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1   **Climate and fragmentation affect forest structure in the southern border of**

2   **Amazonia**

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Abstract

Background: The remaining forests in the extensive contact zone between southern Amazonia (seasonal rain forest) and the Cerrado (savanna) biomes are at risk due to intense land-use and climate change.

Aims: To explore the vulnerability of these transitional forests to changes in land use and climate, we evaluated the effects of fragmentation and climatic variables on forest structure.

Methods: We measured the diameter and height of 14,185 trees with diameter $\geq$ 10 cm at 24 forest plots distributed over an area of 25,000 km^2. For each plot, we obtained data on contemporary fragmentation and climatic variables.

Results: Forest structure variables (height, diameter, height:diameter allometry, biomass) varied significantly both within and among plots. The height, H:D and biomass of trees were positively correlated with annual precipitation and fragment area.

Conclusions: The association between forest structure and precipitation indicates that these forests plots are likely to be vulnerable to dry season intensification anticipated for the southern edge of the Amazon. Additionally, the reduction in the fragment area may contribute to reductions in forest biomass and tree height, and consequently ecosystem carbon stocks. Given the likely susceptibility of these forests, urgent conservation action is needed to prevent further habitat degradation.

Keywords: allometry; Amazon arc of deforestation; biomass; climate change; habitat fragmentation; precipitation; stem diameter; tree height; transition zone

Introduction

49   Across the Earth's biomes, environmental conditions are expected to be

50 more variable close to the edges than in the core area of each biome, posing potentially

51 ecological and evolutionary challenges to biota towards their biogeographical edges

52 (Safriel et al. 1994; Kark and van Rensburg 2006; Kark et al. 2008). This may be

53 particularly the case in regions subject to rapid environmental change, of which perhaps

54 the most extreme example are the forests of the southern edge of the Amazon rain forest

55 biome, an area affected by high deforestation rates and subject to significant recent and

56 forecast climate change. Thus, here the advance of the agricultural frontier has already

57 resulted in converting most forest to pasture and cropland, increasingly fragmenting the

58 landscape over the last few decades (Alencar et al. 2004, 2015; Nogueira et al. 2008).

59 The remaining forests are subject to recent climate change, including lengthening of the

60 dry season and increasing incidence of strong droughts (Marengo et al. 2011; Gloor et

61 al. 2015; Feldpausch et al. 2016), trends which are expected to intensify further (e.g.

62 Boisier et al. 2015). The land surface temperature has been rising steadily recently,

63 especially in the south and east of the Amazon region (Jiménez-Muñoz et al. 2013), and

64 the effects of these climatic changes may be exacerbated by changes in land use

65 (Aragão 2012; Silvério et al. 2015). Finally, research elsewhere in Amazonia clearly has

66 indicated that the structure of the tropical forest vegetation is affected by both climate

67 change (e.g. Phillips et al. 2010; Feldpausch et al. 2016) and fragmentation of habitats

68 (e.g. Laurance et al. 1997, 2000; Laurance 2004).

69   Yet few studies have evaluated structural variation among the forests in the

70 southern border region of the Amazon forest biome and its covariation with climate and

71 landscape factors. Exceptions include one analysis of the effects of the interaction

72 between droughts and wildfires on tree mortality at one experimental site (Brando et al.

73 2014), and a landscape study which showed that habitat fragmentation, combined with

74 droughts, increased the susceptibility of the forests to fire (Alencar et al. 2015). We are

75 not aware of a single study that has evaluated the effects of habitat fragmentation and

76 different climate variables across the region's forests using direct, on-the-ground

77 measurement of vegetation structural variables such as tree diameter, height, and

78 biomass.

79   Habitat fragmentation, by decreasing fragment size and increasing forest

80 edges and numbers of fragments, may modify the forest structure in the remaining

81 fragments (Fahring 2003; Haddad et al. 2015). For example, fragment edges are subject

82 to a greater incidence of insolation and increased velocity of winds, resulting in higher

83 temperatures and a drier microclimate than the forest interior (D'Angelo et al. 2004;

84 Laurance 2004; Haddad et al. 2015), which increases tree mortality rates, principally for

85 larger trees (Laurance et al. 2000; Laurance 2004). The death of bigger trees reduces

86 total biomass, height, mean diameter and basal area, especially in the smaller fragments

87 and the areas closest to the forest edge, although with some mortality effects also

88 propagating a few hundred meters into the forest (Laurance 2004; Haddad et al. 2015;

89 Rocha-Santos et al. 2016). Recently, it has even been suggested, based on interpretation

90 of pantropical satellite imagery, that in tropical forests the negative effects on standing

91 biomass and forest structure penetrate as much as 1.5 km into forests (Chaplin-Kramer

92 et al. 2015).

