

1 **TREE GROWTH AND STEM CARBON ACCUMULATION IN HUMAN-MODIFIED AMAZONIAN FORESTS**  
2 **FOLLOWING DROUGHT AND FIRE**

3

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30 **ABSTRACT**

31 Human-modified forests are an ever increasing feature across the Amazon Basin, but little is known  
32 about their ability to absorb carbon and how it can be affected by extreme climatic events. Here we  
33 assess for the first time the impacts of human-driven disturbance in combination with El Niño-  
34 mediated droughts and fires on tree growth and carbon accumulation. We found that after 2.5 years  
35 of continuous measurements, there was no difference in stem carbon accumulation between  
36 undisturbed and human-modified forests. Furthermore, the extreme drought caused by the El Niño  
37 did not affect carbon accumulation rates in surviving trees. In recently burned forests trees grew  
38 significantly more than in unburned ones, regardless of their history of previous human disturbance.  
39 Wood density was the only significant factor that helped explain the difference in growth between  
40 trees in burned and unburned forests, with low wood density trees growing significantly more in  
41 burned sites. Our results suggest stem carbon accumulation is resistant to human disturbance and  
42 one-off extreme drought events, and it is stimulated immediately after wildfires. However, these  
43 results should be seen with caution – without accounting for carbon losses, we cannot fully  
44 understand the impacts of drought and fire in the carbon balance of human-modified forests.

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47 **KEY WORDS**

48 Drought, wildfire, tree growth, tropical forests, degradation, ENSO

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51 **RUNNING HEAD**

52 El Niño effects on tree growth

53

54 **INTRODUCTION**

55

56 The Amazon stores c. 86Pg of carbon [1], an amount equivalent to almost 10 years of combined  
57 global emissions from fossil fuels and the cement industry [2]. This large carbon reservoir has  
58 historically been threatened by deforestation, with large NGO-led campaigns bringing the issue to  
59 the public and pressuring governments for measures to effectively stop forest loss [3]. However,  
60 wildfires, i.e. fires that escape agricultural lands and invade forests, have been an often neglected  
61 although significant threat to Amazonian forests, substantially decreasing carbon stocks [4] and  
62 biodiversity [5,6]. In the past decades, forest fires were directly linked to deforestation rates [7],  
63 however, this is not the case anymore – although deforestation in the Brazilian Amazon has  
64 remained somewhat stable since 2009 [8], forest fires are increasing in number [9]. This surge in  
65 wildfire occurrence is a consequence of a combination of factors: greater frequency of extreme  
66 droughts [10], the indirect impacts of deforestation that creates flammable edges [11] and reduces  
67 regional rainfall [12], the spread of selective logging that increases forest flammability [13] and the  
68 prevalence of ignition sources used in Amazonian agricultural systems [14]. As a result, wildfires  
69 have become the new norm in the parts of the Amazon Basin most affected by human disturbance,  
70 especially during extreme dry years [15,16].

71

72 More frequent and more intense droughts are expected across Amazonia in this century [17–19].  
73 Extreme droughts are known to double tree mortality rates in tropical rainforests, reverting them  
74 from carbon sinks to sources [20,21]. Drought-affected rainforest trees die either because they  
75 cannot move water from their roots to their leaves, known as hydraulic failure [22], or because they  
76 close their stomata in order not to lose water but, as a consequence, do not have enough sugars to  
77 keep their metabolism, a process known as carbon starvation [23]. This increase in tree mortality  
78 rates leads to more openings in the forest canopy, turning drought-affected forests more flammable  
79 due to the accumulation of fuel (i.e. branches and leaves) on the forest floor and the higher  
80 incidence of sun and wind on the understory [13]. When drought-affected tropical rainforests catch  
81 fire, they experience even higher rates of tree mortality, sometimes close to 50% [24]. This large-  
82 scale mortality is then followed by severe structural and compositional changes [25] and significant  
83 reductions of their carbon stocks [4].

