



1 Sensitivity of tropical woodland savannas to El Niño droughts

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18 Abstract

19 The 2015-2016 El Niño event led to one of the most intense and hottest droughts for many tropical 20 forests, profoundly impacting forest productivity. However, we know little about how this event 21 affected the Cerrado, the largest savanna in South America. Here we report 5 years of productivity 22 of the dominant vegetation types in Cerrado, savanna (cerrado) and transitional forest-savanna 23 (cerradão), continuously tracked before, during, and after the El Niño. We carried out intensive 24 monitoring between 2014 and 2019 of the productivity of key vegetation components (stems, 25 leaves, roots). Before the El Niño total productivity was ~25% higher in the cerradão compared 26 to the cerrado. However, cerradão productivity declined strongly by 29% during the El Niño event. 27 The most impacted component was stem productivity, reducing by 58%. By contrast, cerrado 28 productivity varied little over the years, and while the most affected component was fine roots, 29 declining by 38% during the event, fine root productivity recovered soon after the El Niño. The 30 two vegetation types also showed contrasting patterns in the allocation of productivity to canopy, 31 wood, and fine-root production. Our findings demonstrate that cerradão can show low resistance 32 and resilience to climatic disturbances due to the slow recovery of productivity. This suggests that 33 the transitional Amazon-Cerrado ecosystems between South America's largest biomes may be 34 particularly vulnerable to drought enhanced by climate change. 35 Keywords: 2015-2016 El Niño, productivity, productivity allocation, climate events, cerradão,

- 36 cerrado.
- 37

38 1 Introduction

The 2015-2016 El Niño event led to one of the most intense droughts of the last century as well as record maximum temperatures, coming on top of decades of long-term warming (Jiménez-





Muñoz et al., 2016; Liu et al., 2017). The 2015-16 climate anomaly affected most of the tropics 41 42 but was especially potent in Amazonia (Gloor et al., 2018). Intense droughts can increase tree 43 mortality and affect the carbon sequestration capacity of forests as shown by long-term ground-44 based monitoring (e.g., Phillips et al., 2009; Feldpausch et al., 2016; Rifai et al., 2018; Bennett et al., 2023). Satellite-based analyses also reveal the impacts of climate anomalies on carbon 45 dynamics (Palmer et al., 2018; Fan et al., 2019), providing a synoptic view of ecosystem 46 47 productivity. However, we still lack ground-based, tree-level measurements of net primary productivity (NPP) through extreme tropical climate events, hindering our understanding of key 48 49 aspects of the vegetation carbon cycle response, such as recovery following drought events, and 50 NPP allocation. Measuring these ecosystem responses directly is helped by tracking long-term 51 forest dynamics in permanent plots but especially requires high-fidelity process-based 52 measurements sustained over time. These are exceptionally challenging to make and require 53 long-term dedication to measurements before, during, and after major climate events like the 54 2015-16 El Nino. We know especially little about how the productivity of savanna ecosystems is affected by El Niño events, especially in the extensive Amazonia-Cerrado transition in South 55 56 America.

57 The Amazonia-Cerrado transitional region contains a mixture of Amazonia and Cerrado 58 species, making the species composition of this region unique and diverse (Ratter et al., 1973; 59 Marimon et al., 2006; Morandi et al., 2016). Despite its ecological importance, this region has 60 been greatly impacted by deforestation (~41% between 1984 and 2014) so that today only 61 fragments of native vegetation remain (e.g., Marques et al., 2020). In recent decades, the 62 remaining vegetation has been affected by increasing temperatures, frequent wildfires, extreme 63 drought events, and the long-term lengthening of the dry season (e.g., Reis et al., 2018; Silvério 64 et al., 2019; Nogueira et al., 2019; Matricardi et al., 2020; Araújo et al., 2021a). Deforestation, 65 together with increases in temperature and reduction in precipitation during El Niño events, 66 increases wildfire occurrence and carbon emissions, reducing the capacity of the vegetation to 67 act as a carbon sink (Covey et al., 2021; Gatti et al., 2021). As the Amazonia-Cerrado transition 68 is the driest, warmest, and most fragmented region in the Amazon basin (e.g., Matricardi et al., 69 2020; Marques et al., 2020; Covey et al., 2021; Reis et al., 2022) it is especially vital to understand 70 better how climate change and extreme climate events impact carbon dynamics.

71 This transition is composed naturally of a mosaic of vegetation, being the typical cerrado 72 (referred to as cerrado hereafter) and woodland savanna (i.e., cerradão) the most common in the 73 regions (Ratter et al., 1973; Marimon et al., 2006, Oliveras & Malhi, 2016). Despite co-existing in 74 the same space, cerrado and cerradão vegetation formations show contrasting characteristics 75 (Marimon-Junior & Haridasan, 2005; Marimon et al., 2006). The cerradão is a transitional forest-76 savanna characterized by closed canopy, understory formed by small shrubs and herbs, with few 77 grasses, and average height of the tree stratum varying from 8 to 15 m, tree cover of 50 to 90% 78 (Ribeiro & Walter, 2008, Oliveras & Malhi 2016), while cerrado is a savanna vegetation type with 79 a discontinuous canopy, trees, and shrubs with grass understorey, and a low average height of





just 3 to 6 m, with tree cover of 20 to 50% (Marimon-Junior & Haridasan, 2005; Ribeiro & Walter,
2008).

82 In the cerrado, most species are deciduous, fully shedding their leaves in the dry season, 83 while most cerradão species are brevi-deciduous. Although the dominant species of both 84 vegetation types show strong stomatal efficiency (Jancoski et al., 2022), trees in the cerrado have 85 smaller stomata and higher trichome density than individuals occurring in the cerradão, 86 anatomical features that help the leaves minimise their water loss (Araújo et al., 2021b; Araújo et al., 2023). Furthermore, for species that co-occur in both cerrado and cerradão, individuals in 87 88 cerrado lose their leaves earlier than cerradão in the dry season. The early loss of leaves in the cerrado means that the photosynthetic apparatus is not harmed during the driest and hottest 89 90 period of the year. In the cerradão, individuals take longer to lose their leaves, which makes them 91 more sensitive to changes in temperature increases, both current and projected (Araújo et al., 92 2021b). Cerradão trees are taller than cerrado trees, this characteristic may offer cerradão greater 93 sensitivity to drought, since taller trees have wider xylem vessels (Araújo et al., under review). 94 These characteristics (e.g., larger stomata and greater maximum stomatal pore opening) may 95 give the cerradão greater sensitivity to disturbances generated by climatic anomalies, such as the 96 2015-2016 El Niño.

97 Here, by setting up and sustaining intensive, long-term monitoring plots that experience a similar climate at cerradão and cerrado, we aimed to quantify and compare the effect of the 98 99 2015/2016 El Niño on the carbon cycle (productivity and allocation) of these two vegetation types. 100 Our guiding questions are: 1) Does productivity and allocation differ between cerradão and 101 cerrado? 2) How did the 2015-2016 El Niño affect productivity and allocation in the cerradão and 102 cerrado? 3) Did the cerradão and cerrado regain productivity after the El Niño? 4) What are the 103 trade-offs in resource allocation between canopy, wood and fine roots during drought in the two 104 vegetation types?

