

**Anatomical distribution of starch in the stemwood influences carbon dynamics and suggests storage-growth trade-offs in some tropical trees**

David Herrera-Ramírez<sup>1\*</sup>, Henrik Hartmann<sup>1</sup>, Christine Römermann<sup>2,3</sup>, Susan Trumbore<sup>1</sup>, Jan Muhr<sup>1,4</sup>, Leonardo Maracahipes-Santos<sup>7</sup>, Paulo Brando<sup>5,6,7</sup>, Divino Silvério<sup>8</sup>, Huang Jianbei<sup>1</sup>, Iris Kuhlmann<sup>1</sup> and Carlos A. Sierra<sup>1</sup>

<sup>1</sup>Max Planck Institute for Biogeochemistry, Hans-Knöll-Str 10, 07745 Jena

<sup>2</sup>Friedrich Schiller University Jena, Institute for Ecology and Evolution, Philosophenweg 16, 07743, Jena, Germany

<sup>3</sup>German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig, D- 04103 Leipzig

<sup>4</sup>Georg August University Göttingen, Department of Bioclimatology, Büsgenweg 2, 37077 Göttingen, Germany

<sup>5</sup>Department of Earth System Science, University of California, Irvine, CA 92697, USA

<sup>6</sup>Yale School of the Environment, Yale University, New Haven, CT, United States.

<sup>7</sup>Instituto de Pesquisa Ambiental da Amazônia, Brasília, DF 70863-520, Brazil

<sup>8</sup>Department of Biology, Universidade Federal Rural da Amazônia-UFRA, Capitão Poço 68650-000, Pará, Brazil

**Abstract**

1. Trees balance temporal asynchrony in carbon source and sink activity by accumulating and using non-structural carbon (NSC). Previous work has demonstrated differences in the amount and distribution of NSC stored in stemwood in tropical tree species and related these patterns in NSC distribution to tree growth and mortality rates. However, we still do not know how changes in the amount and location of starch, a major component of NSC in stemwood, influence the seasonal carbon dynamics of mature trees and how this may reflect storage-growth trade-offs.
2. In this work, we hypothesized that combining two life history traits, here leaf habit (evergreen/semi-deciduous) and the anatomical distribution of starch within the stemwood (parenchyma-storage and fiber-storage), would allow us to explain differences in the seasonal interplay between carbon sources and sinks and the use and accumulation of starch in the tree stem. We expected semi-deciduous/fiber-storing species to have greater seasonal amplitudes of carbon source and sink activity, and therefore greater variation in starch content and stronger storage-growth trade-offs than evergreen/parenchyma-storing species.
3. We measured monthly increments in stem radial growth, soluble sugars, and starch every three months during 2019 in *Dacryodes microcarpa* (semi-deciduous/fiber-

storing-species), *Ocotea leucoxylon* (evergreen/parenchyma-storing-species), and *Sacoglottis guianensis* (semi-deciduous/parenchyma-storing-species).

4. We found seasonal changes in starch but not sugars in the semi-deciduous species, with greater amplitude in the fiber-storing species that also had greater storage capacity and stem respiration rates. The fiber-storing species further showed a negative relationship between starch consumption/accumulation and growth during the rainy season, suggesting a trade-off between growth and storage, with starch accumulating in some cases at the expense of growth.
5. **Synthesis:** Our results show the influence of seasonal starch storage on carbon dynamics in three species of tropical trees that differ in leaf phenology and starch storage traits. Semi-deciduous/fiber-storing species have greater temporal variation in carbon sink activities and more seasonally dynamic starch content. Since fiber-storing species tend to be slower-growing and longer-lived trees with lower mortality rates compared to parenchyma-storing species, these results may provide clues about how storage traits could influence their survival and life span.

## 1. Introduction

Trees assimilate CO<sub>2</sub> through photosynthesis to produce non-structural carbon (NSC), consisting mainly of soluble sugars, starch, and lipids. Sugars are transported to all tree organs to fuel metabolism and growth, and are locally accumulated as starch and lipids to provide energy reserves on different time scales, for example, daily, seasonal, or interannual (Chapin et al., 1990; Richardson et al., 2013). Seasonal changes in NSC content in different tree organs reflect imbalances between carbon sources (e.g., photosynthesis) and sinks (e.g., growth and respiration) (Kozlowski, 1992; Körner, 2003). Older stored carbon has been shown to fuel metabolic functions when disturbances such as defoliation or drought interrupt C supply (Hartmann and Trumbore, 2016). Information about how trees accumulate and use their NSC in key organs, such as stemwood, is needed to improve our understanding of how trees maintain their metabolism under stress.

Stemwood plays a central role in NSC dynamics and long-term storage in adult trees (Arx et al., 2017; Furze et al., 2018, 2020; Herrera-Ramírez et al., 2021). It could be the largest reservoir of NSC in trees in terms of mass due to its large volume. However, stems are also important in the transport of NSC and water between tree organs (Plavcová et al., 2016;

71 Furze et al., 2018). Wood is always in contact with the secondary phloem, the main tissue  
72 transporting NSC from leaves to roots and in itself an important NSC storage tissue (Rosell et  
73 al., 2021). There is constant mixing of NSC between the secondary phloem and adjacent wood  
74 tissue, allowing trees to store NSC for a long time (Furze et al., 2018, Peltier et al., 2023).

75 Wood anatomical traits can influence the way trees access or accumulate carbon reserves  
76 (Barbaroux and Bréda, 2002; Furze et al., 2020; Herrera-Ramírez et al., 2021). For instance,  
77 in temperate trees, traits such as vessel distribution influence not only the NSC distribution in  
78 wood but also its seasonal dynamics (Barbaroux and Bréda, 2002; Michelot et al., 2012; Furze  
79 et al., 2020). Other traits such as the formation of living fibers in stemwood (septate fibers  
80 that remain alive after maturation), or the abundance, size, and longevity of parenchyma  
81 cells, may also influence the seasonal response of NSC, overall metabolic rates, and may even  
82 be related to tree longevity or mortality rates (Arx et al., 2017; Herrera-Ramírez et al., 2021).

83 Here, we define two starch storage traits based on the anatomical distribution of starch  
84 storages in the stemwood: 1) trees that only store starch in the parenchyma, which we refer  
85 to as parenchyma-storing species; and 2) trees that store starch in the parenchyma and living  
86 fibers, which we refer to as fiber-storing species (Herrera-Ramirez et al., 2021, Fig. S1).

87 Herrera-Ramirez et al. (2021) demonstrated that these storage traits were related to stem  
88 growth and mortality rates for the species involved. Thus, further understanding of how this  
89 anatomical distribution of starch within the stemwood is related to the metabolism of trees  
90 and the dynamics of NSC would improve our mechanistic understanding of how tropical trees  
91 regulate carbon storage to increase competitiveness and/or survival.

92 On seasonal to interannual timescales, non-structural carbon storage dynamics can be  
93 controlled by physiological processes related to carbon source or sink activity under the  
94 influence of environmental conditions (Würth et al., 2005; Palacio et al., 2014; O'Brien et al.,

2015, 2020; Resco de Dios and Gessler, 2021). However, NSC storage processes can be up-regulated by increased gene expression, so reserve formation competes with other carbon sinks (Wiley and Helliker, 2012; Martínez-Vilalta, 2014; Huang et al., 2021). This may allow trees to maintain carbon stores that enable them to face future compounding perturbations (Sala et al., 2012; Dietze et al., 2014; Resco de Dios and Gessler, 2021). Storage up-regulation may be species specific (Poorter and Kitajima, 2007) and likely greater in species that store NSC for long periods of time (Blumstein et al., 2022). Tree species that rely on living fibers for NSC storage (e.g., fiber-storing species) may indicate a high priority for storage formation that would compete with other carbon sinks like growth and respiration and may be related to higher plasticity of carbon metabolism, such as higher capacity for variability of seasonal carbon fluxes (Plavcová et al., 2016; Herrera-Ramírez et al., 2021). It is possible that trees with larger plasticity in carbon storage and sink fluxes may be better adapted to stressful conditions that severely reduce photosynthesis or increase carbon demand because they may be more tolerant to a wider range of environmental conditions.

When C sources are insufficient, regulation of storage and sink activity can result in trade-offs between NSC storage and some important carbon sinks such as growth, respiration, reproduction, and production of defense compounds (Poorter and Kitajima, 2007). These trade-offs may indicate different plant survival strategies, ranging from long-lived species that may prioritize NSC storage at the expense of growth to ensure future survival, to short-lived species that invest in fast growth while storing less NSC (Wright et al., 2004; O'Brien et al., 2014; Blumstein et al., 2022). Trade-offs between NSC storage, growth, and defense have been observed in some temperate species. For example, studies have shown that the storage of NSC may be prioritized over growth under various stressful conditions such as defoliation, drought, and low atmospheric CO<sub>2</sub> concentrations (Anderegg et al., 2012; Hartmann et al.,

119 2015; Piper et al., 2015; Wiley et al., 2017; Huang et al., 2019; Piper and Paula, 2020).

120 Allocation to defense compounds can be reduced to maintain the minimum operational  
121 storage of NSC required for survival under shade and low CO<sub>2</sub> (Huang et al., 2019; Huang et  
122 al., 2020). However, NSC can also be accumulated to build defenses for the future, and  
123 therefore such trade-offs may not be immediate and exhibit time lags. Thus, in many  
124 instances, trade-offs may not be clear, especially if they are measured on an annual basis.

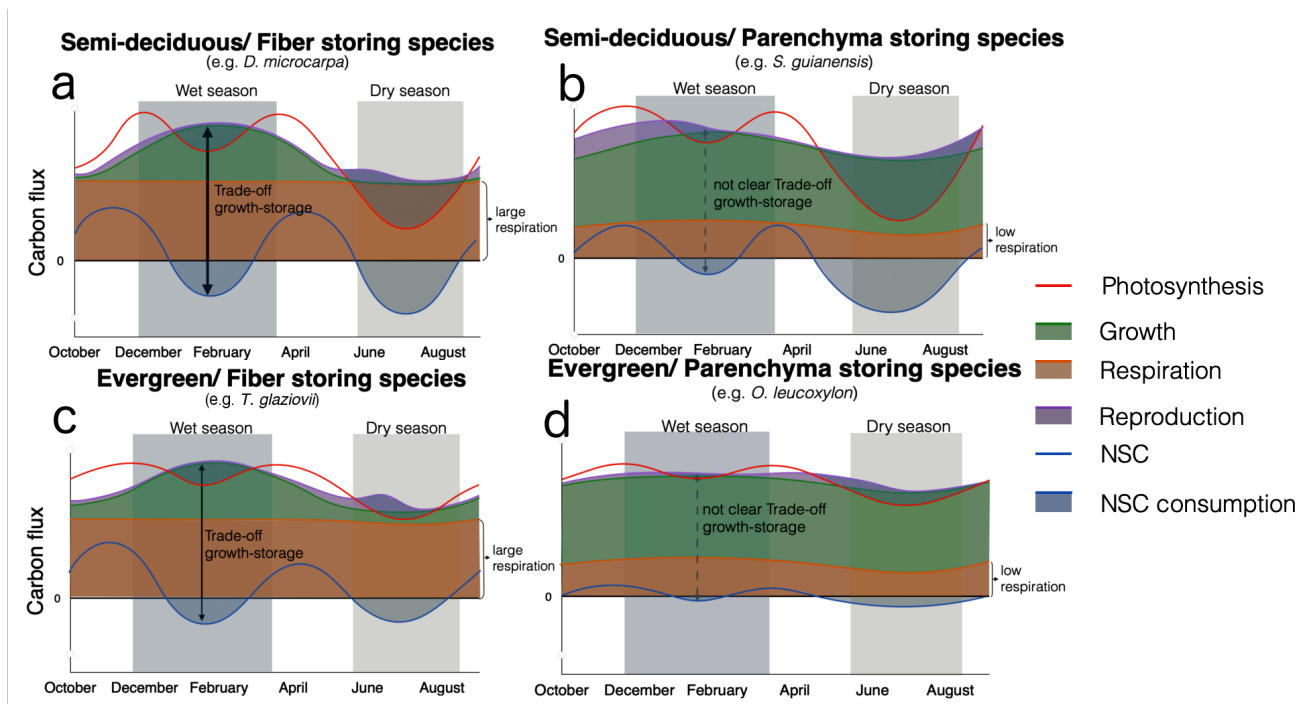
125 In some cases, identifying growth-storage trade-offs can be difficult due to variability in  
126 carbon assimilation or other competing carbon sinks such as respiration and reproduction  
127 (Wiley and Helliker, 2012; Huang et al., 2019; Blumstein et al., 2022). For example, as carbon  
128 supply increases, both growth rate and NSC content may increase, masking trade-offs  
129 between the two competing sinks (Huang et al., 2019; Blumstein et al., 2022). Therefore,  
130 alternative ways to quantify these trade-offs, such as evaluating growth and storage fluxes on  
131 a seasonal basis, are needed to better understand carbon dynamics in mature trees.