84

85 However, the influence of drought or wildfires on the growth of the surviving trees remains poorly  
86 understood. Results from drought experiments on undisturbed forests showed that radial tree  
87 growth was negatively impacted only after years of continuous rainfall exclusion [26,27]. This has

88 been corroborated by results from field monitoring, which showed that radial tree growth was not  
89 affected by a one-off extreme drought [28]. When evaluating the impacts of wildfires on tree growth,  
90 studies in Amazonia have focused solely on re-sprouting dynamics (e.g. 23,24), and have not  
91 examined whether radial growth of the few surviving trees is altered. The one exception is a study  
92 conducted in the Amazon-savannah boundary [31], which found that low-severity fires increased  
93 post-fire tree growth. Notably, no studies to date have investigated the impacts of either extreme  
94 droughts or wildfires on trees growing in human-modified forests. For example, it is unclear whether  
95 droughts and wildfires affect tree growth and carbon accumulation in similar ways between  
96 undisturbed primary forests and those that have been human modified, or whether radial growth is  
97 inhibited in the years following drought and wildfires. It seems therefore crucial that we develop a  
98 better understanding of tree growth and stem carbon uptake in these altered systems, given the  
99 high rates of human-driven forest disturbance across the Amazon [32], the increasing ubiquity of  
100 forest fires and the paucity of studies examining the responses of surviving trees.

101

102 The 2015 El Niño event provided a valuable opportunity to address these knowledge gaps. The  
103 region of Santarém, in the Brazilian Amazon, was particularly affected by drought during this El Niño  
104 [33] and millions of hectares of forests burned. Prior to the El Niño, we had established 18  
105 permanent forest plots in the region, where we had been measuring tree growth monthly in c.1000  
106 individuals. These plots were distributed along a gradient of human disturbance, from undisturbed  
107 primary forests, to logged primary forests, logged-and-burned primary forests and secondary forests  
108 (i.e. those regrowing on land previously cleared for agriculture). All our plots were severely affected  
109 by the El Niño drought, and some were also affected by the extensive wildfires that affected the  
110 region (Withey *et al.* this issue). We draw on this unique dataset to investigate the responses of  
111 human-modified forests to El Niño-mediated droughts and fires, asking four questions 1) How does  
112 tree growth and stem carbon accumulation compare between forest disturbance classes?, 2) Has the  
113 El Niño drought affected relative tree growth and carbon accumulation rates across the disturbance  
114 gradient?, 3) Is the post-El Niño growth and carbon accumulation of trees affected by drought  
115 different from those affected by both drought and fire?, and 4) Which stem or forest structure  
116 factors can influence differences in growth and carbon accumulation between trees located in  
117 drought-affected plots from those located in plots affected by both drought and fire?

118

119

## 120 MATERIAL AND METHODS

121

122 **(a) Study area**

123 The study was conducted in three municipalities of the eastern part of the Amazon Basin: Santarém,  
124 Belterra and Mojuí-dos-Campos (hereafter Santarém region). The climate in the region is hot and  
125 humid, with an annual average of 25 °C, 86% relative humidity and 1920 mm of rain [34]. The region  
126 has a marked dry season that usually lasts for four months, from August to November, when  
127 precipitation is <100 mm/month (Fig. S1). Soils are rich in clay, but nutrient poor [35]. Data were  
128 collected in 18 permanent plots (250 x 10 m) distributed along a gradient of pre-El Niño human  
129 disturbance: undisturbed forests (n = 5), logged forests (n = 5), logged-and-burned forests (n = 4),  
130 and secondary forests (n = 4). Plots were located in *terra firme* forests situated between 1.5 and 97  
131 km apart (Fig. S2). In December 2015, seven of our study plots burned, including three of previously  
132 undisturbed forests, four of previously logged forests, and one of previously logged-and-burned  
133 forest (Table S1).