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106 2 Materials and Methods

107 2.1 Study sites

108 We conducted this study in two long-term plots in cerradão (transitional forest-savanna - NXV-109 02; Forestplots code) and cerrado (typical cerrado - savanna - NXV-01) located in the Bacaba 110 Municipal Park, in Nova Xavantina, Mato Grosso State, Central Brazil. The park covers 111 approximately 500 ha in the transition zone between the Cerrado (Brazilian savanna) and 112 Amazonia. Since the two plots are about 300m apart, they experience very similar climates, which 113 corresponds to the Aw (tropical with dry winters) type in Köppen's classification system (Alvares 114 et al., 2013). As measured by station #83319 of the Brazilian National Institute of Meteorology 115 (INMET), the mean monthly temperature is 24.8 °C, the total annual precipitation is 1440 mm (Peixoto et al., 2017), and the average altitude of the park is ~ 250 m (Marimon Junior & 116 117 Haridasan, 2005). 118 The plots were established in 2002 (Marimon Junior & Haridasan, 2005) and have been

re-censused multiple times. Since 2010, these have been part of the PELD (Cerrado-Amazonia





120 Forest Transition: ecological and socio-environmental bases for Conservation), RAINFOR (Amazonia Forest Inventory Network; ForestPlots.net et al., 2021) and ForestPlots.net 121 122 collaborations, and since 2014, part of GEM (Global Ecosystems Monitoring network; Malhi et al., 123 2021). The plots have facilitated multiple studies, such as soil, composition and diversity of species, biomass, and tree dynamics (e.g., Marimon Junior & Haridasan, 2005; Marimon et al., 124 125 2014). Partial data on carbon cycling have previously been published for the cerradão plot, on 126 litterfall, soil efflux and carbon stocks at fine roots, litter layer, and stem (Peixoto et al., 2017; Peixoto et al., 2018). Here we provide the first comprehensive description of the carbon cycling 127 128 in both plots as well as an extended time series that provides insight into the aftermath of 129 2015/2016 El Niño event.

130 The plots have not been burned since 2008. The cerradão plot is a transitional forest-131 savanna characterized by the overlap of savanna and forest species, a closed canopy, and with dominant species (notably Hirtella glandulosa Spreng. and Tachigali vulgaris L.G. Silva & H.C. 132 133 Lima). This type of cerradão was recognised by Ratter et al. (1973) as Hirtella glandulosa 134 cerradão. Trees and shrubs with grass understorey and open canopy characterize the cerrado. 135 Here the two dominant tree species are Qualea parviflora Mart. and Davilla elliptica A.St.-Hil. 136 (Marimon Junior & Haridasan, 2005; Marimon et al., 2014). The vegetation of the cerrado is 137 becoming denser and there are not many grasses present (Morandi et al., 2015), possibly due to 138 fire exclusion.

139 The soil is similar across the plots - sandy loams of the yellow latosol type, acidic (pH < 140 5.0) and dystrophic (Ca2+ ~ 0.4 cmolc kg-1), with high levels of exchangeable aluminium (Al3+ > 141 1.3 cmol_c kg⁻¹) – however, the cerradão soil presents higher percentages of clay and potential 142 water holding capacity than the cerrado (Marimon Junior & Haridasan, 2005). These differences 143 in soil texture may explain the different vegetation formations in these two closely adjacent sites. 144 The average height of the trees in cerrado is 3.7 m, and a basal area of ~14.9 m² ha⁻¹. For the 145 cerradão, the average tree height is 6.4 m and basal area of ~ 21.4 m² ha⁻¹ (Marimon Junior & 146 Haridasan, 2005). The species number was 77 in both and the number of trees similar (cerrado 147 = 1890 and cerradão = 1884) (Marimon Junior & Haridasan, 2005).

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149 2.2 Site climate and the El Niño 2015/2016 event

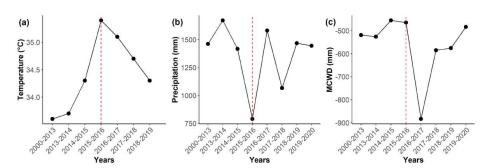
We used the climate variables – air temperature, relative air humidity, and precipitation – in time series from a Meteorological Station (World Weather Station 83319), about 800 m from the plots. We calculated the maximum climatological water deficit (MCWD), a climatological measure of tropical forest water stress (see Aragão et al., 2007). To calculate MCWD, we considered a standardized evapotranspiration (ET) value for wet season tropical forests of 100 mm month-1 (Aragão et al., 2007).

The seasonality of the plots is marked by two well-defined seasons – cooler-dry (April to September) and hot-rainy (October to March). We defined the twelve months from May 2015 to April 2016 as the climate of the 2015-2016 El Niño Southern Oscillation event based on Liu et al. (2017). During the El Niño, the plots experienced record mean and mean monthly maximum





- annual temperatures (26.0 °C and 35.4 °C) and record low total annual precipitation (790.2 mm),
- and in September 2016, record low annual MCWD (-883.7 mm) (Fig. 1; Table S1).
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Fig. 1. Climate variables between 2000 and 2020 for *cerrado* and *cerradão*. We should (a) temperature (°C), (b) precipitation (mm/year) and (c) maximum climatological water deficit (MCWD, mm in a rolling year) with the 1st month of the dry season (May) representing the beginning of each year's climatic calendar. The temperature indicates the average maximum monthly temperatures. The dashed red line indicates the El Niño periods. Climatic data are from meteorological station #83319 of the Brazilian National Institute of Meteorology (INMET). See Table S1 for data.

172 2.3 Field methods

- 173 We followed the GEM protocol manual (Marthews et al., 2014; Malhi et al., 2021) to collect the
- 174 data for this study. We measured the main components of NPP: canopy (leaves, twigs,
- 175 reproductive parts, and others), wood (stems and branches), and fine roots (see Table 1).
- 176

¹⁷⁸ *cerrado* (NXV-01) plots in the transition zone between Amazonia and Cerrado. See also the 179 RAINFOR-GEM manual (Marthews et al., 2014). nc= no collected.

Component	Description	Sampling period	Sampling interval
	Above-ground net primary productivity (NPP _{AG}) and	d biomass	
Abovegroun d coarse wood net primary productivity (<i>NPP</i> _{stem}) and Stem biomass	Forest inventory or plot census were done in the years listed. The default measurement point was set at 30cm (DAS _{30cm}) above soil surface, instead of a typical forest diameter at breast height at 1.3m. All trees ≥ 5 cm DAS _{30cm} were censused, based on which, we calculated mortality and recruitment rate of new trees. Stem biomass for each tree was calculated. The sum of all alive trees in each census is termed stem biomass. As we have noticed trees stem shrinking, we calculate stem NPP as the change of alive trees biomass between two censuses, where dead and new recruit trees are excluded but shrinking trees are included. We have also presented 'Stem Diameter growth' which excludes shrinking trees, dead trees and new recruit trees. In 2014, five 10 m x 10 m subplots were established to census sfraction of smaller trees biomass is measured using similar method and not counted in Stem Biomass.	2013 - 2021 (NXV-01) 2013 - 2020 (NXV-02)	2013, 2015, 2011 and 2021 (NXV-01) 2013, 2015, 2017 and 2020 (NXV-02)

¹⁷⁷ Table 1. Field methods for intensive monitoring of NPP components from cerradão (NXV-02) and