132 Furthermore, investigating the relationship between these trade-offs and tree storage traits  
133 may help us identify the influence of storage strategies on carbon dynamics in mature trees  
134 and their relationship with tree survival and recovery.

135 In this context, we aim to understand the influence of life history traits such as the anatomical  
136 distribution of starch storage in the stemwood (fiber-storage and parenchyma-storage, Fig.  
137 S1) and leaf habit (evergreen and semi-deciduous) on seasonal carbon dynamics of trees in  
138 response to seasonal changes in precipitation and relative humidity. For this, we propose a  
139 conceptual framework to help formulate hypotheses about the seasonal behavior of the  
140 observed carbon fluxes in four groups of trees based on the possible combinations of their life  
141 history traits (Fig. 1). Carbon fluxes were estimated based on our observations of growth,  
142 respiration, and leaf and fruit phenology collected during two years for species

143 representative of each combination of traits. We used phenological data on the percent cover  
144 of mature leaves in the tree crown as a proxy for carbon acquisition. Thus, we expect  
145 evergreen species to have less seasonal variations in photosynthesis rates than semi-  
146 deciduous species. Although all four trait combinations are given in Figure 1, given the great  
147 effort needed for seasonal analyses, we left one trait combination (evergreen/fiber-storing-  
148 species) out of our sampling because previous results suggested that leaf habit did not  
149 influence NSC storage, growth, or mortality in the fiber-storing species (see Herrera-Ramírez  
150 et al., 2021). Thus, we focused only on three combinations of traits (evergreen/parenchyma-  
151 storage, semi-deciduous/parenchyma-storage and semi-deciduous/fiber-storage) in this  
152 study (Fig. 1).

153 Based on the conceptual framework (Fig. 1), we expected that: i) semi-deciduous/fiber-  
154 storing species would have greater seasonal amplitude in stemwood NSC due to greater  
155 seasonal variation in photosynthesis, greater demand of carbon for respiration due to a larger  
156 amount of living cells, and greater seasonal variation in growth rates than the parenchyma-  
157 storing species; ii) negative carbon balance during the wet season would lead to some  
158 consumption of starch reserves, principally because there may be a slight decrease in  
159 photosynthetic capacity (due to increased cloudiness or loss of photosynthetic tissue)  
160 coincident with a large demand of carbon for growth and respiration; in which case iii) we  
161 would expect to observe seasonal storage-growth trade-offs during the wet season where  
162 growth would be negatively correlated with starch consumption/accumulation. In contrast,  
163 we expect a smaller seasonal variation in NSC for parenchyma-storing species



164

165 **Figure 1:** Conceptual framework illustrating temporal patterns of carbon fluxes and their  
 166 intraannual interactions for trees with different combinations of two life history traits: leaf  
 167 habit (semi-deciduous and evergreen) and anatomical distribution of starch storage in the  
 168 stemwood (Fiber storage and parenchyma storage) for a) semi-deciduous/fiber-storing-  
 169 species, b) semi-deciduous/parenchyma-storing-species, c) evergreen/fiber-storing  
 170 species and d) evergreen/parenchyma-storing-species). Carbon fluxes measured over a  
 171 year for each of these species were used to inform the seasonal dynamics of carbon  
 172 sources - photosynthesis (red line)- and carbon sinks - growth (green), respiration  
 173 (orange), and reproduction (purple) - along with their hypothesized interactions with non-  
 174 structural carbohydrates (NSC) storage or consumption (blue line, representing the  
 175 balance between previous storage, sources, and sinks). The patterns of photosynthesis and  
 176 reproduction were deduced from the phenological observations of mature leaves in the  
 177 crown of trees belonging to each of the representative species (see phenology data in Fig.  
 178 S3). The black horizontal solid line represents zero flux for NSC, respiration, and  
 179 photosynthesis. C sinks are represented as negative fluxes; The shaded colored area

180 beneath each carbon sink represents its contribution to the total sink flux. This conceptual  
181 (but data-informed) representation illustrates general trends. Expected seasonal trade-offs  
182 between growth and storage, manifesting as negative correlations between growth (green  
183 line) and storage (blue line), are indicated by black arrows, thicker and darker lines  
184 indicate strong (clearly observable) trade-offs, and dashed and lighter lines indicate  
185 weaker or less clear trade-offs. Under each trait combination label, we mention the  
186 representative species we measured in this study. We did not measure the representative  
187 evergreen/fiber-storing species *Trattinnikia glaziovii*, but we used previous data to inform  
188 our framework.

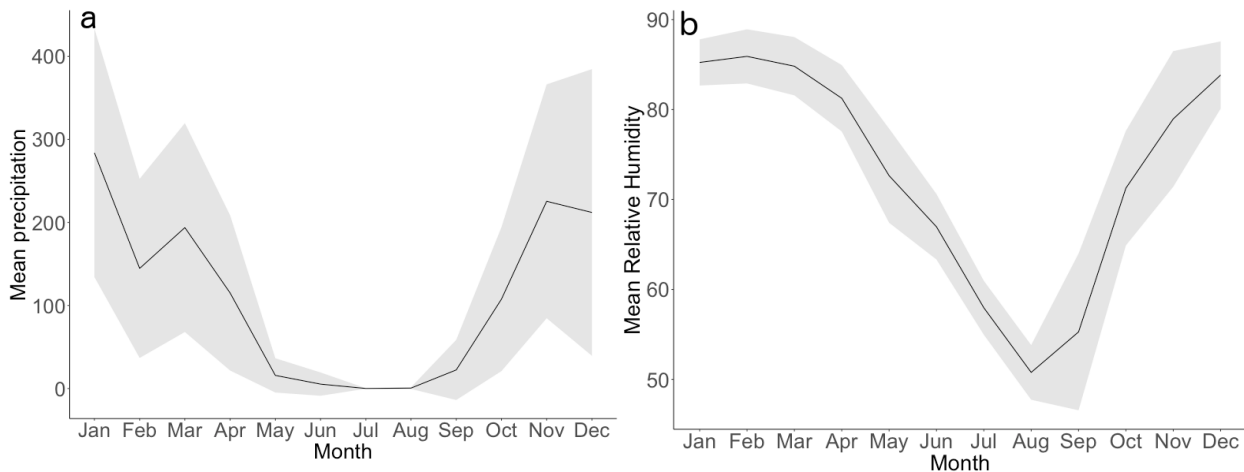
## 189 **2. Methods**

### 190 **2.1. Site description**

191 This study was conducted in a transitional forest between the closed canopy Amazon  
192 rainforest and Cerrado (savanna), located at Tanguro Ranch, Mato Grosso, Brazil. The mean  
193 annual precipitation is 1770 mm, distributed between a dry season (May to September) with  
194 less than 10 mm of precipitation per month and a wet season (October to April) with mean  
195 precipitation of 150 mm per month (Fig. 2). Relative humidity follows a similar seasonal  
196 pattern, falling below 60% from June to September and above 80% from December to  
197 February (Fig. 2). The mean temperature is 25°C with almost no seasonal variation  
198 throughout the year (data obtained from the local station at Tanguro Ranch, operated by the  
199 Instituto de Pesquisa Ambiental da Amazônia, IPAM).

200





**Figure 2:** a) Mean seasonal course of monthly total precipitation (mm) and b) relative humidity (%) during 2016-2020 (black lines). Gray areas correspond to the standard deviation. Data were obtained from a local weather station at Tanguro Ranch (S 13°4'35.39", W 52°23' 8.85").

## 2.2. Species description

We selected three tree species previously shown to have a contrasting anatomical distribution of starch storage in the stemwood: parenchyma-storing species and fiber-storing species (Herrera-Ramírez et al., 2021, Fig. S1), different leaf habits (evergreen or semi-deciduous), and different growth and mortality rates (Table 1). Based on inventory data, these species are among the top 10 % of the most dominant species in this forest. From each tree species, we chose 12 mature and healthy trees that reached the canopy with a diameter at breast height (at 1.3 m, dbh) greater than 20 cm (Table 1), to ensure a representative sample for each species.

**Table 1:** Species names and traits: wood storage strategy, growth rates, mortality rates, phenology, and sampling dates.

Species name	Growth rate (cm/year)	Mortality rate (%/year)	Anatomical	
			distribution c starch	Leaf phenology (~% of annual leaf loss)
<i>Ocotea leucoxylon</i> (Sw.) Laness	0.295	5.7	Parenchyma	Evergreen (~30)
<i>Sacoglottis guianensis</i> Benth.	0.72	5.0	Parenchyma	Semi-deciduous (~45)
<i>Dacryodes microcarpa</i> Cuart.	0.078	1.6	Fiber	Semi-deciduous (~60)

\*Growth rate, Mortality rates, and Leaf phenology were calculated based on inventories conducted between 2004 and 2018 and included more than 100 individuals per species.

### 2.3. Sampling strategy

We sampled the 36 selected trees every three months from May 2019 to February 2020, specifically in May 2019 (transition from wet to dry season), August 2019 (dry season), November 2019 (transition from dry to wet season), and February 2020 (wet season). During each field campaign, we took two wood cores (~ 5mm diameter and 20 cm long) approximately 10 cm from the previous collection point. We used one wood core to quantify the starch distribution along the radial axis from bark to pith using the histological quantification method described in Herrera-Ramírez et al. (2021). The wood cores were placed on ice immediately after collection and frozen at -18°C within two hours to stop respiration. After freezing, they were dried at 60°C for two days.

The second wood core was used for incubation to estimate wood respiration rates. These cores were placed in a wet tissue immediately after collection and kept away from direct

233 sunlight exposure. The cores were placed in the incubation chambers within two hours after  
234 collection.

#### 235 **2.4. NSC seasonality**

236 We quantified soluble sugars with a High-Performance Anion Exchange Chromatography  
237 with Pulsed Amperometric Detection (HPAE-PAD) following the protocols of Landhäusser et  
238 al. (2018). We quantified starch according to the histological method proposed by Herrera-  
239 Ramirez et al. (2021). This method allows us to observe and quantify spatial patterns of  
240 starch distribution in the stemwood with high resolution and identify the cell types that were  
241 used to store starch.

242 To measure the concentration of sucrose, fructose, and glucose (soluble sugars), with the  
243 HPAE-PAD (Landhäusser et al., 2018), we segmented the wood core into two depth ranges: 0-  
244 2 cm and 2-4 cm. These measurements were made only for 5 individuals per species and for  
245 samples taken in January (wet season) and July 2018 (dry season). Each wood core segment  
246 was ground to a fine powder using a ball mill (Retsch MM 400, Haan, Germany) at 25 Hz for  
247 30 seconds and dried at 60°C overnight. We weighed 50 mg of wood powder and mixed it  
248 with 1.5 ml of 80 % ethanol for 10 minutes at 90°C to extract the soluble sugars. After cooling  
249 to room temperature, the samples were centrifuged at 13000 g for 2 minutes. The  
250 supernatant was recovered in a new vial, diluted, filtered, and then used for measuring the  
251 concentration of three specific sugars (sucrose, fructose, and glucose) by HPAE-PAD. For  
252 quality control, we used blanks, standard solutions with known concentrations of glucose,  
253 fructose, sucrose, and internal standards made of a mixture of tree leaf and branch samples as  
254 detailed in the protocol "S3" of Landhäusser et al. (2018). We estimated the average  
255 concentration of soluble sugars per wood core and compared them between time points. The  
256 concentration of soluble sugars was always very low (up to 2 %) for all species and we did

not find significant differences ( $p > 0.05$  from a Wilcoxon signed rank test) between the wet and dry season of 2018 (Fig. S2). Therefore, soluble sugars were not measured for 2019. For the species we analyzed, starch represented 70, 85, and 90 % of the stored NSC for *D. microcarpa*, *S. guianensis*, and *O. leucoxylon*, respectively. Therefore, we assumed that NSC dynamics are mainly represented by starch changes over time.

