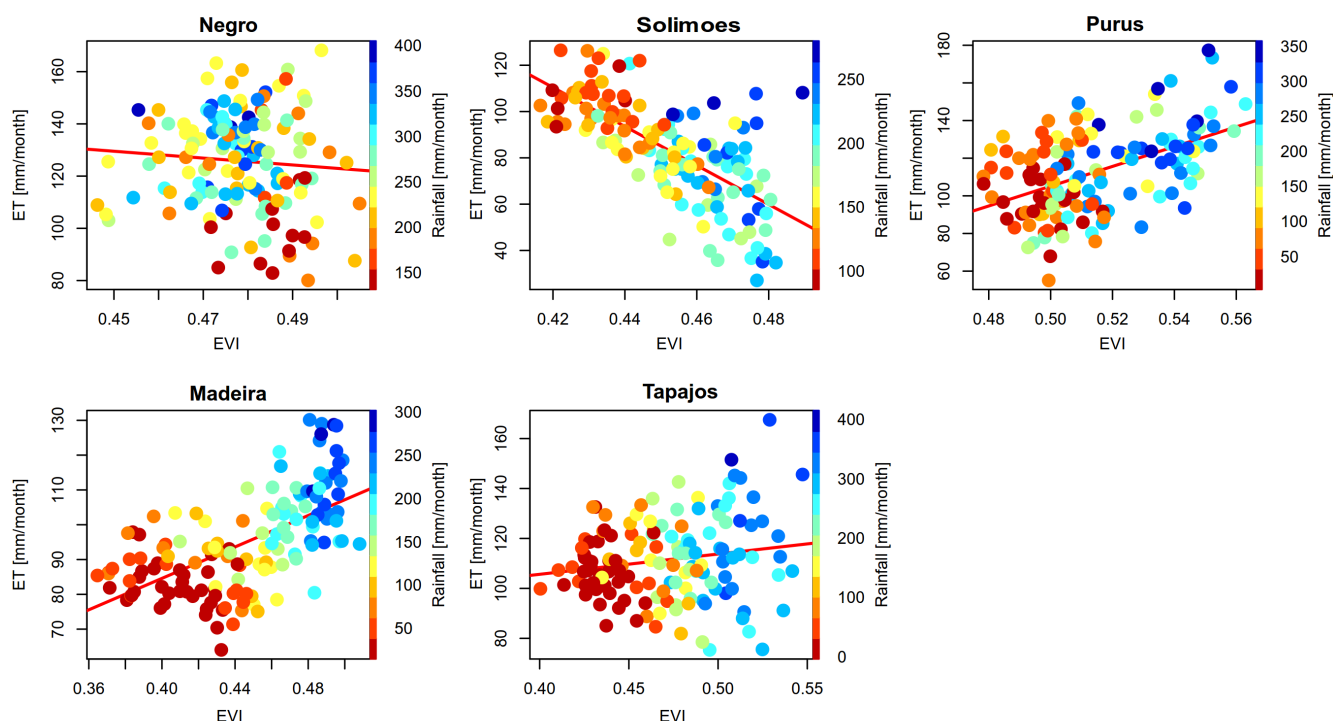
311

### 3.3. Relationship between ET and canopy greenness

312 The relationship between ET and vegetation greenness varied across the Amazon basin (Figure 7 and  
313 Table 2). In the Negro basin, no significant relationship was found between EVI and ET. In this region,  
314 vegetation greening was observed between September and December, followed by a steady decline in  
315 EVI until the following August (Figure 8).

316 Significant positive correlations ( $p < 0.05$ ) between EVI and ET were observed in the Purus, Madeira  
317 and Tapajós basins (Figure 7 and Table 2). In these regions, a clear pattern was observed, in which  
318 higher ET takes place when vegetation is greener and when rainfall is higher. In the Solimões basin,  
319 despite higher EVI values observed during the wet season (Figure 7), an opposite pattern between ET  
320 and EVI was observed, i.e. higher ET takes place when EVI is lower. In Solimões, vegetation greening  
321 also occurs between September and December, with declining from January until August (Figure 8).