134

135 **(b) Tree growth and stem carbon accumulation**

136 In all plots, trees were measured and identified to species level in 2014. We then installed 50  
137 dendrometer bands in each plot, stratifying by tree size class: 10-20cm diameter at breast height  
138 (DBH, 1.3m from the forest floor), 20-30cm DBH, 30-40 cm DBH, 40-50 cm DBH, and >50 cm DBH.  
139 When a plot did not have 10 trees in a given size class, we distributed the remaining dendrometers  
140 evenly across the other size classes. Between July 2015 and December 2017, tree growth was  
141 measured monthly with digital callipers. In the case of a dendrometer been found damaged or a tree  
142 having suddenly died, the band would be removed immediately and promptly reinstalled. In the  
143 burned plots, the heat overstretched the metal springs and all dendrometers were reinstalled within  
144 four weeks of the fires. Monthly tree growth was converted into stem carbon accumulation by using  
145 a biomass equation for tropical trees [36] and assuming carbon content to represent 50% of biomass.  
146 The equation used takes into consideration the tree measured growth, height, and wood density.  
147 We allowed negative growth values, even though these reflect water loss from the bark and not a  
148 true decrease in tree size [37]. This was because some of the positive growth values are due to water  
149 accumulation in the bark, and the keeping of negative values is therefore necessary to balance out  
150 the fluctuating water content over the year [37].

151

152

153 **(c) Factors influencing tree growth and carbon accumulation**

154 Based on the literature, we selected six factors that could possibly influence post-fire tree growth  
155 and the consequent carbon accumulation on the stem: DBH, height, wood density, fire intensity and

156 two measures reflecting the degree of competitive release from fires – the change in liana load, and  
157 the change in basal area in the surrounding forest. The DBH and the height of each tree were  
158 assessed during a re-census of all plots in 2016. Wood density was derived from the Global Wood  
159 Density database [38], based on the species identification and filtering by South American tropical  
160 regions. We measured the maximum char height on all burned stems as a proxy for fire intensity.  
161 Liana loads were determined during both the 2014 and the 2016 censuses. This is an estimate of  
162 how much of the crown of a given tree is infested by liana leaves, ranging from 0, 1-25, 26-50, 51-75,  
163 and 76-100% [39]. Finally, the basal area of live stems was calculated in a 10 x 10 m plot surrounding  
164 each tree in both 2014 and 2016. Changes in both liana load and surrounding basal area were  
165 calculated as the difference between the 2014 and the 2016 values for each tree. We expected that  
166 the high mortality of lianas [40] and trees [41] immediately after wildfires would result in less  
167 competition for light and water among the surviving trees, thus likely influencing tree growth [42].

168

#### 169 **(d) Data analysis**

170 To investigate whether there were any differences in radial growth and stem carbon accumulation  
171 between trees of different forest disturbance classes, we considered only individuals which were  
172 continuously measured over a 2.5 year period from July 2015 until December 2017 ( $n = 385$ ),  
173 therefore excluding all stems located in burned plots from this analysis. We used ANOVAs followed  
174 by *post-hoc* Tukey tests to examine whether there were any differences in the mean cumulative  
175 growth and carbon between the forest disturbance classes. The tests were ran using both the  
176 absolute and normalized (growth/DBH) growth of each stem. For each test we calculated the eta-  
177 square ( $h^2$ ), which is a measure of effect size and corresponds to the proportion of the total variation  
178 in the data that can be attributed to the explanatory variable.