To quantify the starch concentration using the histological method, we took 30  $\mu\text{m}$  thick slices and mounted them on a glass slide. The wood surface was covered with Lugol's iodine solution for 3 minutes to stain starch grains and then covered with a coverslip. The stained samples were photographed using an optical digital microscope with a large depth-of-field (Keyence, VHX-6000, USA) within 3 hours. Panoramic images of the sample were taken at 300x magnification.

We quantified the percentage of starch in the samples by repeatedly measuring the areal percentage of starch coverage for each 1  $\text{mm}^2$  area over the wood surface using the Image J software (Schneider et al., 2012). We divided the images into radial increments of 5-mm from bark to pith to measure the radial distribution of starch in the wood core. We measured starch in all 5-mm increment sections until no starch was found in the wood. While doing so, we also manually eliminated artifacts from the images that may have interfered with the quantification. After preparing the images, we ran an automatic script for identifying and quantifying starch grains in multiple 1  $\text{mm}^2$  regions of interest (ROI), randomly selected along the images (see the Supplementary Material Methods S2 in Herrera-Ramírez et al. (2021) for details). After identifying all starch grains in an ROI, the script calculates the percentage of the surface area covered by starch. We measured 50 ROIs in each image of 5 mm segments of the stem increment core from bark to pith. We took the average of these 50 ROI measurements as an estimate of the percentage of starch for each of the 5 mm sections of

281 the wood increment core. Measurements of the areal percentage covered by starch  
282 approximate the concentration of starch per gram of dry wood, explaining 80% of the  
283 variability of starch concentrations measured by the HPAE-PAD and getting close to the 1:1  
284 ratio with a regression slope of 0.89 (  $p < 0.01$ , Herrera-Ramírez et al., 2021). Finally, we  
285 estimated the total starch mass of starch in the entire wood core (in grams) by integrating the  
286 starch mass in each 5-mm segment along the radial path of the wood core from bark to pith.  
287 The starch content (grams of starch/grams of wood) at each depth was calibrated using the  
288 measured percentage of starch in the histological images. The starch mass at each wood  
289 depth was estimated based on the species' wood density, segment volume, and the starch  
290 content at each 5-mm of wood section.

291 We compared total starch mass and total soluble sugar concentrations in the entire wood  
292 core between different sampling dates using the Wilcoxon signed-rank test with a 95%  
293 confidence level, and adjusting the p values following the Holm method (Holm 1979). This  
294 non-parametric test was chosen due to its robustness against potential outliers, the flexibility  
295 to analyze samples coming from distinct density distributions, and its suitability for small  
296 sample sizes (Hollander et al., 2013). This allowed us to identify the seasonal patterns of  
297 starch mass and soluble sugars in the entire wood core of our evaluated species during 2019.

298 We estimated the seasonal amplitude of the starch concentration as the difference between  
299 the maximum and minimum starch concentrations measured during 2019 across all wood  
300 depths. The maximum concentration corresponds to the peak of the starch accumulation  
301 period, while the minimum concentration represents the period of starch depletion. We used  
302 analysis of variance (ANOVA) to evaluate differences in these starch changes between  
303 different depths to determine at which wood depths the greatest changes in starch  
304 concentration occurred.

305 We estimated the seasonal changes in the total starch mass (mg) in the entire wood core  
 306 between two consecutive sampling periods as the difference between the starch mass  
 307 between the two respective measurements For example, the starch change observed in the  
 308 dry season was the difference in the starch mass between May 2019 and August 2019 and we  
 309 denominated it May19-Aug19; Similarly, the starch change observed in the transition months  
 310 from dry to wet season was the difference in starch mass between August 2019 and  
 311 November 2019 (Aug19-Nov19); and the starch change observed during the wet season  
 312 constituted the difference in starch mass between November 2019 and February 2020  
 313 (Nov19-Feb20). Unfortunately, due to the COVID-19 pandemic, we could not travel to the  
 314 field site in May 2020 to observe the change in starch mass during the transition months from  
 315 the wet to the dry season (February 2020 to May 2020) and close the year of observations.

316 We also estimated the relative change of the total starch mass of the entire wood core  
 317 between seasons, by dividing the absolute starch mass change by the starch mass in the final  
 318 month (Equation 1).

$$319 \quad relative\_starch\_change = \frac{Starch_{mass}(t_2) - Starch_{mass}(t_1)}{Starch_{mass}(t_2)} \quad (Equation 1)$$

320 Where  $t_2$  is chronologically the second time point between seasons, and  $t_1$  is the first time  
 321 point.

322 We evaluated whether the changes in starch mass between seasons were different than zero  
 323 by building 95% non-parametric confidence intervals for the mean using the adjusted  
 324 bootstrap percentile (BCa) interval method (Davison and Hinkley 1997). For this purpose, we  
 325 used the 'boot' package available in R (Canty and Ripley 2022). We use these intervals to  
 326 graphically show differences between starch storage changes between all possible periods of

2019, within and between species. These graphical comparisons were also confirmed by a Wilcoxon rank test following the Holm method for adjusting p values (Holm 1979).

## **2.5. Sink activity measurements**

We measured monthly stem growth with manual dendrometer bands (D1, Labcell Ltd, UK). These dendrometers were installed at breast height for each tree in July 2018, and measurements were collected monthly (by manually reading) until July 2020. We used three-month moving averages of growth rates, which we calculated to reduce the effect of water dynamics in our growth data, capturing the average intra-annual radial growth variations (Zweifel et al., 2006, 2016). From these data, we calculated the annual growth for each tree during 2019 and compared it with the mean annual starch mass and the changes in the starch mass every three months. To assess the significance of variations in monthly radial growth rates within each species, we used Wilcoxon signed-rank tests with a 95% confidence level. This test compared the growth rates between individual months. To ensure the robustness of the results, p-values were adjusted following the Holm method (Holm 1979).

To evaluate seasonal trade-offs between growth and storage, we calculated seasonally three-month cumulative growth, each corresponding to the core of the wet and dry season and the two transition periods. We estimated Pearson's correlations and linear regressions between the cumulated growth and starch mass changes for each season. The heteroscedasticity of the residuals was checked by plotting the fitted values of the model against the residuals, while the normality assumption was checked using a Q-Q plot (Fox 2015).

We measured wood respiration during the wet (May) and dry (August) seasons of 2019 by incubation of stem cores taken from each tree for 36 hours following collection. The wood cores were cut at the depth in the stem we had previously identified where starch is depleted

350 (6 cm for *O. leucoxyton*, 8 cm for *D. microcarpa*, and 12 cm for *S. guianensis*). The cores were  
351 then sealed in cylindrical chambers and incubated for 36 hours at ambient temperature (~25  
352 °C). The CO<sub>2</sub> produced was collected in custom-made glass flasks. We purified the total  
353 amount of collected CO<sub>2</sub> after cryogenic separation on a vacuum line. We estimated the total  
354 sample mass by measuring the pressure at room temperature together with the flask volume  
355 (Muhr et al., 2018). Then we calculated the volume-specific respiration rate for each sample  
356 by dividing the total amount of collected CO<sub>2</sub> (in mg C) by the incubation time and wood  
357 volume of the incubated core segment. We used Wilcoxon rank tests to compare respiration  
358 between seasons and between species.

359 We used previous observations of leaf, flower, and fruit phenology patterns to estimate other  
360 carbon source and sink fluxes such as photosynthesis and reproduction. These observations  
361 were made for each species at this location over the last 8 years. The percentage of crown  
362 coverage of young leaves, mature leaves, flowers, and fruits has been recorded monthly from  
363 selected species. The patterns were classified into 5 categories (0, 25, 50, 75, and 100%),  
364 following the Fournier method (Fournier 1974). We used these data to estimate seasonal  
365 changes in the phenological data by fitting a smoothing spline model to the monthly data (Fig.  
366 S3). We used these phenological data to give an idea of the contribution of flowering and  
367 fruiting to the seasonal carbon sink fluxes, but we did not measure the amount of carbon  
368 allocated to these fluxes for individual trees or species. Then in general we assume a higher  
369 flux of carbon to reproduction whenever there is more presence of fruiting and flowering. We  
370 report these data in Fig. S3.

## 371 **2.6. Data analysis software**



372 All statistical analyzes and calculations were performed using the software R (R core team  
373 2023). All graphs and figures were generated using the software R and Gimp (The GIMP  
374 development team 2019).

### 375 **3. Results**

#### 376 **3.1. Seasonality of starch mass, growth, and respiration**

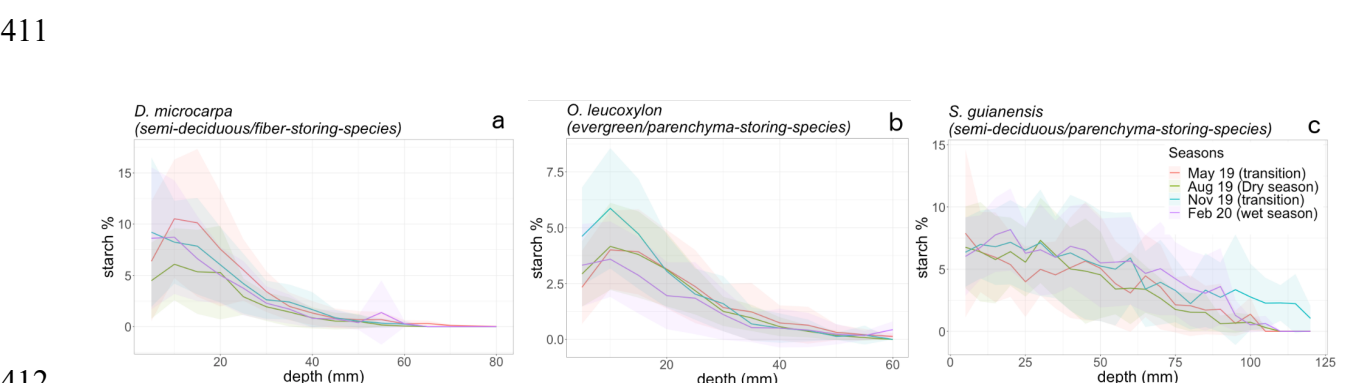
377 Quantifying starch concentrations every 5 mm of wood along the radial axis from bark to pith  
378 allowed us to estimate with high precision the radial profile of starch concentration across  
379 the sampled wood cores (Fig. 3). Starch concentrations decreased radially across the  
380 sapwood from bark to pith for all species at all sampling dates (Fig. 3). Notably, the largest  
381 differences in starch concentration between sampling dates were greater in the first 20 mm  
382 of wood for *D. microcarpa* and *O. leucoxylon*, and in the first 60 mm of wood for *S. guianensis*  
383 (Fig.3, Fig. S6,  $p_{\text{adj}} < 0.05$ ), as compared to the inner regions where starch concentrations  
384 were very low overall.

385 We used these radial profiles of concentrations to estimate the starch mass in the entire  
386 wood core for each tree at each time point (Fig. 4). We found significant differences in the  
387 total starch mass of the entire wood core between the sampling dates for the semi-  
388 deciduous/fiber-storing species *D. microcarpa* ( $p=0.01$ , Fig. 4a) and marginally significant  
389 differences for the semi-deciduous/parenchyma-storing species *S. guianensis* ( $p = 0.059$ , Fig.  
390 4c), while the evergreen/parenchyma-storing species *O. leucoxylon* showed no significant  
391 differences in starch mass in the stemwood during 2019 ( $p > 0.1$ , Fig. 4b). Changes in total  
392 starch mass in the entire wood core between sampling dates were greater for the fiber-  
393 storing species *D. microcarpa* than for the parenchyma-storing species *S. guianensis* and *O.*  
394 *leucoxylon* (Fig. 5). *D. microcarpa* trees significantly reduced starch mass during the dry

395 period May19-Aug19 by about -0.75% (*ci*: -1.15, -0.37, Fig. 5a, Fig S4). with respect to starch  
 396 mass in August 2019, while *S. guianensis* had significant gains in starch mass during the  
 397 period May19-Feb20 by about 0.30% (*ci*: 0.06, 0.53, Fig. 5c, Fig. S4) with respect to the starch  
 398 mass in February 2020. *O. leucoxylon* reduced the starch mass during the wet period Nov19-  
 399 Feb20 by about -0.30% (*ci*: -0.05, -0.63, Fig.5b, Fig. S4) with respect to the starch mass in  
 400 February 2020.