	Biomass of each stem was calculated using Rezende et al. (2006) specific allometric equation for the Cerrado: C = $0.24564+0.01456^{*}(D/10)^{2}$ [*] H where C is aboveground Carbon stocks (kg), D is the diameter (30 cm above the soil), and H is the height (m). The authors assumed that dry stem biomass is 50% carbon. Systematic uncertainty of +25% was assigned to values for error propagation. Errors calculated as the sampling error associated with variation between the transects.		
Branch turnover net primary productivity (<i>NPP</i> branch turnover)	Branchfall > 2 cm diameter (excluding that associated with dead trees) surveyed within four 1 m x 100 m transects; small branches were cut to include only the transect-crossing component, removed, and weighed. Larger branches had their dimensions taken (diameter at three points), and assigned a wood density value according to decomposition class (Harmon et al., 1995). Biomass of each branch was calculated. The first collection of branchfall in 2014 lead to 'woody debris biomass', which accounts for nacromass in the ground litter layer. 'Woody debris biomass' is not included in above nor belowground biomass. The biomass of branch has been implicitly included in <i>NPP</i> stem. Branchfall was then collected from the same transects every 3 months which lead to <i>NPP</i> branch turnover	2014 – 2019	Every 3 months
Litterfall net primary productivity (<i>NPP</i> _{litterfall})	Litterfall production of dead organic material (< 2 cm diameter) calculated by collecting litterfall in 0.2827 m ² circular collectors placed at 1 m above the ground at the center of each of the 25 subplots in each plot. Litter separated into leaves, twigs, reproductive parts (flowers, fruits, and seeds), and unidentifiable. <i>NPP</i> _{itterfall} calculated as follows: <i>NPP</i> _{itterfall} = <i>NPP</i> _{canopy} – Loss to Leaf Herbivory. Litterfall separated into different components, oven-dried at 65°C to constant mass, and weighed. Litter estimated to be 49.2% carbon, based on mean Amazonia values (S. Patiño, unpublished analysis). Errors calculated as the sampling error associated with variation between the litter traps.	Jan 2014 – Dec 2019	Every 14 days
Leaf Area Index (LAI) and Leaf biomass	Hemispherical photos taken with a digital camera (Nikon OP 10mm) and hemispherical lens (Nikon 10mm fisheye lens) near the center of each of the 25 subplots in each plot at a standard height of 1 m and during overcast conditions. LAI estimated from these images using Hemisfer software (licensed version 2.12; http://www.wsl.ch/dienstleistungen/produkte/software/he misfer/index_EN). LAI estimated from hemispherical photos using the standard Li-Cor LAI-2000 method, based on the Miller (1967) equations, and correcting for non-linearity and slope effects (Schleppi et al., 2007) and canopy clumping (Chen & Cihlar, 1995). Thresholds were set to detect separately for each ring (6 rings) according to Nobis & Hunziker (2005). Errors calculated as the sampling error through variation among subplots. Leaf biomass calculated as leaf area index (LAI)/specific leaf area (SLA), where LAI is the plot mean over the study period, and SLA is the basal area-weighted plot mean	Jun 2015 – Jan 2020	Every 3 months





	2014 (Neyret et al., 2016).		
Loss to leaf herbivory (<i>NPP</i> _{herbivory})	Estimated based on Neyret et al. (2016)'s observation that the loss to herbivory was 3.11% in NXV-01 and 4.43% in NXV-02. The data collection was conducted between March and May 2014. Each leaf's fractional herbivory (<i>H</i>) was calculated as $H = (A_{nh} - A_h) / A_{nh}$. Where A_h is the area of each leaf, including the damage incurred by herbivory, and A_{nh} is the leaf area prior to herbivory (Neyret et al., 2016). The average value of <i>H</i> of all leaves collected per litterfall trap was derived, and plot-level means were calculated. Systematic uncertainty of \pm 50% assigned to values for error propagation.	nc	nc
	Below-ground net primary productivity (NPP	BG)	
Coarse root net primary productivity (<i>NPP</i> _{coarse} root)	Root biomass estimated based on Miranda et al. (2014) that is specific for the vegetation types of Cerrado. Based on this study, the Root(belowground): shoot ratio (aboveground) biomass is 1.37 to <i>cerrado</i> and 0.22 to <i>cerradão</i> . Systematic uncertainty of ±20% assigned to values for error propagation. We used these ratios, 1.37 (at NXV-01) and 0.22 (at NXV-02) to derive <i>NPP</i> _{coarse root} from <i>NPP</i> _{stem}	nc	nc
Fine root net primary productivity (<i>NPP</i> fine root) and fine root biomass	In each plot, sixteen ingrowth cores (mesh cages 12 cm diameter, to 30 cm depth) were installed. Roots were manually removed from the soil samples in four 10 min time steps, according to a method that corrects for underestimation of biomass of hard-to-extract roots (Metcalfe et al., 2007) and used to predict root extraction beyond 40 min (up to 120 min); typically, there was an additional 33% correction factor for fine roots not collected within 40 min. Correction for fine roots productivity below 30 cm depth (Galbraith et al., 2013) increased the value by 39%. Errors were calculated as the sampling error associated with variation between the sampling points. Root-free soil was then re-inserted into the ingrowth core. Collected roots were thoroughly rinsed, oven-dried at 65°C to constant mass, and weighed. This process was repeated for each measurement thereafter. Fine root biomass was calculated from harvested fine roots during the first installation of ingrowth. The subsequent fine root collection from the ingrowth cores lead to <i>NPP</i> fine root	Sep 2014 – Feb 2020	Every 3 months

over the study period. We used the SLA value of March

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182 2.4 NPP calculation

- 183 We measured the NPP in the two plots between 2014 and 2020 (Table 1). We calculated all major
- 184 components of NPP using the following equations:
- 185 NPPtotal = NPPcoarse root + NPPfine root + NPPstern + NPPbranch + NPPlitter fall + NPPherbivory (1)
- 186 NPP_{canopy} = NPP_{litter fall} + NPP_{herbivory} (2)
- 187 NPPwoody = NPPcoarse root + NPPstem + NPPbranch turnover (3)
- 188 NPP_{root} = NPP_{fine root} (4)





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191	Our calculations above neglect several small NPP terms, such as NPP lost as volatile
192	organic emissions (NPP_{VOC}), unmeasured litter trapped in the canopy, or litter dropped from
193	understorey flora below the litter traps (1 m). However, in central Amazonia, Malhi et al. (2009)
194	found NPP _{VOC} was a relatively minor NPP term (0.13 + 0.06 Mg C ha ⁻¹ year ⁻¹). For belowground
195	NPP, we do not include root exudates and mycorrhizae that account for < 2 Mg C ha ⁻¹ year ⁻¹ ,
196	representing a modest part of the carbon fluxes (Malhi et al., 2017). Thus, we focus on the canopy,
197	wood, and fine roots productivity, which account for over 85% of NPP (See Riutta et al., 2018 and
198	their references).

We calculated the relative allocation to the main NPP components (woody, canopy, andfine roots NPP) for leaves, fine roots, and stem following the equations:

201 Allocation_x = $(NPP_x * 100)/NPP_{total}$ (6)

 $NPP_{ACW} = NPP_{stem}$ (5)

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203 2.5 Calculation of measurements uncertainty

Estimation of measurements uncertainty for each NPP component is explained in details in Table 1. For components that are not directly measured, for example NPP_{total} as a sum of several components, we combine relevant error by error propagation with standard quadrature rules (Hughes & Hase, 2010; Malhi et al., 2015). During the above process, we also assigned significant systematic errors to capture uncertainties related to sampling methodology or scaling approaches (see Table 1); these factors were consistent with those applied in similar previous studies (Malhi et al., 2009, 2015; 2017; Girardin et al., 2010; Galbraith et al., 2013).