179

180 We used a temporal comparison to assess the impacts of the El Niño-induced extreme drought. For  
181 this we conducted two analyses. First, we compared the total dry season growth and carbon  
182 accumulation of trees measured continuously during the 2015 El Niño-mediated drought with the  
183 two following dry seasons, 2016 and 2017 ( $n = 385$ ). We built generalized linear mixed-effects  
184 models (GLMM) to assess whether dry season growth and carbon accumulation were influenced by  
185 forest disturbance, year or an interaction between both. In these models, tree and plot identities  
186 were set as random effects. Second, we investigated whether relative growth and carbon  
187 accumulation rates were influenced by dry season intensity, measured by the climatological water  
188 deficit (CWD). To calculate the relative growth and stem carbon accumulation rates, we used the  
189 interval growth between months. CWD was defined as precipitation in a given month (mm), minus

190 evapotranspiration (100 mm), minus the previous month CWD; following [43]. Precipitation data  
191 was obtained from CHIRPS [44]. We built two sets of GLMMs, using either the relative growth or  
192 carbon accumulation rates as response variables. CWD was the explanatory variable in these models,  
193 while random effects included tree identity, study plot and year.

194

195 To compare the annual growth and carbon accumulation of trees located in drought-affected plots  
196 with those of trees located in plots affected by both drought and fire, we used data of individuals  
197 with continuous measurements throughout 2017 ( $n = 545$ ), which was the only comparable period  
198 given that fires damaged the dendrometers. We then ran three 2-way ANOVAs: on the first we used  
199 cumulative tree growth at the end of 2017 as the response variable, on the second we used the  
200 normalized growth (growth/DBH), while on the third we used the annual carbon accumulation. All  
201 ANOVAs used pre-El Niño forest disturbance class and fire (burned or unburned in 2015-16) as  
202 explanatory variables. After each test we calculated their eta-square ( $h^2$ ).

203

204 Finally, we used a matching approach commonly used in landscape ecology (e.g.[45]) to investigate  
205 which factors predict post-fire tree growth and carbon accumulation. The matching approach linked  
206 individual trees in drought-and-fire-affected forests with functionally comparable stems in drought-  
207 affected forests. This was essential to answer our research question, as fire potentially imposes a  
208 non-random mortality, killing more small-stemmed and low wood density trees [46]. As such, an  
209 unmatched comparison would not be able to fully distinguish differences in tree growth due to the  
210 newly altered functional characteristics of a forest (for example, if only large stems survived) or due  
211 to post-fire changes in forest conditions that may alter the growth of individual stems (e.g. decrease  
212 in liana infestation due to fire-induced mortality). For trees to be matched, they had to belong to the  
213 same pre-El Niño disturbance class and the matched stem had to be within a 10% margin of both the  
214 DBH and wood density of the burned forest stem. When more than one tree in unburned forests  
215 met the matching criteria, we favoured the one with the closest DBH to the tree in the burned forest.  
216 This choice was based on the fact that DBH is quadratic in the biomass equation used [36], as  
217 opposed to wood density which is only elevated to the power of one. In total, 128 trees could be  
218 matched (i.e. 64 pairs).

219

220 After the matching, we ran linear models between the matched trees in each disturbance class to  
221 examine if either the growth or the carbon accumulation of trees in unburned forests could predict  
222 that of trees in burned forests. For each pair, we then calculated the difference in both total growth  
223 and carbon accumulated by the end of 2017. We ran generalized linear models to investigate which

224 stem and forest structure factors could be influencing these differences in radial growth and stem  
225 carbon accumulation between matched trees. Models included forest disturbance class, the DBH,  
226 the height, the wood density, the char height, the  $\Delta$  liana load (i.e. 2016 - 2014) and the  $\Delta$  basal area  
227 of surrounding live stems (i.e. 2016 - 2014) of the fire-affected tree as explanatory variables. Prior to  
228 running the models, we checked for collinearity between explanatory variables and none was found  
229 (Fig. S3). To facilitate our understanding of the effect size of each explanatory variable, they were all  
230 standardized between 0 and 1. All analyses were performed in R version 3.4.0 using the BBmisc,  
231 corplot, MASS, and sjstats packages [47–50].