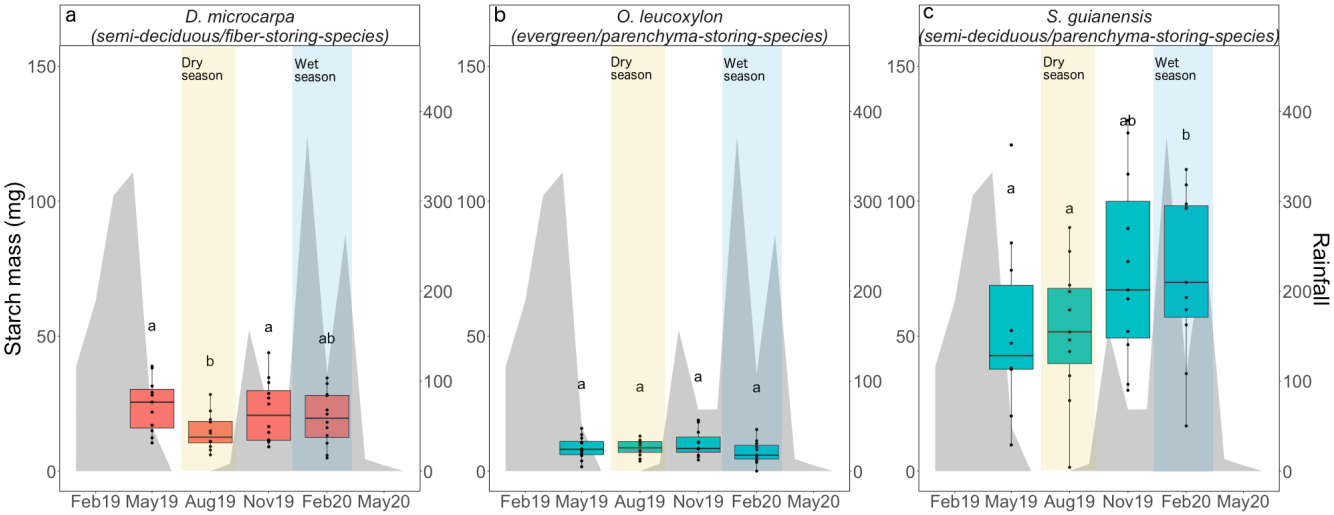
401 Seasonality in monthly growth rates during 2019 was only significant for the semi-  
 402 deciduous/fiber-storing species *D. microcarpa* ( $p < 0.05$ , Fig. 6a). These trees grew slower  
 403 during the dry season and faster during the wet season. The two parenchyma-storing species  
 404 did not show a clear seasonal pattern ( $p > 0.1$ , for all comparisons between all months),  
 405 although a slight decrease in growth rates during the dry season can be noticed for both *O.*  
 406 *leucoxylon* and *S. guianensis* in Figs. 6b and 6c, respectively.

407 Wood respiration was higher in the semi-deciduous/fiber-storing species *D. microcarpa* than  
 408 in the two parenchyma-storing species *S. guianensis* and *O. leucoxylon* ( $p < 0.01$ , Fig. 7).  
 409 However, within-species differences between the 2019 wet and dry seasons were not  
 410 statistically significant for any of the species ( $p > 0.1$ , Fig. 7).



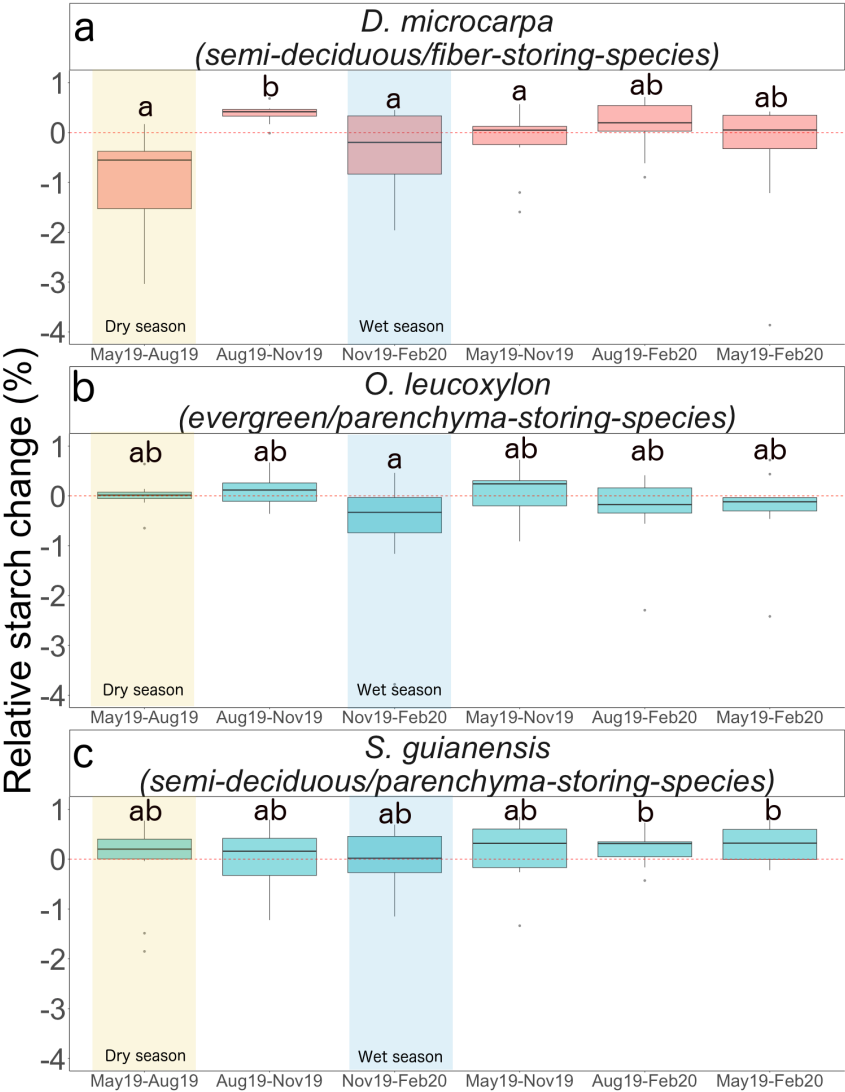
413 **Figure 3:** Radial distribution of the mean starch concentrations in the stemwood measured  
 414 from bark to pith across four different months during 2019. This figure shows the average

starch concentrations, measured every 5-mm interval from bark to pith, across wood cores  
 sampled from adult trees for a) *D. microcarpa*, b) *O. leucoxylon*, and c) *S. guianensis*. We  
 sampled 12 mature trees from each species. Wood cores were sampled during four distinct  
 months in the year 2019, with each month corresponding to a particular seasonal state of  
 precipitation, as indicated in the legend. Starch concentrations were measured along the  
 entire length of the wood cores until reaching a point where starch levels were  
 indistinguishable from zero. The shadowed areas in different colors represent the standard  
 variation around the mean starch concentrations for each of the measured months.



**Figure 4:** Distribution of the wood core starch mass at each time point of measurements  
 for a) *D. microcarpa*, b) *O. leucoxylon*, and c) *S. guianensis*. Starch mass data was obtained  
 from repetitive measurements in wood cores from 12 trees per species across four months  
 of the year 2019. The box-plots show the median and quartile ranges for each group, with  
 individual measurements represented as black dots. These values correspond to the y-axis.  
 Statistical comparisons between different temporal measurements for each species were  
 done using a Wilcoxon signed-rank test. Significant differences between groups at a 90%  
 confidence level are indicated by distinct letters placed above the respective box-plots.

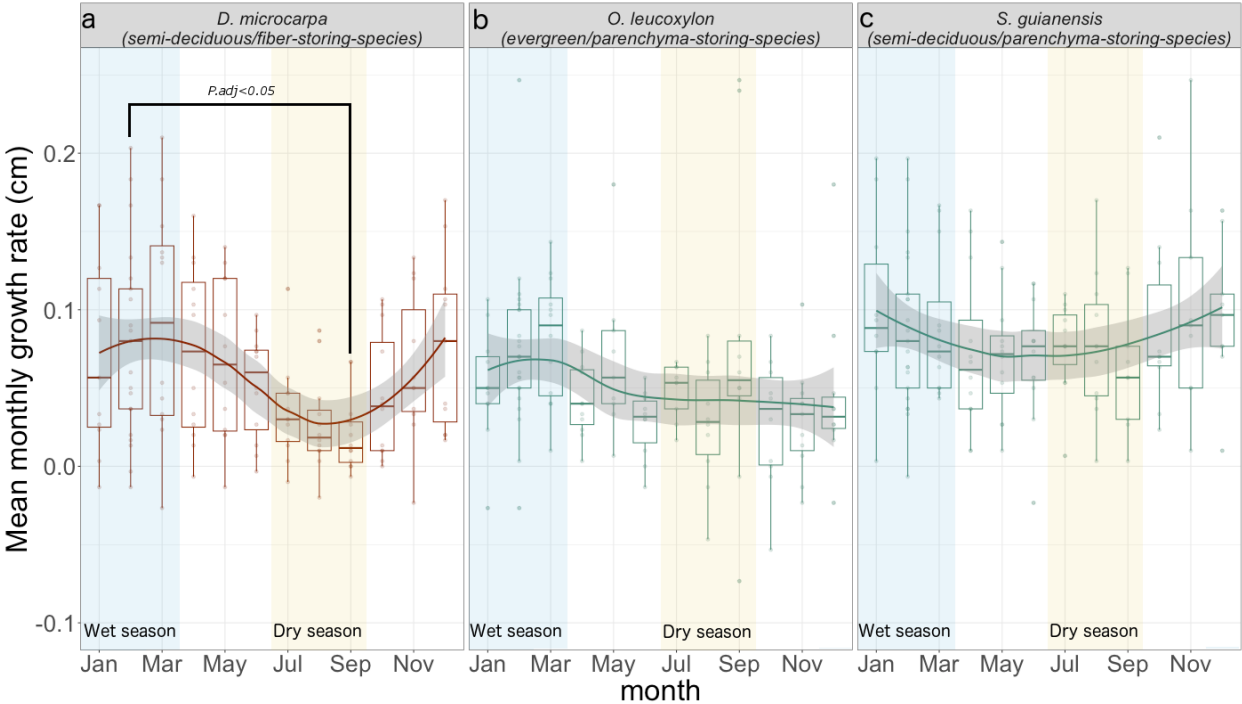
Each panel is labeled with the species name and its specific trait combination. blue box-plots represent parenchyma-storing species, while red box-plots represent the fiber-storing species. The gray-shaded areas behind the box-plots correspond to the monthly precipitation data for the year 2019, with values provided on the right-hand y-axis. We highlighted the dry season with a yellow shading and label, while the wet season is marked with a blue shading and label.



**Figure 5:** Relative starch changes with respect to the starch content in the final month of the evaluated period for a) *D. microcarpa*, b) *O. leucoxylon*, and c) *S. guianensis*. These relative changes were computed using Equation 1, providing a relative measure of the increase of

443 decrease in starch mass within a given time frame. We estimated these changes for all  
 444 possible permutations of months, keeping their temporal order. The x-axis describes the  
 445 evaluated periods, first denoting the initial month and then the final month. Box-plots  
 446 indicate the data distribution derived from measurements of 12 trees per species per time  
 447 point. Statistical comparisons between groups with a 95% confidence level were evaluated  
 448 using nonparametric confidence intervals obtained from bootstrapping methods (see Fig. S4).  
 449 Here, significant statistical differences occurred when the confidence intervals did not  
 450 overlap, denoted by distinct letters. Red box-plot represent fiber-storing species and blue  
 451 box-plots represent parenchyma-storing species. The wet season is highlighted with a shaded  
 452 blue area, while the dry season is highlighted with a shaded yellow area.