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212 2.6 Data analyses

213 Our analyses were focused on comparing NPP among the years (2014 to 2019) - comprising the 214 periods before, during, and after the El Niño 2015/2016 events - in cerrado and cerradão. We 215 compared the stem and canopy biomass of the two vegetation types over time using repeated 216 two-way analysis of variance (ANOVA-two way). We used Tukey's post hoc test to compare the 217 different years in each plot. We used the same analysis to compare productivity and carbon 218 allocation across different compartments. In cases where the residuals violated the ANOVA 219 assumptions, we used Friedman's non-parametric analysis. We performed all analyses in the R 220 environment and adopted a significance level of 0.05. To improve the accessibility of colour figures with COLORBREWER 2.0. 221

222

223 3 Results

224 3.1 Net primary productivity

The net primary productivity (NPP) in the *cerradão* was ~ 30% higher compared to that of the *cerrado* prior to the occurrence of El Niño (*cerradão* = ~9.3±0.57 Mg C ha⁻¹ year⁻¹; *cerrado* = ~6.5±1.12 Mg C ha⁻¹ year⁻¹ Fig. 2; Table S2). This is due to the greater productivity in the canopy and stem in the *cerradão* (Fig. 2; Table S2). During the El Niño, *cerradão* NPP decreased to





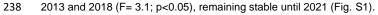
229 6.6 ± 0.6 Mg C ha⁻¹ year⁻¹ and became similar to the *cerrado* (6.6 ± 1.3 Mg C ha⁻¹ year⁻¹; Fig. 2; 230 Table S2).

Cerradão NPP was severely affected in 2016 during the El Niño event (-29%). In 2018 it
 was still 13% lower than pre-El Niño conditions (Fig. 2). Additionally, stem biomass declined
 significantly after El Niño (F= 19.3, p< 0.001) and did not return to the values registered before
 the event (Fig. S1).

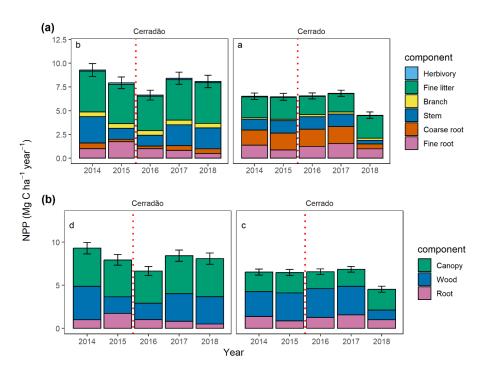
235 In the *cerrado*, NPP did not vary much before and during the El Niño. However, in 2018,

236 productivity reduced by ~30%, due especially to the reduction in stem productivity. Despite this,

237 stem biomass was not significantly influenced by El Niño and increased significantly between



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Fig. 2. Mean total annual net primary productivity (NPP) between 2014 and 2018 split into its components (a) and annual NPP allocation into the canopy, wood, and root components (b) at *cerrado* and *cerradão*. The branch data from *cerradão* was collected in 2014 and repeated in other years. The error bars represent the standard error for total NPP. The dashed red line indicates the El Niño periods.

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In the *cerradão*, the most affected component was stem net primary productivity (NPPs),
which was reduced by 58% during and after El Niño (F= 15.6, p< 0.001; Fig. 3A). In 2019 it was
still -21% lower than pre-El Niño conditions. When we consider only those trees that were alive
before El Niño and remained alive after the event, the *cerradão* reduced NPPs significantly during
the event, but after the event, NPPs was greater than before the El Niño (Fig. 3B; F= 25.6, p<





- 0.001). This is mainly due to two critical species for this transitional forest, *Hirtella glandulosa*Spreng. and *Tachigali vulgaris* L.G.Silva & H.C.Lima, which contributed 22% and 17% to NPPs
 after El Niño. Before El Niño, *T. vulgaris* was the species that most contributed to NPPs (26%).
 In the *cerrado*, trees showed less diameter growth during and after the event (Fig. 3B; F= 109.7,
- p < 0.001). However, stem productivity was not affected during the event (Fig. 3A).
 - (A) (B) 3 Stem NPP (MgC ha⁻¹ year⁻¹) Lree DBH Growth (cm year⁻¹) 0.3 0.2 0.1 0.0 2014 2015 2016 2017 2018 2019 2013-2015 2015-2017 2017-2021 Year Year Plot Cerradão Cerrado Plot Cerradão Cerrado

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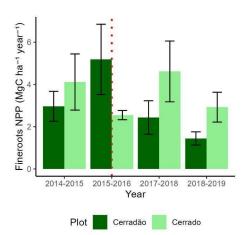
Fig. 3. Stem dynamics in *cerrado* (light green) and *cerradão* (dark green). (a) Stem net primary productivity (NPPstem, MgC ha⁻¹ year⁻¹) for stem larger than 5cm diameter. We note that there are negative stem NPP values due to those trees that lose bark or water from the stem in the dry periods, especially in *cerrado* after 2017. (b) The growth of tree diameter (measured at 30cm above soil surface) (cm year⁻¹), calculated as the increase in DAS between two censuses divided by time. Only growth is included, in other words, trees with shrinking stems are excluded The dashed red line indicates the El Niño periods.

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In the *cerradão*, fine root net primary productivity (NPPfr) production increased significantly (+42%) during El Niño (F= 17.3, p< 0.001), but in later years productivity declined (Fig. 4). The *cerrado* presented the opposite pattern observed in the *cerradão*. NPPfr reduced by 38% during the event (F= 5.6, p= 0.001; Figs. 2 and 4). However, the NPPfr of this component re-established itself soon after El Niño, but experienced a decline of ~ 38% in 2018.







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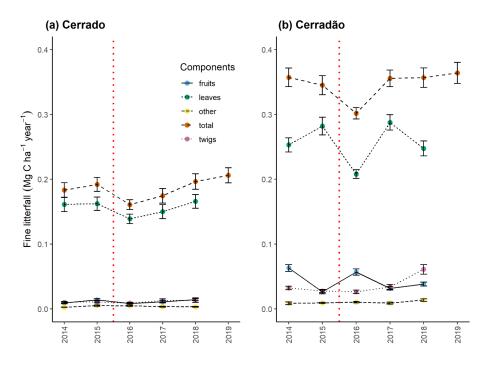
Fig. 4. Fine root net primary productivity (NPPfr) for *cerrado* (light green) and *cerradão* (dark green) between September 2014 and August 2019. The error bars represent the standard error.
 The dashed red line indicates the El Niño periods.

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281 Canopy productivity was affected after the El Niño event in both cerradão (F= 2.8, p= 0.01) 282 and cerrado (F= 6.7, p< 0.001) (Fig. 5). However, the NPP of this component was re-established 283 two years after the event. For the cerradão, it is worth highlighting the drop in fruit production after 284 the event, which had not yet re-established itself two years after El Niño (Fig. 6). Furthermore, 285 after El Niño, both plots show declining and then recovering LAI (Fig. S2). We also noted that 286 following El Niño, the variability of LAI increased among subplots, potentially due to clearings emerging from heightened tree mortality. The average annual mortality rate increased during and 287 288 after El Niño, especially in the cerradão (Fig. 6). 289







290

Fig. 5. Mean monthly productivity in canopy litterfall and its components for *cerrado* (a) and *cerradão* (b) between 2014 and 2019: (fruits) flower, fruit, and seed fall; (leaves) leaf fall; (other) not identified and (total) total canopy fine litterfall (as measured in litter traps); (twigs) twig fall (< 2 cm). The error bars represent the standard error. The dashed red line indicates the El Niño periods.</p>

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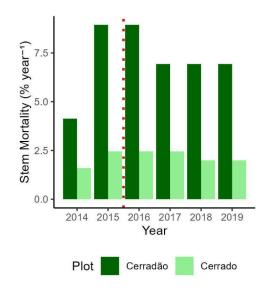






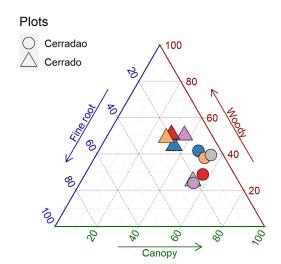
Fig. 6. Stem mortality, shown as the percentage of dead trees to the number of trees in the first census divided by time. The dashed red line indicates the El Niño periods.