322



323

324 **Figure 7.** Relationship between monthly evapotranspiration (ET) and MODIS enhanced vegetation  
325 index (EVI) at each Amazon sub-basin using the data from 2001 to 2014. Colour gradient indicates the  
326 monthly rainfall value, from high (blue) to low (red).

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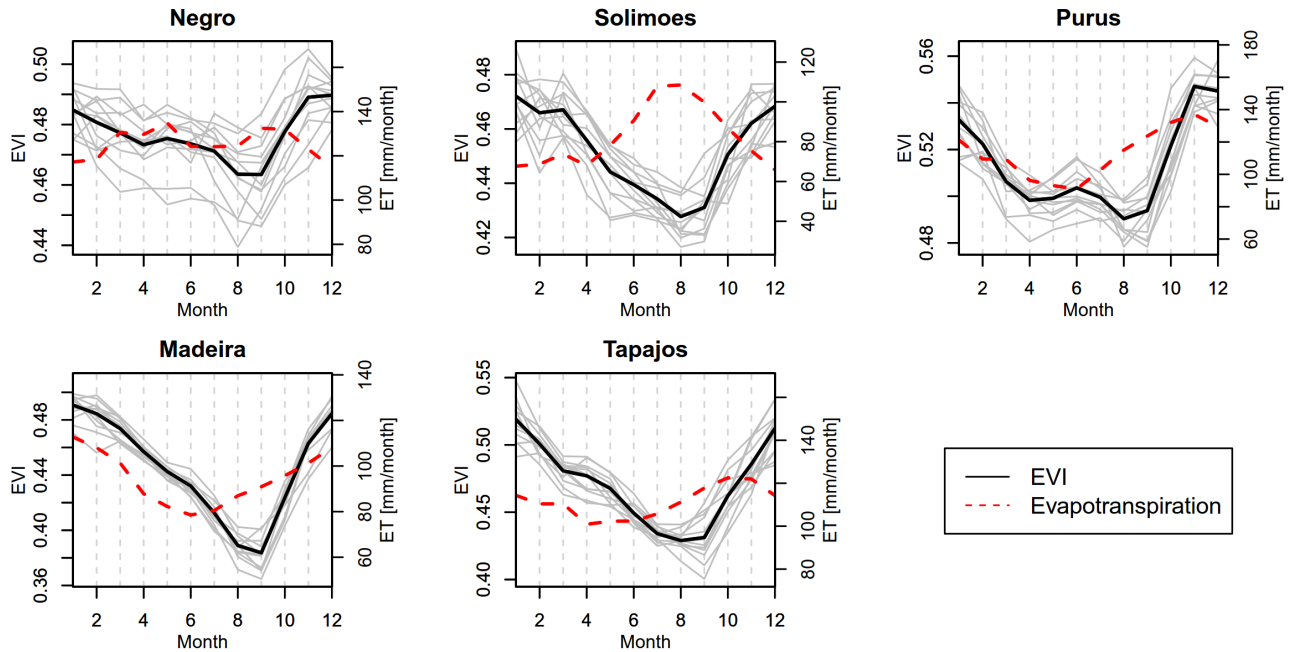
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329

330 **Table 2.** Coefficients of the linear regression between evapotranspiration (ET) and MODIS enhanced  
331 vegetation index (EVI) for each of the five sub-basins (\*  $p < 0.05$ ).

332

|                 | <i>Intercept</i> | <i>Slope</i> | <i>R</i> <sup>2</sup> |
|-----------------|------------------|--------------|-----------------------|
| <i>Negro</i>    | 6.0              | -4.06        | 0.006                 |
| <i>Solimões</i> | 14.9             | -27.0        | 0.463*                |
| <i>Purus</i>    | -5.3             | 17.5         | 0.259*                |
| <i>Madeira</i>  | -0.4             | 7.9          | 0.383*                |
| <i>Tapajós</i>  | 2.2              | 3.1          | 0.035*                |



333

334 **Figure 8.** Seasonal patterns of MODIS EVI in the five Amazon sub-basins. The black lines show the  
 335 monthly average values from 2001 to 2014, while gray lines show individual monthly values for each  
 336 year. The mean seasonal variations in ET for each sub-basin are represented as red dashed lines.