453

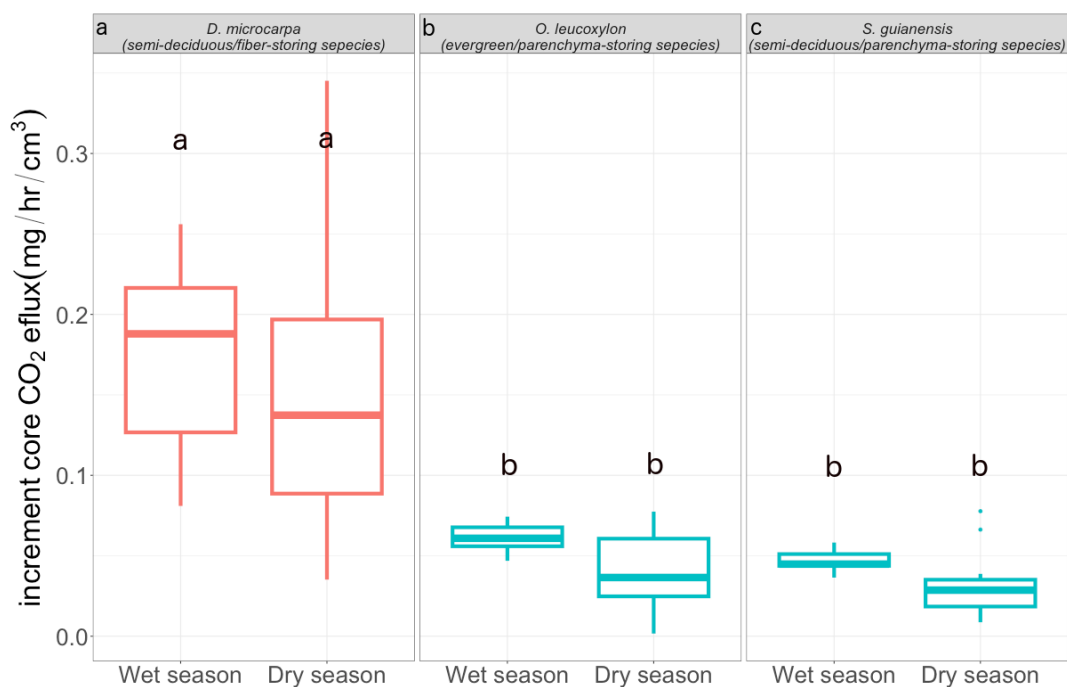


454

455 **Figure 6:** Seasonal course of growth rates for a) *D. microcarpa*, b) *O. leucoxylon*, and c) *S.*  
 456 *guianensis*. The figure shows the mean monthly growth rates, derived from measurements  
 457 of 12 trees per species throughout the year 2019. The distribution of data can be observed

458 in the box-plots, while the seasonal trend is shown by a smooth colored solid line across  
 459 the data groups. The gray shaded areas correspond to the 95% confidence interval of the  
 460 smooth line. The wet season is indicated by blue shading, and the dry season is indicated  
 461 by orange shading. Comparisons between monthly growth rates within each species were  
 462 statistically assessed using a Wilcoxon signed-rank test, with P values adjusted following  
 463 the Holm method (Holm 1979). We only found significant differences between growth  
 464 rates in February 2019 and September 2019 for *D. microcarpa* (p.adj=0.032), and this is  
 465 indicated in the figure by a solid black line that connects the compared months and depicts  
 466 the p-value. Each panel is labeled with the name of the species and its specific combination  
 467 of traits. Red box-plots correspond to the fiber-storing species, while blue box-plots  
 468 correspond to the parenchyma-storing species.

469



470

471 **Figure 7:** Seasonal variation of stemwood respiration for the evaluated species. The  
 472 figures present wood respiration rates per unit of wood volume, measured in 12 trees per  
 473 species during both the wet and dry seasons of 2019. The distribution of the data is shown

by box-plots. Statistical comparisons of wood respiration rates between seasons and species were evaluated using the Wilcoxon ranked test. Significant differences at the 95% confidence level are denoted by distinct letters. Each panel is labeled with the species name and its specific trait combination. The red box-plots represent the fiber-storing species, whereas the blue box-plots represent the parenchyma-storing species.

### 3.3. Accumulation and consumption of starch

We observed a strong significant starch consumption during the dry season (May19-Aug19) only for the fiber-storing species *D. microcarpa* ( Fig. 5a, Fig. S4). The parenchyma-storing species (*S. guianensis* and *O. leucoxylon*) did not show significant starch consumption during the dry season (Fig. 5b, Fig. 5c, Fig. S4). Nevertheless, our data suggest that *S. guianensis* may consume starch during the transition months between the wet and the dry season from February to May, since the total starch mass appears to be lower in May than in February (Fig. 4c), but unfortunately, we did not measure this period in 2020.

We observed a marginally significant consumption of starch during the wet season (Nov19-Feb2020) for the fiber-storing species *D. microcarpa* (Fig. 5a and Fig. S4) and a significant consumption for the evergreen/parenchyma-storing species *O. leucoxylon* (Fig. 5b and Fig. S4).

Starch accumulation was observed during the transition from dry to wet season (Aug19-Nov20) in the semi-deciduous/fiber-storing species trees (*D. microcarpa*) (Fig. 5a). The semi-deciduous/parenchyma-storing species (*S. guianensis*) accumulated starch from May 2019 to February 2020 (Fig. 5c and Fig. S4). We did not detect a significant accumulation of starch for the evergreen/parenchyma-storing species *O. leucoxylon* ( $p < 0.05$ , Fig. 5b).

### 3.4 Seasonal storage-growth trade-offs in stemwood

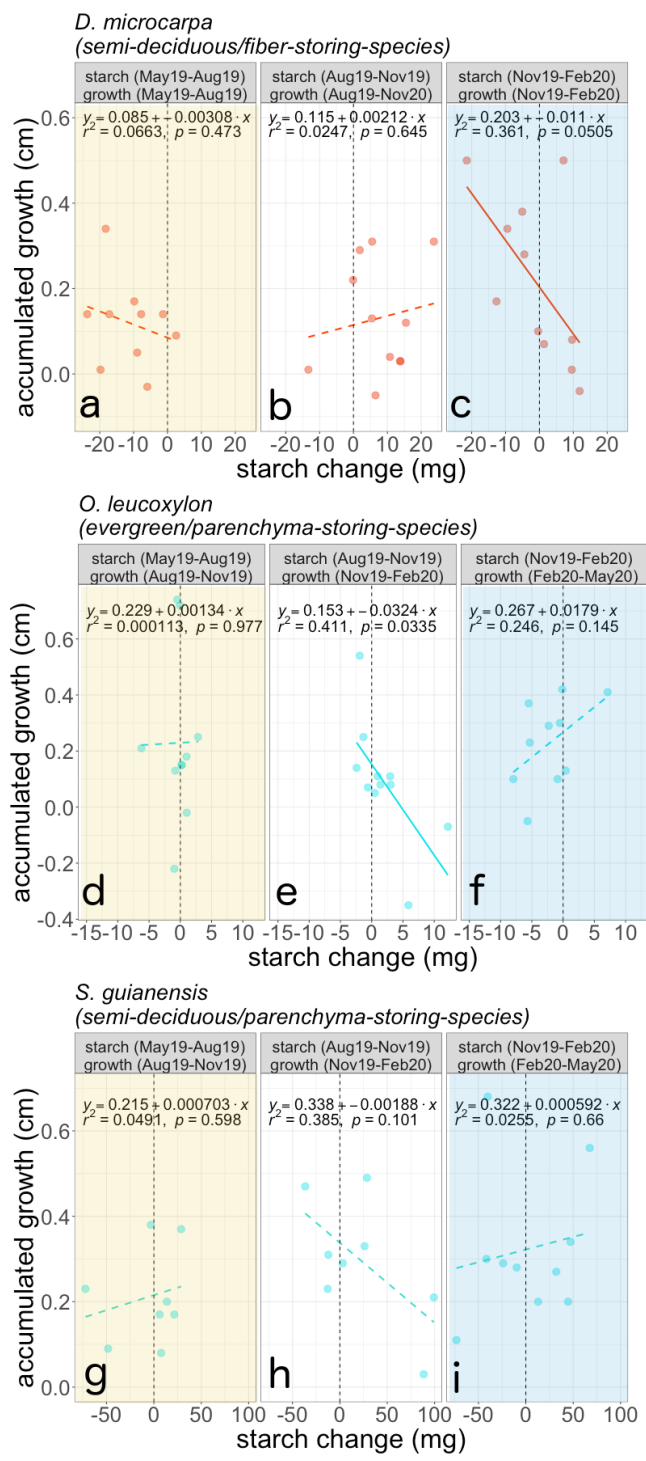
During the wet season (Nov19-Feb20), we observed a negative correlation between the cumulative three months of growth (Nov19-Feb20) and the changes in starch mass during the same period, for the semi-deciduous/fiber-storing species *D. microcarpa* ( $p = 0.05$ ,  $r^2 = 0.36$ , Fig. 8c). During this period, some trees consumed starch while growing rapidly, and some other individuals accumulated starch instead and had lower growth rates (Fig. 8c). For this species, most individuals had high starch consumption during the dry season, but this was not associated with growth (Fig. 8a).

For the parenchyma-storing species *O. leucoxylon* and *S. guianensis*, the storage-growth trade-off was less clear or absent. We did not observe a significant correlation when comparing concurrent growth and starch change during each season ( $p > 0.05$ ). However, we found significant correlations when considering a three-month lag between growth and starch change for these species (Fig. 8, see only *O. leucoxylon* and *S. guianensis*). For example, we observed a negative correlation between the starch mass change during the transition months from dry to wet season (from August 2019 to November 2019, Aug19-Nov19) and the three months of cumulative growth in the wet season (from November 2019 to February 2020, Nov19-Feb20), for *O. leucoxylon* ( $p = 0.03$ ,  $r^2 = 0.41$ , Fig. 8e). This trend was also observed for *S. guianensis*, but it was not significant ( $p = 0.1$ ,  $r^2 = 0.38$ , Fig. 8h).

We observed that starch mass in the whole wood core was related to annual stem growth, but there were differences between storage strategies. The storage-growth trade-off trend between the starch mass in February 2020 and the annual growth calculated from February 2019 to February 2020 was still distinguishable for the fiber-storing species *D. microcarpa*, although not significant ( $p = 0.11$ ,  $r^2 = 0.31$ , Fig. S5d). The semi-deciduous/parenchyma-storing species (*S. guianensis*) showed a positive correlation between annual growth and starch mass



520 in the dry season (August 2019,  $p=0.003$ ,  $r^2=0.79$ , Fig. S5j), and also a positive correlation  
521 between annual growth and starch mass in the wet season, although weaker (February  
522 2020,  $p=0.05$ ,  $r^2=0.51$ , Fig. S5l). The evergreen/parenchyma-storing species (*O. leucoxylon*)  
523 showed a marginal positive relationship between starch mass in May 2019 (transition  
524 between wet and dry seasons) and annual stem growth ( $p=0.088$ ,  $r^2=0.4$ , Fig. S5e).



**Figure 8:** Relationship between changes in starch mass and the cumulative three-month growth in 2019 for the species studied. For the fiber-storing species (red

541 lines and dots), growth and starch changes are evaluated over the same months. Conversely,  
542 For the two parenchyma-storing species (blue lines and dots), no significant correlations  
543 were observed between growth and starch changes during the same months. Instead, starch  
544 storage shifts are correlated with growth in the following 3 months. The specific periods of  
545 comparison for starch change and growth are indicated in each panel's title. Solid lines  
546 denote significant correlations, while dashed lines represent non-significant correlations. The  
547 vertical dotted line marks the 0 in the starch change axis. Shaded yellow areas indicate the  
548 dry season, and shaded blue areas indicate the wet season. Species names and their specific  
549 combination of traits appear over the three respective sub-panels, and each sub-panel is  
550 labeled with distinct consecutive letters from a to i.