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303 3.2. Allocation between plots, components, and years

304 In general, NPP allocation differed significantly between cerradão and cerrado plots (Fig. 7), but 305 this varies within the components (F= 41.7, p< 0.001). The allocation to canopy was greater in 306 cerradão (53+3%) than in cerrado (37+10%). In contrast, allocation to woody and roots was 307 greater in the cerrado (woody= 44+11%; roots= 19+4%) than in the cerradão (woody= 34+8%; 308 roots= 13+6%). Over the studied time, the NPP allocation showed inter annual variation in cerrado 309 but no clear drought signal the main axis of interannual variation was a trade-off between canopy 310 investment and root allocation, with woody allocation remaining constant (Fig. 7). However, in 311 2018, ~ three years after El Niño, the allocation of canopy and wood changed drastically, showing 312 an opposite pattern to previous years. In the cerradão, there is a clear drought signal with 313 increased investment in fine roots during the drought, and reduced investment in woody growth; 314 canopy allocation remained fairly constant.

315



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Fig. 7. Relative allocation (% of total) of net primary productivity (NPP) to canopy, woody, and fine root NPP in *cerrado* and *cerradão*. 2014= blue, 2015= pink, 2016= red, 2017=orange and 2018= grey. Woody components include stem, coarse root and branch turnover; Fine root includes fine root NPP only (no roots exudates); Canopy includes litterfall and herbivory

321 322

323 4 Discussion

324 Cerradão and cerrado showed contrasting responses to the 2015/2016 El Niño-associated

325 drought event. The cerrado proved to be more resistant and resilient, i.e., most of the parameters





assessed (e.g., stem carbon stock, canopy productivity) changed little during the event, and those
that experienced a reduction soon re-established themselves (e.g., production of fine roots). In
contrast, the *cerradão* showed lower resistance and resilience: stem carbon stock and mortality,
productivity and allocation were affected during and after the El Niño event and even three years
after, most of the parameters assessed were not similar to those observed before the event. Our
findings demonstrate the high sensitivity of the *cerradão* to extreme drought events.

332 The productivity found in our cerradão (9.3+0.57) was similar to that observed in transitional forests in Africa (9.2 to 13.1; Moore et al., 2018), and some low-fertility humid forest sites in 333 334 lowland Amazonia in Colombia and Brazil (~8.1 to 10.3; Aragão et al., 2009; Girardin et al., 2010). It is markedly greater than observed in seven premontane and montane sites in Peru (~3.9 to 6.4; 335 336 Girardin et al., 2010) and lower than observed in lowland tropical forest plots in south-western 337 Amazonia $(15.1 \pm 0.8 \text{ and } 14.2 \pm 1.0; \text{ Malhi et al., } 2015)$ and nutrient-rich soils forests $(17.0\pm1.4;$ 338 Aragão et al., 2009). Nevertheless, the total productivity of cerradão was more affected (-29%) 339 than the Amazonia rainforest (-7.6% to -8.5%) during the El Niño drought of 2015/2016 (Machado-340 Silva et al., 2021). Moreover, the reduction in stem productivity was much larger (-58%) than that 341 estimated for tropical forests as a whole (-8.3% in 1997/1998, and -9% in 2015/2016 (Rifai et al., 342 2018). This demonstrates the high sensitivity of this vegetation to climate anomalies.

343 The higher mortality of cerradão may be related to the hydraulic characteristics of the main 344 species, such as Tachigali vulgaris, a pioneer species with stomatal control tending to an 345 anisohydric condition and, therefore, more susceptible to hydraulic failure (Jankoski et al., 2022). 346 In addition, many species are brevi-deciduous, so the plants continue to photosynthesize even 347 during water scarcity (e.g., Jancoski et al., 2022). Other factors, such as a lack of strategies to avoid water loss, may also contribute, like low trichome density in their leaves and smaller stomata 348 349 (Araújo et al., 2021b). Another possible cause of the high mortality in the cerradão is the unusually 350 intense winds that hit trees with xylem tissue already weakened by the effects of drought and heat 351 due to the El Niño event (Reis et al., 2022). Cerradão trees are taller than cerrado trees, which 352 makes them more susceptible to wind disturbances. Once broken, even just part of the crown, 353 the tree is at greater risk of death (Reis et al., 2022).

354 Despite the high mortality, the trees that remained alive showed higher stem productivity 355 than before the El Niño. This may be related to the greater opening of clearings, favouring carbon 356 uptake and plant growth due to the greater availability of light, water and nutrients to the remaining 357 trees. During the El Niño drought, a decline in the growth of Tachigali vulgaris trees was observed, 358 leading to a shift in the primary contributor to stem productivity from T. vulgaris to Hirtella 359 glandulosa. The role reversal of these two species can be explained by the high mortality and low 360 growth rate of T. vulgaris during and after El Niño. T. vulgaris is considered a key species for 361 cerradão due to its high biomass gain after disturbances such as fire (Reis et al., 2015; 2017), but it is sensitive to drought. The high sensitivity of T. vulgaris to drought events may be attributed 362 363 to the increased xylem tension required to extract water from the soil, making it more prone to 364 embolism (Jancoski et al., 2022). Consequently, this results in reduced growth and higher 365 mortality compared to H. glandulosa. On the other hand, H. glandulosa proved to be more





366 resistant: it has high foliar trichome density, which works as a strategy to prevent water loss (e.g. 367 Gianoli & González-Teuber, 2005; Araújo et al., 2021b). In the cerrado, we observed the opposite 368 pattern; the productivity of trees that remained alive continued to decline after the event. In this 369 vegetation type, plant mortality was low, and the surviving plants had to compete to stay alive, which may explain the lower productivity after El Niño. Furthermore, many trees in the cerrado 370 371 shed their outer bark, which may have affected the diameter measurement and, consequently, 372 the productivity of the stem. The loss of bark from cerrado plants, especially after fire and drought 373 events, makes the measurement of trunk productivity inaccurate.

374 The high production of fine roots in cerradão during drought is probably a strategy to 375 increase soil water uptake during the period of soil water scarcity (Metcalfe et al., 2008). However, 376 this strategy on partially ameliorates drought risk, as tree mortality was high even with a high 377 investment in fine roots. The cerrado, on the other hand, showed the opposite strategy, investing 378 less in fine roots during the event. However, shortly afterwards, the productivity of this component 379 was similar to that observed before El Niño. Lowland terra firme have less root growth during the 380 dry season but had greater specific root length and surface area where soil moisture was depleted 381 (Metcalfe et al., 2008), and the cerradão presented a strategy similar to these Amazonia forests, 382 but not the cerrado.

383 Both cerrado and cerradão adopted the strategy of losing more leaves during El Niño. It is 384 well known that during periods of water stress in the soil, plants lose their leaves as a strategy to 385 avoid water loss and consequent mortality (e.g., Brando et al., 2008). This strategy can also have 386 nutrient cycling benefits: the nutrients released to the litter layer and soil after leaf drop and can 387 later be reabsorbed by the plants when they re-establish leaf growth after a high stress period 388 (e.g., Oliveira et al., 2017). The high leaf loss during El Niño may have contributed to lower 389 photosynthetic activity of plants (e.g., Luo et al., 2018; Kaewthongrach et al., 2020), consequently 390 affecting carbon accumulation.