232

233

## 234 **RESULTS**

235

### 236 **Tree growth and stem carbon accumulation across human-modified forests**

237 After 2.5 years of continuous measurements, thus focusing only on trees located in unburned sites,  
238 the mean individual growth was significantly higher in trees located in secondary forests (Fig. 1, Fig.  
239 S4) than when compared to trees in all other forest classes ( $F_{(3, 381)} = 14.27, p < 0.001, h^2 = 0.10$ ;  
240 Tukey tests involving secondary forests, all  $p < 0.001$ ). However, there was no significant difference in  
241 carbon accumulation between any of the forest classes. The higher growth of trees in secondary  
242 forests was consistent across DBH size classes (Fig. S5). These results were also consistent whether  
243 using absolute or normalized tree growth.

244

### 245 **El Niño impacts on dry season growth and carbon accumulation**

246 While dry season growth was significantly higher in the post-El Niño years (Fig. 2a; both  $p < 0.05$ ); dry  
247 season carbon accumulation was not significantly influenced by the El Niño-mediated drought (Fig.  
248 2b). Regardless of the year, trees in logged forests grew significantly less and accumulated  
249 significantly less carbon (both  $p < 0.05$ ). In trees situated in undisturbed, logged and secondary  
250 forests (Fig. S6), there was a weak but significant relationship between growth rates during the dry  
251 season and the climatological water deficit (all  $p < 0.001$ ) – the more negative the deficit, the lower  
252 the growth. However, monthly carbon accumulation rates were only significantly affected by CWD in  
253 logged and secondary forests (Fig. S7).

254

### 255 **Growth and stem carbon accumulation between trees in burned and unburned forests**

256 When analysing data from all surviving stems ( $n = 545$ ) in forests affected by drought and those  
257 affected by drought and fire during the 2015 El Niño, both growth and carbon accumulation in the

258 end of 2017 were significantly higher in trees located in burned plots ( $F_{(1,389)} = 41.64$ ,  $h^2_{\text{fire}} = 0.09$  and  
259  $F_{(1,389)} = 22.68$   $h^2_{\text{fire}} = 0.06$ , respectively; both  $p < 0.0001$ , Fig. 3). This pattern was maintained  
260 regardless of tree size or pre-El Niño forest disturbance class (Fig. S8-S10). Results were consistent  
261 whether using absolute or normalized tree growth.

262

### 263 **Factors influencing differences in tree growth and stem carbon accumulation**

264 When focusing only on the matched trees ( $n = 128$  trees, 64 pairs), neither the growth nor carbon  
265 accumulated in trees located in forests that burned during the 2015 El Niño could be predicted by  
266 their matched pairs in forests only affected by drought (all  $R^2 \leq 0.28$ ,  $p > 0.05$ ; Fig. S11). Of all the  
267 factors examined with a generalized linear model to possibly explain differences in growth and  
268 carbon accumulation between matched trees, only wood density was significant ( $p = 0.05$ ,  $\beta = -1.94$ ;  
269 and  $p < 0.05$ ,  $\beta = -3.67$ , respectively). Wood density had a negative relationship with the differences  
270 in growth and carbon accumulation between burned and unburned trees, thus the lighter the wood  
271 density, the greater the increase in growth in stems in recently burned forests (Fig. 4).

272

273

## 274 **DISCUSSION**

275 Our novel results provide important insights into tree growth and carbon accumulation in human-  
276 modified Amazonian forests, and the interaction between forest disturbance and extreme drought  
277 and fire events. Surprisingly, there was no significant difference in overall carbon accumulation  
278 between trees in undisturbed and human-modified forests. Furthermore, the extreme El Niño-  
279 mediated drought did not seem to inhibit carbon accumulation in surviving trees. We were also able  
280 to assess the impacts of wildfires on the few surviving trees and the factors affecting post-fire  
281 growth, something never done before in humid tropical forests. We found that trees situated in  
282 forests that burned during the 2015 El Niño presented a significantly higher radial growth and stem  
283 carbon accumulation than trees in forests only affected by drought, and that this difference was  
284 more pronounced in lighter wood density stems. We discuss these results in light of the increasing  
285 ubiquity of human-modified Amazonian forests and of the increased frequency of drought and fire  
286 events.