551

#### 552 **4. Discussion**

553 Our results relate differences in carbon dynamics in mature trees to life history traits such as  
554 the anatomical distribution of starch in the stemwood and leaf habit, which we summarized  
555 in our conceptual framework (Fig. 1). Evergreen trees should not exhibit large seasonal  
556 differences in carbon acquisition, while the semi-deciduous trees studied here shed between  
557 50 and 70% of their leaves during the dry season (May-August) and then invest carbon in  
558 new leaves (Fig. S3). Eddy covariance data from this forest indicate an overall decrease in  
559 canopy photosynthesis during the dry season (Brando et al., 2019). During the wet season  
560 (November-February), a slight reduction in the percentage of mature leaves may also indicate  
561 a slight reduction in photosynthesis (Fig. S3, Fig. 1).

562 The slow-growing semi-deciduous/fiber-storing species (*Dacryodes microcarpa*) showed  
563 seasonality in radial growth but not in wood respiration. This species had the largest starch  
564 consumption during the dry season, when carbon availability was low and radial growth was

almost absent, and low overall consumption during the wet season, when carbon availability and growth were higher (Fig. 5, Fig. 6). As expected, we observed a strong negative correlation between growth and starch storage change in this species during the wet season, with starch consumption when trees grew rapidly, and starch accumulation when radial growth was low (Fig 1, Fig. 8c).

The two parenchyma-storing species partially agreed with our hypothesis presented in our conceptual figure. Neither showed marked seasonality in radial growth or wood respiration. Starch mass was marginally seasonal in the semi-deciduous/parenchyma-storing species (*Sacoglottis guianensis*), with starch accumulation towards the wet season from November to February. Although we did not detect significant starch consumption during the wet or dry season, lower starch mass values in May and August 2019 suggest that these trees consume starch during the dry season. For the evergreen/parenchyma-storing species (*Ocotea leucoxylon*), we did not observe seasonality in the total starch mass, but we did detect significant starch consumption during the wet season (Fig. 5, Fig. S4). For these parenchyma-storing species, we did not observe clear seasonal trade-offs (Fig. 1, Fig. 8).

Although we only evaluated three of the four conceptualized groups of our framework, omitting the evergreen/fiber-storing species (Fig. 1), we expect that differences in carbon fluxes between evergreen and semi-deciduous fiber-storing species to be small because previous analyses indicated little difference in starch storage, mortality, respiration, or growth between these groups (Hererra-Ramirez et al., 2021). We included the expected behavior of this group in our conceptual framework as reference, which was based on previous observations of species belonging to that group (e.g., *Trattinnickia glaziovii*). Our conceptual framework, in conjunction with our results, helps to improve our understanding of the carbon dynamics in mature trees and provides insights into the mechanisms behind

589 starch accumulation and use and its interplay with carbon sources and sinks that may  
590 influence plant survival in tropical forests.

#### 591 **4.1. Semi-deciduous and fiber-storing species have seasonal changes in starch**

592 As we expected, leaf habit influences the seasonal fluctuation of starch in the stemwood.  
593 Semi-deciduous species showed significant starch seasonality ( $p=0.001$ ), while the evergreen  
594 species we analyzed did not (Fig. 4). Semi-deciduous trees may experience large imbalances  
595 between carbon sources and sinks during the dry season due to the high demand for carbon  
596 to rebuild the canopy, and therefore show larger seasonal changes in starch mass than  
597 evergreen species (Kozlowski, 1992; Hoch et al., 2003; Würth et al., 2005; Richardson et al.,  
598 2013; Furze et al., 2018). This is consistent with studies showing strong seasonality of NSC in  
599 temperate deciduous species but not in evergreen species (Chapin et al., 1990; Piispanen and  
600 Saranpää, 2001; Trumbore et al., 2015; Martínez-Vilalta et al., 2016; Furze et al., 2018). Our  
601 results support the idea that semi-deciduous trees are more dependent on NSC storage than  
602 evergreen species in seasonally dry tropical forests. Nevertheless, a larger number of species  
603 and a longer time series of NSC dynamics should be examined to confirm these patterns.  
604 Histological methods provide a rapid and convenient method for quantifying starch storage  
605 that can support observations in a large number of species over the longer term.

606 Our results also suggest that the anatomical distribution of starch in the stemwood influences  
607 the seasonal dynamics of NSC (Fig. 5, Fig. S4). The fiber-storing species (*D. microcarpa*) had  
608 larger seasonal changes in starch mass than the parenchyma-storing species (Fig. S4). This  
609 provides some insight into the mechanisms behind carbon dynamics. In fiber-storing species,  
610 the high seasonal carbon consumption may be caused by the larger proportion of living cells  
611 in the stem, which require more carbon for respiration than the smaller proportion of living  
612 cells in parenchyma-storing species (Fig. 7). This high metabolic cost may compete for carbon

resources with other carbon sinks, such as radial growth during the dry season, probably leading to seasonal growth and a high reliance on storage to keep cells alive during low photosynthetic periods. However, the presence of living fibers allows trees to use a larger volume of wood to store starch and cover all metabolic needs, which could reduce their vulnerability to carbon starvation. This would likely benefit plant competition/survival and result in low mortality rates (e.g., Table 1 and Fig S7).

**4.2. Fiber-storing species have greater plasticity in radial growth, but not in respiration**

Trees growing in a seasonally dry environment should balance their carbon sources and sinks in order to survive recurrent stress conditions. The fast-growing, parenchyma-storing species *S. guianensis* kept radial growth relatively constant throughout the year, while the slow-growing, fiber-storing species *D. microcarpa* significantly reduced growth during the dry season (Fig. 6). Thus, these trees, with contrasting starch storage traits and respiration requirements, differentially adjust their balance between carbon sources and sinks during seasonally dry conditions (Fig. 1). It is important to note that radial growth is not the only form of growth in trees, and other forms of growth may be differentially adjusted or prioritized. Nevertheless, reducing radial growth rates during periods of low productivity may make trees more tolerant to a wider range of environmental conditions (Rowland et al., 2023).

Contrary to the growth patterns described above, wood respiration did not show seasonal changes for any of the species studied during the year of measurement (Fig. 7). A previous study also reported no changes in CO<sub>2</sub> efflux between dry and wet seasons in tropical forest trees (Asao et al., 2015). CO<sub>2</sub> efflux includes not only wood respiration, but also phloem respiration, CO<sub>2</sub> re-fixation, and transport of CO<sub>2</sub> dissolved in sap (Helm et al., 2023).

637 Nevertheless, it has been shown that some trees can regulate stem CO<sub>2</sub> efflux and growth to  
638 adapt to specific environmental conditions, such as seasonal changes in precipitation and  
639 temperature, thereby releasing resources for other metabolic activities (Teskey et al., 2008;  
640 Huang et al., 2019; Sierra et al., 2022). Thus, it is also possible that the 2019 dry period was  
641 not strong enough to induce changes in CO<sub>2</sub> efflux, and longer time series of wood respiration  
642 are needed to confirm this pattern.

643 Overall, wood respiration was greatest in the fiber-storing species (*D. microcarpa*), probably  
644 because the proportion of living cells in the wood of this species, which includes parenchyma  
645 and living fibers, is greater than in parenchyma-storing species (*S. guianensis* and *O.*  
646 *leucoxylon*). Therefore, fiber-storing species may have greater carbon requirements to keep  
647 all living wood tissue alive, potentially reducing their ability to regulate respiration  
648 seasonally. These results suggest that wood anatomical features of wood, such as living fibers,  
649 may indicate a large carbon demand for respiration that may compete strongly with other  
650 carbon sinks, such as growth, leading these trees to have a stronger regulation of growth  
651 rates and a high reliance on storage. Thus, the proportion of living wood biomass may have  
652 greater control over wood respiration than environmental conditions.

#### 653 **4.3. Starch is consumed during the peak of the dry and the wet season**

654 Our results show that, for most trees, starch seems to be consumed not only during the peak  
655 of the dry season, but also during the peak of the wet season (Fig. 5, Fig S4). This is clear for  
656 *D. microcarpa* and *O. leucoxylon* trees, but we did not observe starch consumption during the  
657 wet season for *S. guianensis* (Fig. 5, Fig. S4). We hypothesize that trees may recharge the  
658 carbon storage pools during the wet season due to favorable growth and less photosynthetic  
659 limitation due to more water availability and lower air temperatures (Dietze et al., 2014).  
660 Nevertheless, starch consumption during the wet season suggests a negative carbon balance

661 in these trees. There may be several reasons for this: For fiber-storing trees (*D. microcarpa*),  
662 it is possible that the carbon demand of wood growth and respiration exceeds the supply of  
663 new photoassimilates. In addition, some other carbon sinks, such as reproduction, may also  
664 contribute to a greater carbohydrate demand during the wet season (Hartmann and  
665 Trumbore, 2016, Fig. S3). For instance, both *D. microcarpa* and *S. guianensis* produced flowers  
666 and fruits during the wet season (Fig. 1, Fig. S3), which may represent a large demand on  
667 carbon reserves (Hartmann and Trumbore, 2016; Blumstein et al., 2022). Alternatively, the  
668 wet season presents its own challenges that put pressure on available NSC. During the rainy  
669 season, blowdowns, storms, and lightning can cause physical damage to trees, which, along  
670 with increased herbivory or fungal infections, may increase the risk of individual damage or  
671 even mortality (Zuleta et al., 2022, Aleixo et al., 2019). Rebuilding lost tissue during the wet  
672 season or building defensive compounds could represent a large demand on NSC (Huang et  
673 al., 2021). Thus, trees that have not replenished their NSC reserves and repaired damage  
674 during the transition months between the wet and dry seasons may be more vulnerable and  
675 susceptible to die when facing stress (Anderegg et al., 2013; Arellano et al., 2019; Aleixo et al.,  
676 2019).

#### 677 **4.4. Storage-growth trade-offs occur in fiber-storing species during the wet season**

678 Seasonal patterns in starch mass showed us when trees consumed and accumulated stem  
679 NSC. Our results indicate a trade-off between starch consumption/accumulation and growth  
680 only during the wet season for the fiber-storing species *D. microcarpa*. For the parenchyma-  
681 storing species (*O. leucoxylon* and *S. guianensis*) a storage-growth trade-off was only observed  
682 when lagged correlations were considered (Fig. 8), probably reflecting the use of reserves for  
683 other metabolic needs, such as reproduction, that are also important for species survival (Fig.  
684 S3, Blumstein et al., 2022, Hartmann and Trumbore 2016). Nevertheless, it is important to

clarify that these storage-growth trade-offs may change when other forms of growth are considered. At the whole tree level, the inclusion of other starch storage tissues, such as phloem (Rosell et al., 2021) or roots (Hillman et al., 2021) may provide further insight into these trade-offs.

Identifying trade-offs between C sinks with field measurements is challenging (Blumstein et al., 2022). Here, we show that annual estimates of starch mass and radial stem growth may mask the seasonal trade-offs between these two variables (Fig. S5). This may be influenced by the fact that a single time point measurement of starch mass integrates past fluctuations in plant carbon balance and therefore may not reflect the actual current growing conditions of trees. Also, trade-offs may be influenced by variations in carbon supply. Individuals with higher carbon acquisition may have more leverage to invest more in both growth and storage than individuals with lower carbon acquisition (Blumstein et al., 2022), as may have been the case for our parenchyma-storing species. These two parenchyma-storing species grew faster than the fiber-storing species, and the correlation between annual growth and starch mass was positive for both species during the dry season (August 2019) for *S. guianensis* (Fig. S5e) and during the transition months from wet to dry season (May 2019) for *O. leucoxydon* (Fig. S5j). For more conservative species, such as the fiber-storing species, a weak trade-off trend between starch mass and annual growth was still observed during the wet season (Fig. S5d). Our results suggest that these trade-offs may not be reflected in the annual carbon balance, but only on a seasonal basis. Therefore, evaluating seasonal changes in starch mass (rather than mean storage mass per year or maximum starch storage per year) in relation to seasonal growth may better inform us about when and under what conditions such trade-offs occur, avoiding the influence of some confounding factors when annual averages are used.