391 The canopy-wood-fine root trade-offs identified here are different from those reported by 392 Doughty et al. (2014) for a somewhat similar Amazonian forest-dry forest transition in Bolivia, with 393 similar rainfall but more fertile soils. There, the site with better water availability (related to soil 394 properties) hosted an Amazonian forest which showed wood-canopy trade-offs during drought. 395 The drier site hosted chiquitano dry forest with wood-fine root trade-offs during drought. Our 396 cerradão site shows similar wood-fine root trade-offs to the chiquitano forest, whereas our cerrado 397 site shows a different canopy-fine root trade off. One possibility is that these shifting trade-off 398 strategies reflect points on an aridity continuum from sub-humid Amazonian forest (wood-canopy 399 trade-off) through transitional or seasonally dry forests (wood-fine root trade-off) through to 400 savanna (wood-canopy trade-off). Alternatively, the differences in soil fertility may play a role, 401 changing the costs and advantages of investment in fine-root production.

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406 5 Conclusions

407	<i>Cerradão</i> is a vital transitional vegetation type at the Amazon-Cerrado ecotone, as it is in contact
408	with the two main Brazilian biomes, Cerrado and Amazonia. However, this vegetation type
409	appears to be susceptible to climatic events (present study), wildfires (Reis et al., 2015; 2017)
410	and wind storms (Reis et al., 2022). One of the most frequent species in cerradão (T. vulgaris),
411	that is especially important for carbon uptake, proved to be very sensitive to the climatic event.
412	Thus, if these extreme drought events continue to become more frequent and intense, cerradão
413	may release more carbon than absorbs, as observed here. In addition, the cerradão serves as a
414	connection between the savanna and the forest, acting as a kind of buffer-barrier for the Amazonia
415	to the effects of environmental stressors along its contact with the <i>cerrado</i> . Our results suggest
416	that the more frequent occurrence of El Niño events can break this natural barrier, creating
417	conditions for the progressive degradation of the forest along the edges.
418	
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420	R.F., B.S.M., H.Z., C.A.J.G., E.A.O., K.S.P., L.J.S., E.L.S., E.B.S., K.P.S., M.D.A.G., C.A.L.D.,
421	$O.L.P. \ and \ I.O.M.); \ Y.M., \ B.H.M.Jr. \ and \ I.O.M. \ were \ involved \ in \ planning \ and \ supervised \ the \ work;$
422	$S.M.R.,\ R.F.,\ E.A.O.,\ K.S.P.,\ L.J.S.,\ E.L.S.,\ E.B.S.,\ K.P.S.\ and\ M.D.A.G.\ performed\ the\ field$
423	measurements; S.M.R., H.Z., and C.A.J.G. performed the analyses and made the figures. All
424	authors discussed the results and contributed to the final manuscript.
425	
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452	
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454	The contact author has declared that none of the authors has any competing interests.
455	
456	References
457	Alvares, C. A., Stape, J. L., Sentelhas, P. C., Gonçalves, J. D. M., & Sparovek, G. (2013).
458	Köppen's climate classification map for Brazil. Meteorologische Zeitschrift, 22(6), 711-728.
459	Aragão, L. E. O., Malhi, Y., Roman-Cuesta, R. M., Saatchi, S., Anderson, L. O., & Shimabukuro,
460	Y. E. (2007). Spatial patterns and fire response of recent Amazonian droughts. Geophysical
461	Research Letters, 34(7).
462	Aragão, L. E. O. C., Malhi, Y., Metcalfe, D. B., Silva-Espejo, J. E., Jiménez, E., Navarrete, D.,
463	& Vásquez, R. (2009). Above-and below-ground net primary productivity across ten
464	Amazonian forests on contrasting soils. Biogeosciences, 6(12), 2759-2778.
465	Araújo, I., Marimon, B. S., Scalon, M. C., Cruz, W. J., Fauset, S., Vieira, T. C., & Gloor, M. U.
466	(2021b). Intraspecific variation in leaf traits facilitates the occurrence of trees at the
467	Amazonia–Cerrado transition. Flora, 279, 151829.
468	Araújo, I., Marimon, B. S., Scalon, M. C., Fauset, S., Junior, B. H. M., Tiwari, R., & Gloor, M.
469	U. (2021a). Trees at the Amazonia-Cerrado transition are approaching high temperature
470	thresholds. Environmental Research Letters, 16(3), 034047.
471	Araújo, I., Scalon, M. C., Amorim, I., Menor, I. O., Cruz, W. J., Reis, S. M., & Marimon, B. S.
472	(2023). Morpho-anatomical traits and leaf nutrient concentrations vary between plant
473	communities in the Cerrado-Amazonia transition?. Flora, 306, 152366.
474	Araújo et al. (under review). Taller trees exhibit greater hydraulic vulnerability in southern
475	Amazonian forests. Environmental and Experimental Botany.
476	Bennett, A. C., Rodrigues de Sousa, T., Monteagudo-Mendoza, A., Esquivel-Muelbert, A.,
477	Morandi, P. S., Coelho de Souza, F., & Phillips, O. L. (2023). Sensitivity of South American
478	tropical forests to an extreme climate anomaly. Nature Climate Change, 13(9), 967-974.
479	Brando, P. M., Nepstad, D. C., Davidson, E. A., Trumbore, S. E., Ray, D., & Camargo, P. (2008).
480	Drought effects on litterfall, wood production and belowground carbon cycling in an Amazon
481	forest: results of a throughfall reduction experiment. Philosophical Transactions of the Royal
482	Society B: Biological Sciences, 363, 1839-1848.
483	Chen, J. M., & Cihlar, J. (1995). Quantifying the effect of canopy architecture on optical
484	measurements of leaf area index using two gap size analysis methods. IEEE transactions
485	on geoscience and remote sensing, 33(3), 777-787.





486	Covey, K., Soper, F., Pangala, S., Bernardino, A., Pagliaro, Z., Basso, L., & Elmore, A. (2021).
487	Carbon and beyond: The biogeochemistry of climate in a rapidly changing Amazon. Frontiers
488	in Forests and Global Change, 4, 11.
489	Doughty, C. E., Malhi, Y., Araujo-Murakami, A., Metcalfe, D. B., Silva-Espejo, J. E., Arroyo, L.,
490	& Ledezma, R. (2014). Allocation trade-offs dominate the response of tropical forest growth
491	to seasonal and interannual drought. Ecology, 95(8), 2192-2201.
492	Fan, L., Wigneron, J. P., Ciais, P., Chave, J., Brandt, M., Fensholt, R., & Peñuelas, J. (2019).
493	Satellite-observed pantropical carbon dynamics. Nature plants, 5(9), 944-951.
494	Feldpausch, T. R., Phillips, O. L., Brienen, R. J. W., Gloor, E., Lloyd, J., Lopez-Gonzalez, G.,
495	& Vos, V. A. (2016). Amazon forest response to repeated droughts. Global Biogeochemical
496	Cycles, 30(7), 964-982.
497	ForestPlots.net, Blundo, C., Carilla, J., Grau, R., Malizia, A., Malizia, L., & De Araujo, R. O.
498	(2021). Taking the pulse of Earth's tropical forests using networks of highly distributed plots.
499	Biological Conservation, 260, 108849.
500	Galbraith, D., Malhi, Y., Affum-Baffoe, K., Castanho, A. D., Doughty, C. E., Fisher, R. A., &
501	Lloyd, J. (2013). Residence times of woody biomass in tropical forests. Plant Ecology &
502	Diversity, 6(1), 139-157.
503	Gatti, L. V., Basso, L. S., Miller, J. B., Gloor, M., Gatti Domingues, L., Cassol, H. L., & Neves,
504	R. A. (2021). Amazonia as a carbon source linked to deforestation and climate change.
505	Nature, 595(7867), 388-393.
506	Gloor, E., Wilson, C., Chipperfield, M. P., Chevallier, F., Buermann, W., Boesch, H., & Sullivan,
507	M. J. (2018). Tropical land carbon cycle responses to 2015/16 El Niño as recorded by
508	atmospheric greenhouse gas and remote sensing data. Philosophical Transactions of the
509	Royal Society B: Biological Sciences, 373(1760), 20170302.
510	Gianoli, E., & González-Teuber, M. (2005). Environmental heterogeneity and population
511	differentiation in plasticity to drought in Convolvulus chilensis (Convolvulaceae). Evolutionary
512	Ecology, 19(6), 603-613.
513	Girardin, C. A. J., Malhi, Y., Aragao, L. E. O. C., Mamani, M., Huaraca Huasco, W., Durand, L.,
514	& Whittaker, R. J. (2010). Net primary productivity allocation and cycling of carbon along
515	a tropical forest elevational transect in the Peruvian Andes. Global Change Biology, 16(12),
516	3176-3192.
517	Harmon, M. E., Whigham, D. F., Sexton, J., & Olmsted, I. (1995). Decomposition and mass of
518	woody detritus in the dry tropical forests of the northeastern Yucatan Peninsula, Mexico.
519	Biotropica, 305-316.
520	Hughes, I. G., & Hase, T. P. A. (2010). Measurements and their uncertainties: A practical guide
521	to modern error analysis. Oxford: Oxford University Press.
522	Jiménez-Muñoz, J. C., Mattar, C., Barichivich, J., Santamaría-Artigas, A., Takahashi, K., Malhi,
523	Y., & Van Der Schrier, G. (2016). Record-breaking warming and extreme drought in the
524	Amazon rainforest during the course of El Niño 2015–2016. Scientific reports, 6(1), 1-7.