287

### 288 **The importance of human-modified forests for carbon accumulation**

289 Over a 2.5-year period of continuous monitoring, trees in secondary forests grew significantly faster  
290 than those in undisturbed and disturbed primary Amazonian forests, a result that is consistent with  
291 others from elsewhere in the Neotropics [51]. However, these higher levels of individual growth did

292 not lead to more carbon accumulation, with trees in undisturbed, disturbed and secondary forests  
293 accumulating comparable amounts of carbon. The apparent discrepancy between the results of  
294 radial growth and stem carbon accumulation can be explained by the dominance of lower wood  
295 density species in secondary forests [52]. For example, when we consider a 20-cm DBH and 15-m tall  
296 stem of a low-wood density species commonly found in secondary forests, *Jacaranda copaia*, a 2-cm  
297 growth results in an increment of 0.66 kg of C. However, in a hyper-abundant primary forest species,  
298 *Eschweilera coriacea* [53], a stem of the same size and height experiencing the same growth will  
299 incorporate 1.57 kg of C, a difference of 236%. To achieve a similar amount of carbon accumulation,  
300 this hypothetical individual of *Jacaranda copaia* would have to grow 3.1 cm; i.e. it would have to  
301 grow 1.6 times more than the *Eschweilera coriacea* to accumulate the same amount of carbon.  
302 Therefore, although trees in secondary forests are showing higher rates of radial growth, this is  
303 compensated by their lower wood density, resulting in similar levels of carbon accumulation across  
304 all forest classes.

305

#### 306 **Drought effects on tree growth and carbon accumulation**

307 The El Niño-mediated drought negatively affected tree growth, but had no significant impact on  
308 overall stem carbon accumulation. This appears to indicate that low wood density trees, i.e. those  
309 that contribute less to carbon accumulation, were the most affected by the 2015 drought. In  
310 Amazonian forests, low wood density tree species tend to be less resistant to extreme droughts [20],  
311 as they present high turgor loss points and high osmotic potential [54]. In other words, when there is  
312 less water available, the leaves of low wood density trees are more likely to wilt, impacting  
313 photosynthesis [55] and, as a consequence, growth rates. However, the effects of the El Niño-  
314 mediated drought appeared to be transient, given that growth rates remained uninhibited in the  
315 following dry seasons. Furthermore, the weak relationships between climatological water deficit and  
316 both dry season growth and carbon accumulation rates suggest that trees in both undisturbed and  
317 human-modified forests are adapted to seasonal droughts. This result is to be expected, as the  
318 distribution of Amazonian tree species follows a dry-tolerance pattern, which consists in more  
319 drought-tolerant taxa occurring in the parts of the basin that every year experience some months of  
320 little rainfall [56], such as the Santarém region. It is important to note however, that the dry seasons  
321 of 2015 and 2017 were stronger than those between 1970-1999 – even in 1997, the year of the  
322 strongest El Niño on record [57], the maximum climatological water deficit in eastern Amazonia was  
323 approximately -200 mm [43], while in 2015 and 2017 it was of -368 mm and -316 mm, respectively.  
324 So far, eastern Amazonian trees seem resistant to the current drier climate, continually  
325 accumulating carbon despite more intense dry seasons than in the previous 30 years.