337

338 **3.4. Comparison with ET estimated by models**

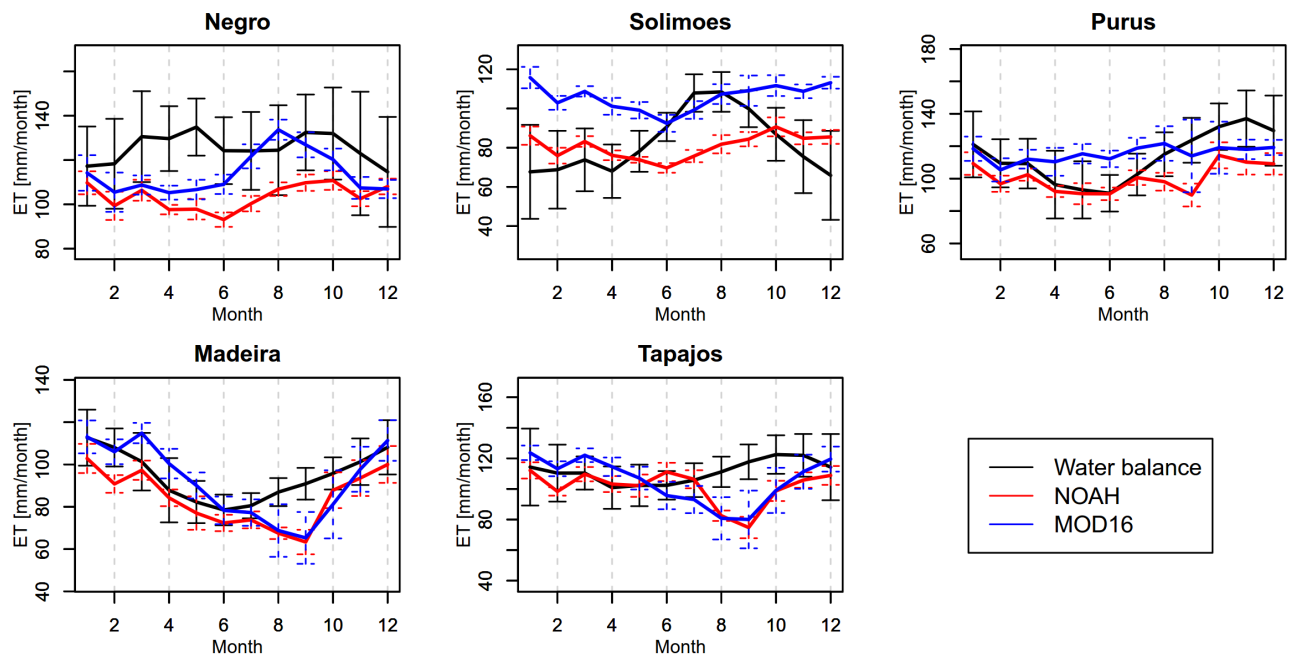
339 We further assessed the ability of two ET models, NOAH-LSM and MOD16 P-M, to replicate the  
 340 seasonality of ET as derived from observation-based water balance calculation. Our results showed that  
 341 neither of these two models was able to reproduce the timing and magnitude of seasonal ET patterns as  
 342 calculated from the water-balance approach (Figure 9). In the Negro basin, NOAH-LSM estimates  
 343 were consistently below the water balance and MOD16 P-M values, with an annual average of 1241  
 344 mm year<sup>-1</sup>. In this region, both NOAH-LSM and MOD16 P-M show a decreasing ET trend from  
 345 January to May, followed by an increasing trend (Figure 9). NOAH-LSM ET reached its maximum in  
 346 September, while MOD16 P-M ET maximum was observed in October (Figure 9).