525	Kaewthongrach, R., Chidthaisong, A., Charuchittipan, D., Vitasse, Y., Sanwangsri, M.,
526	Varnakovida, P., & LeClerc, M. Y. (2020). Impacts of a strong El Niño event on leaf
527	phenology and carbon dioxide exchange in a secondary dry dipterocarp forest. Agricultural
528	and Forest Meteorology, 287, 107945.
529	Liu, J., Bowman, K. W., Schimel, D. S., Parazoo, N. C., Jiang, Z., Lee, M., & Eldering, A. (2017).
530	Contrasting carbon cycle responses of the tropical continents to the 2015–2016 El Niño.
531	Science, 358(6360).
532	Luo, X., Keenan, T. F., Fisher, J. B., Jiménez-Munoz, J. C., Chen, J. M., Jiang, C., & Tadić, J.
533	M. (2018). The impact of the 2015/2016 El Niño on global photosynthesis using satellite
534	remote sensing. Philosophical Transactions of the Royal Society B: Biological Sciences,
535	373(1760), 20170409.
536	Machado-Silva, F., Peres, L. F., Gouveia, C. M., Enrich-Prast, A., Peixoto, R. B., Pereira, J. M.,
537	& Libonati, R. (2021). Drought resilience debt drives NPP decline in the Amazon Forest.
538	Global Biogeochemical Cycles, 35(9), e2021GB007004.
539	Malhi, Y., Aragao, L. E. O., Metcalfe, D. B., Paiva, R., Quesada, C. A., Almeida, S., & Teixeira,
540	L. M. (2009). Comprehensive assessment of carbon productivity, allocation and storage in
541	three Amazonian forests. Global Change Biology, 15(5), 1255-1274.
542	Malhi, Y., Doughty, C. E., Goldsmith, G. R., Metcalfe, D. B., Girardin, C. A., Marthews, T. R., &
543	Phillips, O. L. (2015). The linkages between photosynthesis, productivity, growth and
544	biomass in lowland Amazonian forests. Global Change Biology, 21(6), 2283-2295.
545	Malhi, Y., Girardin, C. A., Goldsmith, G. R., Doughty, C. E., Salinas, N., Metcalfe, D. B., &
546	Silman, M. (2017). The variation of productivity and its allocation along a tropical elevation
547	gradient: a whole carbon budget perspective. New Phytologist, 214(3), 1019-1032.
548	Malhi, Y., Girardin, C., Metcalfe, D. B., Doughty, C. E., Aragão, L. E., Rifai, S. W., & Phillips,
549	O. L. (2021). The Global Ecosystems Monitoring network: Monitoring ecosystem productivity
550	and carbon cycling across the tropics. Biological Conservation, 253, 108889.
551	Marimon Junior, B. H., & Haridasan, M. (2005). Comparação da vegetação arbórea e
552	características edáficas de um cerradão e um cerrado sensu stricto em áreas adjacentes
553	sobre solo distrófico no leste de Mato Grosso, Brasil. Acta Botanica Brasilica, 19, 913-926.
554	Marimon, B. S., Lima, E. S., Duarte, T. G., Chieregatto, L. C., & Ratter, J. A. (2006). Observations
555	on the vegetation of Northeastern Mato Grosso, Brazil. IV. An analysis of the Cerrado-
556	Amazonian Forest ecotone. Edinburgh Journal of Botany, 63, 323-341.
557	Marimon, B. S., Marimon-Junior, B. H., Feldpausch, T. R., Oliveira-Santos, C., Mews, H. A.,
558	Lopez-Gonzalez, G., & Phillips, O. L. (2014). Disequilibrium and hyperdynamic tree
559	turnover at the forest-cerrado transition zone in southern Amazonia. Plant Ecology &
560	Diversity, 7(1-2), 281-292.
561	Marques, E. Q., Marimon-Junior, B. H., Marimon, B. S., Matricardi, E. A., Mews, H. A., & Colli, G.
562	R. (2020). Redefining the Cerrado-Amazonia transition: implications for conservation.
563	Biodiversity and conservation, 29(5), 1501-1517.





564	Marthews, T. R., Riutta, T., Oliveras-Menor, I., Urrutia, R., Moore, S., Metcalfe, D., & Cain, R.
565	(2014). Measuring Tropical Forest Carbon Allocation and Cycling: A RAINFOR-GEM Field
566	Manual for Intensive Census Plots (v3.0). Global Ecosystems Monitoring Network, Oxford.
567	Matricardi, E. A. T., Skole, D. L., Costa, O. B., Pedlowski, M. A., Samek, J. H., & Miguel, E. P.
568	(2020). Long-term forest degradation surpasses deforestation in the Brazilian Amazon.
569	Science, 369(6509), 1378-1382.
570	Metcalfe, D. B., Meir, P., Aragao, L. E. O. C., Malhi, Y., Da Costa, A. C. L., Braga, A., & Williams,
571	M. (2007). Factors controlling spatio-temporal variation in carbon dioxide efflux from surface
572	litter, roots, and soil organic matter at four rain forest sites in the eastern Amazon. Journal of
573	Geophysical Research: Biogeosciences, 112(G4).
574	Metcalfe, D. B., Meir, P., Aragão, L. E. O., da Costa, A. C., Braga, A. P., Gonçalves, P. H., &
575	Williams, M. (2008). The effects of water availability on root growth and morphology in an
576	Amazon rainforest. Plant and Soil, 311(1), 189-199.
577	Miller, J. B. (1967). A formula for average foliage density. Australian Journal of Botany, 15, 141-
578	144.
579	Miranda, S. D. C., Bustamante, M., Palace, M., Hagen, S., Keller, M., & Ferreira, L. G. (2014).
580	Regional variations in biomass distribution in Brazilian savanna woodland. Biotropica, 46(2),
581	125-138.
582	Moore, S., Adu-Bredu, S., Duah-Gyamfi, A., Addo-Danso, S. D., Ibrahim, F., Mbou, A. T., &
583	Malhi, Y. (2018). Forest biomass, productivity and carbon cycling along a rainfall gradient in
584	West Africa. Global change biology, 24(2), e496-e510.
585	Morandi, P. S., Marimon-Junior, B. H., De Oliveira, E. A., Reis, S. M., Valadão, M. X., Forsthofer,
586	M., & Marimon, B. S. (2015). Vegetation succession in the Cerrado-Amazonian forest
587	transition zone of Mato Grosso state, Brazil. Edinburgh Journal of Botany, 73(1), 83-93.
588	Morandi, P. S., Marimon, B. S., Eisenlohr, P. V., Marimon-Junior, B. H., Oliveira-Santos, C.,
589	Feldpausch, T. R., & Phillips, O. L. (2016). Patterns of tree species composition at
590	watershed-scale in the Amazon 'arc of deforestation': implications for conservation.
591	Environmental Conservation, 43(4), 317-326.
592	Neyret, M., Bentley, L. P., Oliveras, I., Marimon, B. S., Marimon-Junior, B. H., Almeida de Oliveira,
593	E., & Malhi, Y. (2016). Examining variation in the leaf mass per area of dominant species
594	across two contrasting tropical gradients in light of community assembly. Ecology and
595	evolution, 6(16), 5674-5689.
596	Nobis, M., & Hunziker, U. (2005). Automatic thresholding for hemispherical canopy-photographs
597	based on edge detection. Agricultural and forest meteorology, 128(3-4), 243-250.
598	Nogueira, D. S., Marimon, B. S., Marimon-Junior, B. H., Oliveira, E. A., Morandi, P., Reis, S. M.,
599	& Phillips, O. L. (2019). Impacts of Fire on Forest Biomass Dynamics at the Southern
600	Amazon Edge. Environ. Conserv. 46, 285-292.
601	Oliveira, B., Marimon Junior, B. H., Mews, H. A., Valadão, M. B. X., & Marimon, B. S. (2017).
602	Unraveling the ecosystem functions in the Amazonia-Cerrado transition: evidence of
603	hyperdynamic nutrient cycling. Plant Ecology, 218(2), 225-239.