326

327 **Wildfire effects on tree growth and carbon accumulation**

328 Trees in burned forests both grew more and accumulated more carbon than trees located in plots  
329 that only experienced drought during the El Niño. This is a completely novel finding from tropical  
330 rainforests. In other ecosystems, fire effects on tree growth lead to conflicting results: while low-  
331 intensity fires can increase tree growth in savannahs [58], it can suppress radial growth in temperate  
332 forests [59]. The mechanisms behind these changes in growth rates remain unclear. In our sites,  
333 changes in post-fire tree growth were not explained by tree size, tree height, forest disturbance class,  
334 or proxies of fire intensity and competitive release (from lianas and other trees). Wood density was  
335 the only significant factor explaining differences in tree growth and carbon accumulation between  
336 stems located in burned plots and those located in drought-affected plots, with lower wood density  
337 trees in burned forests growing more than their counterparts in unburned forests. Given that our  
338 measures of competitive release were not important predictors of differences in tree growth  
339 between burned and unburned trees, it is unlikely that low wood density trees experienced an  
340 enhanced growth due to greater light or water availability. Most likely, low wood density trees were  
341 benefitting from the large pulse of nutrients released by the combustion of organic matter. In  
342 general, low wood density trees have acquisitive life strategies, heavily investing in rapid growth [60];  
343 while high wood density tree species are more conservative, with considerably slower growth rates  
344 [61]. The sudden input of nutrients has probably led to a disproportional investment in growth by  
345 low wood density trees.

346

347 **Amazonian forests in the Anthropocene**

348 Tropical ecosystems face growing pressure from a combination of both global and local stressors  
349 [64]. Across Amazonia, a global stressor, climate change, is predicted to increase the frequency of  
350 two local stressors – extreme droughts and associated fires [9,17]. Other local stressors, such as  
351 selective logging, newly-created forest edges and large infrastructure projects, are turning human-  
352 modified forests into a prevalent feature across the basin [65,66]. Understanding ecosystem-level  
353 responses to these growing anthropogenic pressures can help predict their consequences, and  
354 opens up opportunities to mitigate their worst effects. Our study shows the relative resilience of  
355 tree growth and subsequent carbon accumulation to one-off droughts, and suggests that growth  
356 rates can even increase after wildfires. Still, stem growth is just one part of a forest's carbon balance:  
357 despite the spike in stem carbon accumulation, the carbon balance in burned forests is still largely  
358 negative – tree mortality following fires is extremely high [24,63] and cannot be compensated by the  
359 growth of the few surviving trees. Previous studies in Amazonia have shown that three years after

360 fires, forests can lose c. 50% of its individuals and 75 Mg C ha<sup>-1</sup>. This can hardly be compensated by  
361 the remaining trees accumulating an extra 1kg C, and demonstrate the importance of avoiding  
362 wildfires in humid tropical forests.

363

364

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377

378

### 379 **AUTHORS CONTRIBUTIONS**

380 EB, YM and JB designed the study. EB and JF were responsible for plot selection and subsequent  
381 authorizations from landowners. EB, PB, ACNC, FF, LCR, and MMMS performed data collection. EB  
382 conducted all statistical analyses. EB and JB wrote the manuscript with critical inputs from all  
383 authors.

384

385

### 386 **DATA ACCESSIBILITY**

387 The data used in this paper has been deposited under the DOI:

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389

390

### 391 **COMPETING INTERESTS**

392 We have no competing interests.

393

394

395

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564 **FIGURES**

565

566 **Fig 1.** Mean individual growth (left) and carbon (right) accumulated over three years in undisturbed  
567 primary forests (green), logged forests (blue), logged-and-burned forests (orange), and secondary  
568 forests (red).

569

570 **Fig 2.** Cumulative tree (a) growth and (b) carbon accumulation during the dry seasons of 2015, 2016,  
571 and 2017 in undisturbed, logged, logged-and-burned, and secondary forests.

572

573 **Fig 3.** Mean individual tree (a) growth and (b) carbon accumulated along 2017 in trees situated in  
574 forests that burned (red) or were only affected by drought (blue) during the 2015 El Niño. Forest  
575 classes correspond to forest condition prior the onset of the El Niño-mediated fires.

576

577 **Fig. 4.** Coefficient plots of the factors affecting the difference in (a) growth and (b) carbon  
578 accumulation between trees in burned and unburned forests.

579