347 In the Solimões basin, NOAH-LSM and MOD16 P-M ET showed similar seasonal patterns, but  
 348 MOD16 P-M ET values were on average 25 mm month<sup>-1</sup> larger than the NOAH-LSM estimates  
 349 throughout the year (Figure 9). Nonetheless, both models showed ET seasonal patterns largely

350 discrepant with the water balance calculation. Both models indicate highest ET in December/January,  
351 when the water balance showed the lowest seasonal values (Figure 9).

352 The MOD16 P-M ET showed almost no seasonality in the Purus basin, while NOAH-LSM and water  
353 balance ET indicate a decrease in ET during May (Figure 9). However, the NOAH-LSM  
354 underestimated the ET recovery in the following months, in particular between August and November  
355 (Figure 9). The same pattern was observed in the Madeira and Tapajós basins, where both models show  
356 significantly lower ET values in August, September and October (Figure 9).

357



358

359 **Figure 9.** Seasonal ET patterns obtained using the water balance method (black line), NOAH land  
360 surface model (red) and MODIS MOD16 P-M model (blue). Vertical bars indicate the  $\pm 1$  standard  
361 deviation of monthly observations from 2001 to 2014.

362

363

364

#### 4. Discussion

365 Previous estimates of ET in the Amazon basin vary considerably in terms of magnitude and seasonal  
366 patterns. Water balance assessments undertaken at larger scales (e.g. the entire Amazon basin) found  
367 mean annual ET estimates varying from  $767 \text{ mm year}^{-1}$  to  $1642 \text{ mm year}^{-1}$  (Callede et al., 2002; Karam  
368 and Bras, 2008; Ramillien et al., 2006; Rao et al., 1996; Werth and Avissar, 2004). The ET values we  
369 describe for Amazon sub-basins are within this range. We show that in some wet regions, such as the  
370 Rio Negro basin, mean annual ET can be above  $1400 \text{ mm year}^{-1}$ , while in southern basins it vary from  
371  $1130 \text{ mm year}^{-1}$  to  $1350 \text{ mm year}^{-1}$ . Hence, we find that the lower range of  $767 \text{ mm year}^{-1}$  described in  
372 previous studies (Karam and Bras, 2008) is likely to underestimate the average ET for the entire  
373 Amazon basin.



374 Our results show that the seasonal patterns of ET of five sub-basins across the Amazon vary in timing  
375 and magnitude. This spatial heterogeneity in ET seasonality is in agreement with previous studies  
376 carried out at local scale using EC method (Christoffersen et al., 2014; Fisher et al., 2009).  
377 Christoffersen et al. (2014) reported either a flat seasonal cycle or a slight dry season decrease of ET at  
378 transitional southern forests, while equatorial forest ET showed ET peaking with net radiation during  
379 the dry season. Despite agreeing on the main climatic forcing of ET process across these different  
380 ecoregions, our results unveil some differences on the timing of seasonal increases in ET and peak in  
381 relation to climatic variables. These differences are discussed in detail below.

#### 382 *4.1. Climatic drivers of Amazon ET seasonality*

383 Discussions on the drivers of ecosystem function seasonality in the Amazon have often resulted in  
384 conflicting results. Our results revealed that in most cases ET seasonality is driven by a balance  
385 between radiation, rainfall and vegetation regulations, rather than being exclusively limited by any one  
386 of these factors. For instance, the peak timing of ET at five sub-basins did not correspond to the peak  
387 timing of either rainfall or radiation, demonstrating that the arbitrary partition of the Amazon basin into  
388 either energy-limited or water-limited is unrealistic and would result in large uncertainty in predicted  
389 ET patterns, as we showed in this study.