Evidence for growth-storage trade-offs is increasing in the literature and has been linked to individual survival (Wright et al., 2004; Poorter and Kitajima, 2007; O'Brien et al., 2014, 2015; Klein and Hoch, 2015; D'Andrea et al., 2019, 2020). Studies comparing highly conservative species and highly competitive species in surviving stressful environmental conditions are needed to further understand how variation in storage-growth trade-offs are maintained across space and time. Our results are a good indication that highly conservative, slow-growing, and high-storage species, such as fiber-storing species (e.g., *D. microcarpa*), may have stronger trade-offs between starch storage and growth, indicating greater plasticity in their sink activity, and/or higher prioritization of storage than parenchyma-storing species. Life history traits such as low growth and low mortality rate are associated with highly conservative species and are a good proxy for tree longevity (Wright et al., 2004; Herrera-Ramírez et al., 2021; Piovesan and Biondi, 2021). Mortality was also lower for the fiber-storing species compared to the parenchyma-storing species evaluated here, not only at background levels in the forest, but also when the forest experienced annual fires (Fig. S7). Therefore, we would expect that as stressors intensify, highly conservative trees, such as fiber-storing species, would have higher survival rates and would become more competitive in seasonally dry forests. Future work should focus on understanding how these species with contrasting anatomical distributions of starch in the stemwood are distributed in forests, and how their different storage-growth trade-offs affect productivity and species survival in tropical forests under climate change.

## 5. Acknowledgments

We want to thank the Max Planck Society, the Friedrich-Schiller University and the German Research Foundation (SI 1953/2-1) for the founding of this project. Also, we received funding from the EU, through the German Bundesanstalt für Landwirtschaft und Ernährung FKZ:

2816ERA03W (to S. T.); the Brazilian council for Scientific and Technological Development - CNPq (PELD/CNPq No. 441703/2016-0; No. 441940/2020-0; UNIVERSAL No. 430149/2018-2); and DH and ST acknowledge the support from the Balzan foundation.

## 6. Author contributions

DH-R, CAS, HH, CR, ST, and JM, conceived the idea, conceptualized the hypothesis, and planned the data collection and the experiments. DH, LM-S and DS collected and analyzed data. DH and IK performed sample analysis and quantification. DH-R wrote the manuscript. DH-R, HH, CR, ST, JM, LM-S, PB, DS, HJ, IK, and CAS contributed significantly to the writing of the manuscript and gave important and critical input. All authors revised the manuscript.

## 7. Data accessibility statement

The data and scripts for calculation and statistical analysis presented here are available in the open GitHub repository ""

## 8. References

- Adams, H. D., Zeppel, M. J. B., Anderegg, W. R. L., Hartmann, H., Landhäusser, S. M., Tissue, D. T., et al. (2017). A multi-species synthesis of physiological mechanisms in drought-induced tree mortality. *Nature Ecology & Evolution* 1, 1285–1291. doi:10.1038/s41559-017-0248-x.
- Aleixo, I., Norris, D., Hemerik, L., Barbosa, A., Prata, E., Costa, F., et al. (2019). Amazonian rainforest tree mortality driven by climate and functional traits. *Nature Climate Change* 9, 384–388. doi:10.1038/s41558-019-0458-0.
- Anderegg, W. R. L., Berry, J. A., Smith, D. D., Sperry, J. S., Anderegg, L. D. L., and Field, C. B. (2012). The roles of hydraulic and carbon stress in a widespread climate-induced forest die-off. *Proceedings of the National Academy of Sciences* 109, 233–237. doi:10.1073/pnas.1107891109.
- Anderegg, W. R. L., Kane, J. M., and Anderegg, L. D. L. (2013). Consequences of widespread tree mortality triggered by drought and temperature stress. *Nature Climate Change* 3, 30–36. doi:10.1038/nclimate1635.
- Arellano, G., Medina, N. G., Tan, S., Mohamad, M., and Davies, S. J. (2019). Crown damage and the mortality of tropical trees. *New Phytologist* 221, 169–179. doi:https://doi.org/10.1111/nph.15381.

761 Arx, G. von, Arzac, A., Fonti, P., Frank, D., Zweifel, R., Rigling, A., et al. (2017). Responses of  
 762 sapwood ray parenchyma and non-structural carbohydrates of *pinus sylvestris* to drought  
 763 and long-term irrigation. *Functional Ecology* 31, 1371–1382. doi:10.1111/1365-2435.12860.

764 Asao, S., Bedoya-Arrieta, R., and Ryan, M. G. (2015). Variation in foliar respiration and wood  
 765 CO<sub>2</sub> efflux rates among species and canopy layers in a wet tropical forest. *Tree Physiology* 35,  
 766 148–159. doi:10.1093/treephys/tpu107.

767 Barbaroux, C., and Bréda, N. (2002). Contrasting distribution and seasonal dynamics of  
 768 carbohydrate reserves in stem wood of adult ring-porous sessile oak and diffuse-porous  
 769 beech trees. *Tree Physiology* 22, 1201–1210. doi:10.1093/treephys/22.17.1201.

770 Blumstein, M., Sala, A., Weston, D. J., Holbrook, N. M., and Hopkins, R. (2022). Plant  
 771 carbohydrate storage: Intra- and inter-specific trade-offs reveal a major life history trait. *New*  
 772 *Phytologist* 235, 2211–2222. doi:https://doi.org/10.1111/nph.18213.

773 Canty, A., Ripley, B. D. (2022). *boot: Bootstrap R (S-Plus) Functions*. R package version  
 774 1.3-28.1.

775 Carlquist, S. (2013). Fibre dimorphism: Cell type diversification as an evolutionary strategy in  
 776 angiosperm woods. *Botanical Journal of the Linnean Society* 174, 44–67.  
 777 doi:10.1111/boj.12107.

778 Chapin, F. S., Schulze, E., and Mooney, H. A. (1990). The ecology and economics of storage in  
 779 plants. *Annual Review of Ecology and Systematics* 21, 423–447.  
 780 doi:10.1146/annurev.es.21.110190.002231.

781 Davison, A.C., and Hinkley, D. V. (1997). Bootstrap methods and their application, Chapter5.  
 782 Cambridge University Press.

783 Dietze, M. C., Sala, A., Carbone, M. S., Czimczik, C. I., Mantooth, J. A., Richardson, A. D., et al.  
 784 (2014). Nonstructural carbon in woody plants. *Annual Review of Plant Biology* 65, 667–687.  
 785 doi:10.1146/annurev-arplant-050213-040054.

786 D’Andrea, E., Rezaie, N., Battistelli, A., Gavrichkova, O., Kuhlmann, I., Matteucci, G., et al.  
 787 (2019). Winter’s bite: Beech trees survive complete defoliation due to spring late-frost  
 788 damage by mobilizing old c reserves. *New Phytologist* 224, 625–631.  
 789 doi:https://doi.org/10.1111/nph.16047.

790 D’Andrea, E., Rezaie, N., Prislan, P., Gričar, J., Collalti, A., Muhr, J., et al. (2020). Frost and  
 791 drought: Effects of extreme weather events on stem carbon dynamics in a mediterranean  
 792 beech forest. *Plant, Cell & Environment* 43, 2365–2379.  
 793 doi:https://doi.org/10.1111/pce.13858.

794 Fox, J. (2015). Applied regression analysis and generalized linear models (3rd ed.). Sage  
 795 Publications.

796 Fournier, L. A., 1974. Un método cuantitativo para la medición de características fenológicas  
 797 en árboles. *Turrialba* 24(4):422-423.

798 Furze, M. E., Huggett, B. A., Chamberlain, C. J., Wieringa, M. M., Aubrecht, D. M., Carbone, M. S.,  
 799 et al. (2020). Seasonal fluctuation of nonstructural carbohydrates reveals the metabolic

800 availability of stemwood reserves in temperate trees with contrasting wood anatomy. *Tree*  
801 *Physiology* 40, 1355–1365. doi:10.1093/treephys/tpaa080.

802 Furze, M. E., Trumbore, S., and Hartmann, H. (2018). Detours on the phloem sugar highway:  
803 Stem carbon storage and remobilization. *Current Opinion in Plant Biology* 43, 89–95.  
804 doi:<https://doi.org/10.1016/j.pbi.2018.02.005>.

805 Galiano, L., Timofeeva, G., Saurer, M., Siegwolf, R., Martínez-Vilalta, J., Hommel, R., et al.  
806 (2017). The fate of recently fixed carbon after drought release: Towards unravelling c storage  
807 regulation in *tilia platyphyllos* and *pinus sylvestris*. *Plant, Cell & Environment* 40, 1711–1724.  
808 doi:<https://doi.org/10.1111/pce.12972>.

809 Hartmann, H., and Trumbore, S. (2016). Understanding the roles of nonstructural  
810 carbohydrates in forest trees from what we can measure to what we want to know. *New*  
811 *Phytologist* 211, 386–403. doi:10.1111/nph.13955.

812 Hartmann, H., McDowell, N. G., and Trumbore, S. (2015). Allocation to carbon storage pools in  
813 Norway spruce saplings under drought and low CO<sub>2</sub>. *Tree Physiology* 35, 243–252.  
814 doi:10.1093/treephys/tpv019.

815 Hartmann, H., Ziegler, W., and Trumbore, S. (2013). Lethal drought leads to reduction in  
816 nonstructural carbohydrates in norway spruce tree roots but not in the canopy. *Functional*  
817 *Ecology* 27, 413–427. doi:<https://doi.org/10.1111/1365-2435.12046>.

818 Helm, J., Salomón, R.L., Hilman, B., Muhr, J., Knohl, A., Steppe, K. et al. (2023). Differences  
819 between tree stem CO<sub>2</sub> efflux and O<sub>2</sub> influx rates cannot be explained by internal CO<sub>2</sub>  
820 transport or storage in large beech trees. *Plant, Cell & Environment*, 1– 14.  
821 <https://doi.org/10.1111/pce.14614>

822 Herrera-Ramírez, D., Sierra, C. A., Römermann, C., Muhr, J., Trumbore, S., Silvério, D., et al.  
823 (2021). Starch and lipid storage strategies in tropical trees relate to growth and mortality.  
824 *New Phytologist* 230, 139–154. doi:<https://doi.org/10.1111/nph.17239>.

825 Hilman, B., Muhr, J., Helm, J., Kuhlmann, I., Schulze, E.-D., & Trumbore, S. (2021). The size and  
826 the age of the metabolically active carbon in tree roots. *Plant, Cell & Environment* 44( 8),  
827 2522– 2535. <https://doi.org/10.1111/pce.14124>

828 Hoch, G., Richter, A., and Körner, C. (2003). Non-structural carbon compounds in temperate  
829 forest trees. *Plant, Cell & Environment* 26, 1067–1081. doi:10.1046/j.0016-  
830 8025.2003.01032.x.

831 Holm, S. (1979). A Simple Sequentially Rejective Multiple Test Procedure. *Scandinavian*  
832 *Journal of Statistics*, 6(2), 65–70. <http://www.jstor.org/stable/4615733>

833 Hollander, M., Wolfe, D. A., and Chicken, E. (2013). Nonparametric statistical methods (3rd  
834 ed.). John Wiley & Sons.

835 Huang, J., Hammerbacher, A., Gershenzon, J., Dam, N. M. van, Sala, A., McDowell, N. G., et al.  
836 (2021). Storage of carbon reserves in spruce trees is prioritized over growth in the face of  
837 carbon limitation. *Proceedings of the National Academy of Sciences* 118, e2023297118.  
838 doi:10.1073/pnas.2023297118.

839 Huang, J., Rücker, A., Schmidt, A., Gleixner, G., Gershenzon, J., Trumbore, S., et al. (2020).  
840 Production of constitutive and induced secondary metabolites is coordinated with growth  
841 and storage in Norway spruce saplings. *Tree Physiology* 40, 928-942.  
842 doi:<https://doi.org/10.1093/treephys/tpaa040>.

843 Huang, J., Hammerbacher, A., Weinhold, A., Reichelt, M., Gleixner, G., Behrendt, T., et al.  
844 (2019). Eyes on the future – evidence for trade-offs between growth, storage and defense in  
845 norway spruce. *New Phytologist* 222, 144–158. doi:<https://doi.org/10.1111/nph.15522>.