604	Oliveras, I., & Malhi, Y. (2016). Many shades of green: the dynamic tropical forest-savannah
605	transition zones. Philosophical Transactions of the Royal Society B: Biological Sciences,
606	371(1703), 20150308.
607	Palmer, P. I. (2018). The role of satellite observations in understanding the impact of El Nino on
608	the carbon cycle: current capabilities and future opportunities. Philosophical Transactions of
609	the Royal Society B: Biological Sciences, 373(1760), 20170407.
610	Peixoto, K. S., Marimon-Junior, B. H., Marimon, B. S., Elias, F., de Farias, J., Freitag, R., &
611	Malhi, Y. (2017). Unravelling ecosystem functions at the Amazonia-Cerrado transition: II.
612	Carbon stocks and CO2 soil efflux in cerradao forest undergoing ecological succession. Acta
613	oecologica, 82, 23-31.
614	Peixoto, K. D. S., Marimon-Junior, B. H., Cavalheiro, K. A., Silva, N. A., das Neves, E. C., Freitag,
615	R., & Valadao, M. B. X. (2018). Assessing the effects of rainfall reduction on litterfall and
616	the litter layer in phytophysiognomies of the Amazonia–Cerrado transition. Brazilian Journal
617	of Botany, 41(3), 589-600.
618	Phillips, O. L., Aragão, L. E., Lewis, S. L., Fisher, J. B., Lloyd, J., López-González, G., & Torres-
619	Lezama, A. (2009). Drought sensitivity of the Amazon rainforest. Science, 323(5919), 1344-
620	1347.
621	Ratter, J. A., Richards, P. W., Argent, G., & Gifford, D. R. (1973). Observations on the vegetation
622	of northeastern Mato Grosso: I. The woody vegetation types of the Xavantina-Cachimbo
623	Expedition area. Philosophical Transactions of the Royal Society of London. B, Biological
624	Sciences, 266(880), 449-492.
625	Reis, S. M., Lenza, E., Marimon, B. S., Gomes, L., Forsthofer, M., Morandi, P. S., & Elias, F.
626	(2015). Post-fire dynamics of the woody vegetation of a savanna forest (Cerradão) in the
627	Cerrado-Amazon transition zone. Acta Botanica Brasilica, 29, 408-416.
628	Reis, S. M., de Oliveira, E. A., Elias, F., Gomes, L., Morandi, P. S., Marimon, B. S., & Lenza,
629	E. (2017). Resistance to fire and the resilience of the woody vegetation of the "Cerradão" in
630	the "Cerrado"–Amazon transition zone. Brazilian Journal of Botany, 40(1), 193-201.
631	Reis, S. M., Marimon, B. S., Marimon Junior, B. H., Morandi, P. S., Oliveira, E. A. D., Elias, F.,
632	& Phillips, O. L. (2018). Climate and fragmentation affect forest structure at the southern
633	border of Amazonia. Plant Ecology & Diversity, 11(1), 13-25.
634	Reis, S. M., Marimon, B. S., Esquivel-Muelbert, A., Marimon Jr, B. H., Morandi, P. S., Elias, F.,
635	& Phillips, O. L. (2022). Climate and crown damage drive tree mortality in southern
636	Amazonian edge forests. Journal of Ecology, 110(4), 876-888.
637	Reis, S. (2023). Savanna is more resistant and resilient to tropical drought than transitional forest.
638	Dryad Dataset: https://doi.org/10.5061/dryad.rjdfn2zhw
639	Rezende, A. V., Vale, A. D., Sanquetta, C. R., Figueiredo Filho, A., & Felfili, J. M. (2006).
640	Comparação de modelos matemáticos para estimativa do volume, biomassa e estoque de
641	carbono da vegetação lenhosa de um cerrado sensu stricto em Brasília, DF. Scientia
642	Forestalis, 71(2), 65-73.





643	Ribeiro, J. F., Walter, B. M. T. (2008). As principais fitofisionomias do bioma Cerrado. Cerrado
644	Ecologia e Fauna. Brasília: Embrapa Informação Tecnológico 1, 153 – 221.
645	Rifai, S. W., Girardin, C. A., Berenguer, E., del Aguila-Pasquel, J., Dahlsjö, C. A., Doughty, C. E.,
646	& Malhi, Y. (2018). ENSO Drives interannual variation of forest woody growth across the
647	tropics. Philosophical Transactions of the Royal Society B: Biological Sciences, 373(1760),
648	20170410.
649	Riutta, T., Malhi, Y., Kho, L. K., Marthews, T. R., Huaraca Huasco, W., Khoo, M., & Ewers, R.
650	M. (2018). Logging disturbance shifts net primary productivity and its allocation in Bornean
651	tropical forests. Global Change Biology, 24(7), 2913-2928.
652	Schleppi, P., Conedera, M., Sedivy, I., & Thimonier, A. (2007). Correcting non-linearity and slope
653	effects in the estimation of the leaf area index of forests from hemispherical photographs.
654	Agricultural and Forest Meteorology, 144(3-4), 236-242.
655	Silvério, D. V., Brando, P. M., Bustamante, M. M., Putz, F. E., Marra, D. M., Levick, S. R., &
656	Trumbore, S. E. (2019). Fire, fragmentation, and windstorms: A recipe for tropical forest
657	degradation. Journal of Ecology, 107(2), 656-667.
658	Soares Jancoski, H., Schwantes Marimon, B., C. Scalon, M., de V. Barros, F., Marimon-Junior,
659	B. H., Carvalho, E., & Oliveras Menor, I. (2022). Distinct leaf water potential regulation of
660	tree species and vegetation types across the Cerrado-Amazonia transition. Biotropica,
661	54(2), 431-443.
662	
663	
664	
665	
666	
667	
668	
669	
670	
671	
672	
673	
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