390 We further demonstrated the degree of radiation and rainfall limitation, as well as their interactive  
391 effects on ET based on a modified Budyko analysis (Fig. 4-6). Our results show that the evaporative  
392 index ( $ET / (P - ds)$ ) exhibited a positive, nonlinear-type, dependency on climatic dryness index ( $PET /$   
393  $(P - ds)$ ), which falls well within the modified Budyko framework. The modification of the classic  
394 Budyko model is the consideration of temporal changes in water-storage, in which total water-  
395 availability for evaporation should be quantified as the sum of monthly precipitation and water-storage  
396 change, termed as effective precipitation. Our results thus revealed the importance of considering plant  
397 controls in water-balance accounting over Amazon basin forests, as these evergreen trees, with their  
398 lengthy root-systems, have the ability to tap deep soil-/ground-water to meet atmospheric water  
399 demand.

400 ET in the Solimões basin does not necessarily peak with solar radiation, but reaches a maximum when  
401 the ratio between radiation and rainfall is highest (Figure 5). In this case, where ET is normally not  
402 limited by water or energy input, plants do not need to regulate water loss, and seasonality of  
403 productivity can be regulated to reach an optimization that maximize the use of both available water  
404 and energy resources. In the Purus, Tapajós and Madeira basins, which encompass regions often  
405 considered to be water limited (Guan et al., 2015; Jones et al., 2014; Xu et al., 2015), ET does not  
406 necessarily reach the lowest values during the driest periods (Figure 3). Instead, we found increased ET  
407 before the end of the dry season, and ET rates can increase even in rainfall deficit conditions (Figure 4).  
408 This pattern can be explained by plants access to deep soil water (Nepstad et al., 1994). This argument  
409 is reinforced by the seasonal patterns of TWS demonstrated in Figure 3, which show that in southern  
410 basins TWS lags rainfall by approximately three months. Hence, during the meteorological dry season  
411 (i.e. when rainfall is low), soil water storage still remains relatively high. When the soils reach their  
412 lower storage volumes, 3 months after the peak of dry season, the rainy season has already started,  
413 providing water supply to be used by plants.

414 These results concur with previous findings showing a weak relationship between rainfall anomalies  
415 and EVI anomalies (Maeda et al., 2015), indicating a lower sensitivity of ecosystem functioning to  
416 rainfall extremes at transition forests in the southern Amazon. Furthermore, we show that besides  
417 dealing with seasonal rainfall deficit, southern basins remain limited by radiation energy availability  
418 during a certain period of the year (Figure 4), which explains the ET recovery before the driest period,  
419 i.e. when radiation starts to increase (Figure 3).

420 However, it is important to highlight the fact that, although these analyses are based on sub-basins  
421 across the Amazon, they still enclose relatively large areas with substantial heterogeneities. In  
422 particular, the Madeira and Tapajós basins are characterized by a large latitudinal gradient and,  
423 consequently, different ecosystems are present within these sub-basins. Hence, it is likely that, although  
424 on average the Tapajós and Madeira basins are limited by water availability during the dry season,  
425 water limitation may not occur in northern (wetter) parts of these basins.

426

#### 427 ***4.2. Relationship between ET and canopy greenness***

428 The biophysical causes of EVI seasonality in Amazon evergreen forests have been intensively  
429 discussed in recent years (Bi et al., 2015; Hilker et al., 2015; Maeda et al., 2014; Morton et al., 2014;  
430 Myneni et al., 2007). Recent studies indicate that in wet equatorial forests, EVI is driven by a net  
431 increase in leaf production (Lopes et al., 2016). The seasonal variation in EVI was shown to be more  
432 evident in the dry season, when most plants release old leaves while simultaneously producing new  
433 leaves and, therefore, increase EVI.