846 Klein, T., and Hoch, G. (2015). Tree carbon allocation dynamics determined using a carbon  
847 mass balance approach. *New Phytologist* 205, 147–159. doi:10.1111/nph.12993.

848 Kozlowski, T. T. (1992). Carbohydrate sources and sinks in woody plants. *The Botanical*  
849 *Review* 58, 107–222. doi:10.1007/BF02858600.

850 Körner, C. (2003). Carbon limitation in trees. *Journal of Ecology* 91, 4–17.  
851 doi:<https://doi.org/10.1046/j.1365-2745.2003.00742.x>.

852 Landhäusser, S. M., Chow, P. S., Dickman, L. T., Furze, M. E., Kuhlman, I., Schmid, S., et al.  
853 (2018). Standardized protocols and procedures can precisely and accurately quantify non-  
854 structural carbohydrates. *Tree Physiology* 38, 1764–1778. doi:10.1093/treephys/tpy118.

855 Martínez-Vilalta, J. (2014). Carbon storage in trees: pathogens have their say. *Tree Physiology*  
856 34, 215–217. doi:10.1093/treephys/tpu010.

857 Martínez-Vilalta, J., Sala, A., Asensio, D., Galiano, L., Hoch, G., Palacio, S., et al. (2016). Dynamics  
858 of non-structural carbohydrates in terrestrial plants: A global synthesis. *Ecological*  
859 *Monographs* 86, 495–516. doi:<https://doi.org/10.1002/ecm.1231>.

860 Meir, P., Mencuccini, M., and Coughlin, S. I. (2020). Respiration in wood: Integrating across  
861 tissues, functions and scales. *New Phytologist* 225, 1824–1827.  
862 doi:<https://doi.org/10.1111/nph.16354>.

863 Michelot, A., Simard, S., Rathgeber, C., Dufrêne, E., and Damesin, C. (2012). Comparing the  
864 intra-annual wood formation of three European species (*Fagus sylvatica*, *Quercus petraea*  
865 and *Pinus sylvestris*) as related to leaf phenology and non-structural carbohydrate dynamics.  
866 *Tree Physiology* 32, 1033–1045. doi:10.1093/treephys/tps052.

867 Muhr, J., Trumbore, S., Higuchi, N. and Kunert, N. (2018), Living on borrowed time –  
868 Amazonian trees use decade-old storage carbon to survive for months after complete stem  
869 girdling. *New Phytologist*, 220: 111-120. <https://doi.org/10.1111/nph.15302>

870 Newell, E. A., Mulkey, S. S., and Wright, J. S. (2002). Seasonal patterns of carbohydrate storage  
871 in four tropical tree species. *Oecologia* 131, 333–342. doi:10.1007/s00442-002-0888-6.

872 O'Brien, M. J., Burslem, D. F. R. P., Caduff, A., Tay, J., and Hector, A. (2015). Contrasting  
873 nonstructural carbohydrate dynamics of tropical tree seedlings under water deficit and  
874 variability. *New Phytologist* 205, 1083–1094. doi:<https://doi.org/10.1111/nph.13134>.

875 O'Brien, M. J., Leuzinger, S., Philipson, C. D., Tay, J., and Hector, A. (2014). Drought survival of  
876 tropical tree seedlings enhanced by non-structural carbohydrate levels. *Nature Climate*  
877 *Change* 4, 710 EP. Available at: <http://dx.doi.org/10.1038/nclimate2281>.

878 O'Brien, M. J., Valtat, A., Abiven, S., Studer, M. S., Ong, R., and Schmid, B. (2020). The role of  
879 soluble sugars during drought in tropical tree seedlings with contrasting tolerances. *Journal*  
880 *of Plant Ecology* 13, 389–397. doi:10.1093/jpe/rtaa017.

881 Palacio, S., Hoch, G., Sala, A., Körner, C., and Millard, P. (2014). Does carbon storage limit tree  
882 growth? *New Phytologist* 201, 1096–1100. doi:10.1111/nph.12602.

883 Peltier, D. M., Nguyen, P., Ebert, C., Koch, G. W., Schuur, E. A., Ogle, K., (2023). Moisture stress  
884 limits radial mixing of non-structural carbohydrates in sapwood of trembling aspen, *Tree*  
885 *Physiology*, tpad083, <https://doi.org/10.1093/treephys/tpad083>

886 Piispanen, R., and Saranpää, P. (2001). Variation of non-structural carbohydrates in silver  
887 birch (*Betula pendula* Roth) wood. *Trees* 15, 444–451. doi:10.1007/s004680100125.

888 Piovesan, G., and Biondi, F. (2021). On tree longevity. *New Phytologist* 231, 1318–1337.  
889 doi:<https://doi.org/10.1111/nph.17148>.

890 Piper, F. I., and Paula, S. (2020). The role of nonstructural carbohydrates storage in forest  
891 resilience under climate change. *Current Forestry Reports* 6, 1–13. doi:10.1007/s40725-019-  
892 00109-z.

893 Piper, F. I., Gundale, M. J., and Fajardo, A. (2015). Extreme defoliation reduces tree growth but  
894 not C and N storage in a winter-deciduous species. *Annals of Botany* 115, 1093–1103.  
895 doi:10.1093/aob/mcv038.

896 Plavcová, L., Hoch, G., Morris, H., Ghiasi, S., and Jansen, S. (2016). The amount of parenchyma  
897 and living fibers affects storage of nonstructural carbohydrates in young stems and roots of  
898 temperate trees. *American Journal of Botany* 103, 603–612. doi:10.3732/ajb.1500489.

899 Poorter, L., and Kitajima, K. (2007). Carbohydrate storage and light requirements of tropical  
900 moist and dry forest tree species. *Ecology* 88, 1000–1011. doi:[https://doi.org/10.1890/06-](https://doi.org/10.1890/06-0984)  
901 0984.

902 R Core Team (2023). *R: A Language and Environment for Statistical Computing*. R  
903 Foundation for Statistical Computing, Vienna, Austria. <<https://www.R-project.org/>>.

904

905 Resco de Dios, V., and Gessler, A. (2021). Sink and source co-limitation in the response of  
906 stored non-structural carbohydrates to an intense but short drought. *Trees* 35, 1751–1754.  
907 doi:10.1007/s00468-021-02116-9.

908 Richardson, A. D., Carbone, M. S., Keenan, T. F., Czimczik, C. I., Hollinger, D. Y., Murakami, P., et  
909 al. (2013). Seasonal dynamics and age of stemwood nonstructural carbohydrates in  
910 temperate forest trees. *New Phytologist* 197, 850–861. doi:10.1111/nph.12042.

911 Rosell, J. A., Piper, F. I., Jiménez-Vera, C., Vergílio, P. C., Marcati, C. R., et al. (2021). Inner bark  
912 as a crucial tissue for non-structural carbohydrate storage across three tropical woody plant  
913 communities. *Plant, Cell & Environment* 44, 156–170. doi:10.1111/pce.13903

914 Rowland, L., Ramírez-Valiente, J. A., Hartley, I. P. and Mencuccini, M. (2023), How woody  
915 plants adjust above- and below-ground traits in response to sustained drought. *New*  
916 *Phytologist*, 239: 1173-1189. <https://doi.org/10.1111/nph.19000>

917 Sala, A., Woodruff, D. R., and Meinzer, F. C. (2012). Carbon dynamics in trees: Feast or famine?  
 918 *Tree Physiology* 32, 764–775. doi:10.1093/treephys/tpv143.

919 Salleo, S., Lo Gullo, M. A., Trifilò, P., and Nardini, A. (2004). New evidence for a role of vessel-  
 920 associated cells and phloem in the rapid xylem refilling of cavitated stems of *laurus nobilis* l.  
 921 *Plant, Cell & Environment* 27, 1065–1076. doi:10.1111/j.1365-3040.2004.01211.x.

922 Schneider, C. A., Rasband, W. S., and Eliceiri, K. W. (2012). NIH image to imagej: 25 years of  
 923 image analysis. *Nature Methods* 9, 671–675. doi:10.1038/nmeth.2089.

924 Sevanto, and M., Sanna, Dickman, L Turin, Pangle, Robert E., and Pockman, W. T. (2014). How  
 925 do trees die? A test of the hydraulic failure and carbon starvation hypotheses. *Plant, Cell &*  
 926 *Environment* 37, 153–161. doi:10.1111/pce.12141.

927 Sierra, C. A., Ceballos-Núñez, V., Hartmann, H., Herrera-Ramírez, D., and Metzler, H. (2022).  
 928 Ideas and perspectives: Allocation of carbon from net primary production in models is  
 929 inconsistent with observations of the age of respired carbon. *Biogeosciences* 19, 3727–3738.  
 930 doi:10.5194/bg-19-3727-2022.

931 Teskey, R. O., Saveyn, A., Steppe, K., and McGuire, M. A. (2008). Origin, fate and significance of  
 932 CO<sub>2</sub> in tree stems. *New Phytologist* 177, 17–32. doi:https://doi.org/10.1111/j.1469-  
 933 8137.2007.02286.x.

934 The GIMP Development Team. (2019). *GIMP*. Retrieved from <https://www.gimp.org>

935 Trumbore, S., Czimczik, C. I., Sierra, C. A., Muhr, J., and Xu, X. (2015). Non-structural carbon  
 936 dynamics and allocation relate to growth rate and leaf habit in California oaks. *Tree*  
 937 *Physiology* 35, 1206–1222. doi:10.1093/treephys/tpv097.

938 Wiley, E., and Helliker, B. (2012). A re-evaluation of carbon storage in trees lends greater  
 939 support for carbon limitation to growth. *New Phytologist* 195, 285–289.  
 940 doi:https://doi.org/10.1111/j.1469-8137.2012.04180.x.

941 Wiley, E., Casper, B. B., and Helliker, B. R. (2017). Recovery following defoliation involves  
 942 shifts in allocation that favour storage and reproduction over radial growth in black oak.  
 943 *Journal of Ecology* 105, 412–424. doi:https://doi.org/10.1111/1365-2745.12672.

944 Wright, I. J., Reich, P. B., Westoby, M., Ackerly, D. D., Baruch, Z., Bongers, F., et al. (2004). The  
 945 worldwide leaf economics spectrum. *Nature* 428, 821–827. doi:10.1038/nature02403.

946 Würth, M. K. R., Peláez-Riedl, S., Wright, S. J., and Körner, C. (2005). Non-structural  
 947 carbohydrate pools in a tropical forest. *Oecologia* 143, 11–24. doi:10.1007/s00442-004-  
 948 1773-2.

949 Zuleta, D., Arellano, G., Muller-Landau, H. C., McMahon, S. M., Aguilar, S., Bunyavejchewin, S., et  
 950 al. (2022). Individual tree damage dominates mortality risk factors across six tropical forests.  
 951 *New Phytologist* 233, 705–721. doi:https://doi.org/10.1111/nph.17832.

952 Zweifel, R., Haeni, M., Buchmann, N., and Eugster, W. (2016). Are trees able to grow in periods  
 953 of stem shrinkage? *New Phytologist* 211, 839–849. doi:https://doi.org/10.1111/nph.13995.

954 Zweifel, R., Zimmermann, L., Zeugin, F., and Newbery, D. M. (2006). Intra-annual radial  
 955 growth and water relations of trees: implications towards a growth mechanism. *Journal of*  
 956 *Experimental Botany* 57, 1445–1459. doi:10.1093/jxb/erj125.

957  
 958 **9. Tables**

959 Table 1: Species names and traits: wood storage strategy, growth rates, mortality rates,  
 960 phenology, and the sampling dates.

	Growth rate	Mortality rate	Leaf phenology	
Species name	(cm/year)	(%/year)	Storage strategy	(~% of leaf loss)
<i>Ocotea leucoxylon</i> (Sw.)				
Laness	0.295	5.7	Parenchyma	Evergreen (~30)
<i>Sacoglottis guianensis</i>				
Benth.	0.72	5.0	Parenchyma	Semi-deciduous (~45)
<i>Dacryodes microcarpa</i>				
Cuart.	0.078	1.6	Fiber	Semi-deciduous (~60)

961

962

963

964