434 Furthermore, studies have shown that southern and Equatorial forests have different cues for leaf  
435 flushing, i.e. plant growing season is initiated by different climatic factors (Wagner et al., 2016).  
436 Hence, our results indicate a decoupling between ET fluxes and seasonal cycles of canopy foliage. In  
437 general, relationships were better in southern basins where rainfall deficits were observed, in particular  
438 Purus and Madeira. In these cases, the climatic triggers for leaf flushing/litter and productivity drivers  
439 are likely to be in phase. In the southern Amazon, leaf growth was shown to be initiated by water input  
440 (Wagner et al., 2016), which means that peak greening should be observed some months after the  
441 beginning of the wet season. In these regions, ET was found to decline as rainfall decreased between  
442 March and May. Nonetheless, ET trends recovered before the peak of the dry season, increasing with  
443 higher solar radiation – suggesting that soil water was available to the trees even during the peak of the  
444 dry season.

445 In the Negro basin, ET was not significantly correlated with EVI, while in the Solimões Basin, ET and  
446 EVI were inversely related. In these cases, different mechanisms are likely to drive ET and canopy  
447 greenness patterns. In the wet equatorial forests, leaf flushing was shown to be initiated by the increase  
448 in solar radiation (Lopes et al., 2016; Wagner et al., 2016). The subsequent decrease in greening,  
449 however, follows a different pattern, where a slow decrease in EVI might be associated with leaves  
450 aging, epiphylls, herbivores, and leaf fall.

451 Lags between forest functioning and canopy greening have been previously reported from local scale  
452 experiments. Wu et al (2016) suggested that these discrepancies could be explained by leaf

453 demography, given a higher photosynthetic capacity of mature leaves. In other words, while LAI  
454 increases during the dry season due to new leaves flushing, young leaves have lower photosynthetic  
455 capacity, which gradually increases as leaves become mature – but then declines as leaves senesce (Wu  
456 et al., 2016). They, hence conclude that phenology of photosynthetic capacity, and not climate  
457 variability, is the main driver of ecosystem productivity (Wu et al., 2016). Our results confirm this  
458 decoupling of vegetation functioning and leaf production in wet evergreen forests. Nonetheless, we  
459 demonstrate that vegetation function seasonality, as described by sub-basin scale ET, is not  
460 independent from climate intra-annual variability. In fact, in some regions, such as the Solimões basin,  
461 vegetation seems to maximize ET (hence productivity) by balance the use of available light and water  
462 resources across time.

463

#### 464 *4.3. Uncertainties of the water-balance approach and comparison with model estimates*

465 Assessing uncertainties of ET estimates in Amazon forests is challenging, given the lack of reference  
466 datasets. Previous studies indicate that ET estimates based on GRACE water balance approach may  
467 have higher uncertainties than LSM estimates (Long, 2014). This assessment was, however, carried out  
468 in a region with good data quality for model parameterization, and where the drivers of ecosystem  
469 functioning are better understood. In the Amazon, where parameterization of models are usually more  
470 challenging due to low data quality and unknown biophysical parameters, water balance methods are  
471 still considered an adequate alternative.

472 Assessing ET at local scales, using eddy covariance methods, Christoffersen et al. (2014) concluded  
473 that most models are not able to represent ET seasonality at different locations across the Amazon.  
474 They argue that models are unable to properly represent canopy dynamics mediated by leaf phenology,  
475 which is believed to play a significant role in regulating ET seasonality. Assessing spatially averaged  
476 ET for the Amazon basin, Karam and Bras (2008) reported that mean annual values calculated using  
477 water balance methods (including Callede et al., 2002; Ramillien et al., 2006) show significantly lower  
478 estimates when compared with output from LSMs. Although the models compared in this study are not  
479 the same, our results diverge from these claims. At the Negro, Purus, Madeira and Tapajós basins,  
480 mean annual ET values calculated with the water balance method were higher than NOAH and  
481 MOD16 estimates. Only at the Solimões basin, annual mean ET from MOD16 was higher than the  
482 other methods.

483 ET estimates from NOAH-LSM and MOD16 P-M could not provide a consistent representation of ET  
484 seasonality between each other in all sub-basins (Figure 9). Although a full comparison with ET  
485 models is beyond the scope of this study, our results confirm that models still disagree with each other  
486 in estimating Amazon ET seasonality, indicating uncertainties associated with either input datasets or  
487 model assumptions. Both models seem to overestimate water stress in the southern basins, i.e. while  
488 models predict a decline in ET after the driest period, the water balance estimate shows an early  
489 recovery from the dry season, followed by a steady increase until the end of wet season (Figure 9).

490 One potential source of uncertainty in the NOAH-LSM estimates is the fractional total vegetation cover  
491 ( $f_c$ ), which contributes for defining both transpiration and wet surface evaporation. In NOAH,  $f_c$   
492 seasonal variation is estimated from Normalized Difference Vegetation Index (NDVI) climatology  
493 obtained by the Advanced Very High Resolution Radiometer (AVHRR) (Gutman and Ignatov, 1998;

494 Marshall et al., 2013). Nonetheless, studies have shown that, due to saturation over dense tropical  
495 forests, as well as illumination artefacts, NDVI may not correctly describe seasonal changes in  
496 vegetation structure over the Amazon forests (Huete et al., 2002; Maeda et al., 2016).

497 The PET estimates used for the modified Budyko analysis (Figure 4) is also based on models, and  
498 therefor is likely to carry some level of uncertainty. Given that PET is a physical measure of  
499 atmospheric water demand, and do not depend on vegetation interactions, the reliability of estimates for  
500 the Amazon basin are likely to be the same as for other regions. Having said that, uncertainties in PET  
501 and ET have noticeable effects on the derived Budyko curves. For instance, underestimated PET values  
502 may lead to dryness index values higher than evaporative index, leading to plotted values that exceed  
503 the energy limit line. Previous studies, however, reported that monthly-average evaporation may  
504 exceed potential estimates by about 10 % during wet months (Shuttleworth, 1988). On the other hand,  
505 overestimated PET can lead to misleading conclusions of higher water limitation in Figure 4. This is  
506 likely to be the case in the Solimões basin, as the seasonal patterns presented in Figure 3, which are  
507 based only on observational data, indicate that in the Solimões basin average rainfall is always higher  
508 than average ET. Water limitation conditions in this region are still likely, given inter-annual variability  
509 in rainfall and ET, but it should not be a condition that is repeated consistently every year.

510

## 511 **Conclusions**

512 Our results demonstrate strong spatial heterogeneity in ET across five ecoregions within the Amazon  
513 basin. Seasonal cycles of ET are shown to vary in timing and magnitude, driven by intra-annual climate  
514 variability across sub-basins. Based on a modified Budyko analysis, we show the interactive effects of  
515 rainfall, solar radiation and soil water storage on ET fluxes. Nonetheless, our results indicate that  
516 neither energy or water input alone is sufficient to explain ET seasonality across five sub-basins,  
517 regardless of the average degree of dryness, demonstrating a dynamic shift in the degree of energy-  
518 /water-limitation across space and time. Although eddy covariance studies have shown that ET in the  
519 Amazon can be limited by different climatic factors, this fact had not yet been verified at basin scales  
520 using observational data.

521 We demonstrate a decoupling between ET and vegetation greenness seasonal patterns in wet  
522 Amazonian forests. In the Solimões basin, ET is inversely correlated with EVI, indicating higher ET  
523 when canopy foliage density is lower. This finding indicates that ecosystem models based on remotely  
524 sensed vegetation indices, including remote sensing based ET models, need to be further assessed to  
525 better represent ecosystem function seasonality in wet tropical forests.

526 A comparison with two ET models, NOAH-LSM and MOD16 P-M, showed that models are still  
527 unable to consistently represent ET seasonal patterns in the Amazon forest. In the Solimões and Negro  
528 basins, both models presented a different seasonal pattern when compared with our water balance  
529 approach. In southern basins, where rainfall is lower, models seem to overestimate water limitation  
530 during the dry season, and therefore underestimate ET.

531

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540

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