93   In addition to landscape-scale factors, regional climate is related to variation

94 in the forest structure (e.g. Banin et al. 2015). For example, where precipitation and

95 temperature are higher, forests generally have taller trees that accumulate more biomass

96 (Koch et al. 2004; Way and Oren 2010; Feldpausch et al. 2011; Pan et al. 2013; Chave

97 et al. 2014). However, in the very warmest forests the forest structural responses are

98 unclear. There is some evidence that here plants may photosynthesise less and expend

99 more energy on respiration, so potentially accumulating less biomass (Lloyd and

100 Farquhar 2008; Lewis et al. 2013). However, the temperature sensitivity of key

101 respiration processes appears to decline in warmer environments (Atkin et al. 2015,

102 Heskel et al. 2016), rather than increasing exponentially as simple Q_{10} formulations in

103 earlier global vegetation models suggested (Cox et al. 2000), suggesting that the overall

104 sensitivity of biomass stocks to high temperatures might be weaker than many models

105 indicated.

106 Extreme drought events may alter the forest structure. Drought causes

107 mortality, principally in the bigger trees, which are more susceptible to damage in their

108 vascular system (Phillips et al. 2010; Rowland et al. 2015; Bennett et al. 2015;

109 Feldpausch et al. 2016). During drought events, tropical trees may also grow less (e.g.

110 Worbes 1999; Doughty et al. 2015), and if droughts are prolonged or repeated forests

111 eventually accumulate less biomass (Feldpausch et al. 2016; Rowland et al. 2015).

112 In the context of regional land-use and climatic changes in southern Amazonia,

113 and the projected high regional climate sensitivity to global warming (IPCC 2015), it is

114 therefore extremely important to understand how the forest structure is affect by abiotic

115 factors. It may for example help to improve the conservation measures to protect the

116 remaining forest fragments. In this study, we evaluated whether, and to what extent,

117 climatic factors and fragmentation determine variation in the forest structure of the

118 southern Amazon border. We assembled data from permanent plots established across

119 the region close to the natural border of Amazonia with the neighboring Cerrado

120 (savanna) biome, to test hypotheses related to the variation in the forest structure and

121   the factors that determine this variation. We addressed two questions. First, does habitat

122   fragmentation affect the forest structure? We expected that forest cover loss and forest

123   plots present in smaller fragments and/or nearer the edge would have trees with lower

124   height and smaller diameter stems, or with smaller height:diameter (H:D) allometric

125   relationships and reduced biomass, since work elsewhere has shown mortality rates are

126   greater in smaller, more edge-affected fragments, especially for bigger trees (e.g.

127   Laurance et al. 1997, 1998, 2000; Laurance 2004; Chaplin-Kramer et al. 2015). Second,

128   how does the forest structure vary in relation to the climate? We expected that the

129   height and the diameter of stems, the H:D ratio, and biomass were all greater in forest

130   plots that have greater precipitation, and consequently less deficit water, since the

131   greater water availability favours the height growth of the trees, accumulating more

132   biomass (e.g. Feldpausch et al. 2011; Pan et al. 2013; Chave et al. 2014).

133

134   **Materials and methods**

135   Study area

136   We studied 24 forest plots distributed in the so-called 'arc of deforestation'

137   (Nogueira et al. 2008) over an area of ca. 25,000 km^2 (Figure 1 and Table 1). The

138   regional climate is of the Aw (tropical with dry winters) and Am (tropical monsoon)

139   types in the Köppen classification system (Alvares et al. 2013), and originally supported

140   evergreen or semi-evergreen forest vegetation in all cases. Mean annual precipitation

141   and temperature range from 1511 to 2353 mm and from 24.1 to 27.3 °C, respectively

142   (Table 1).
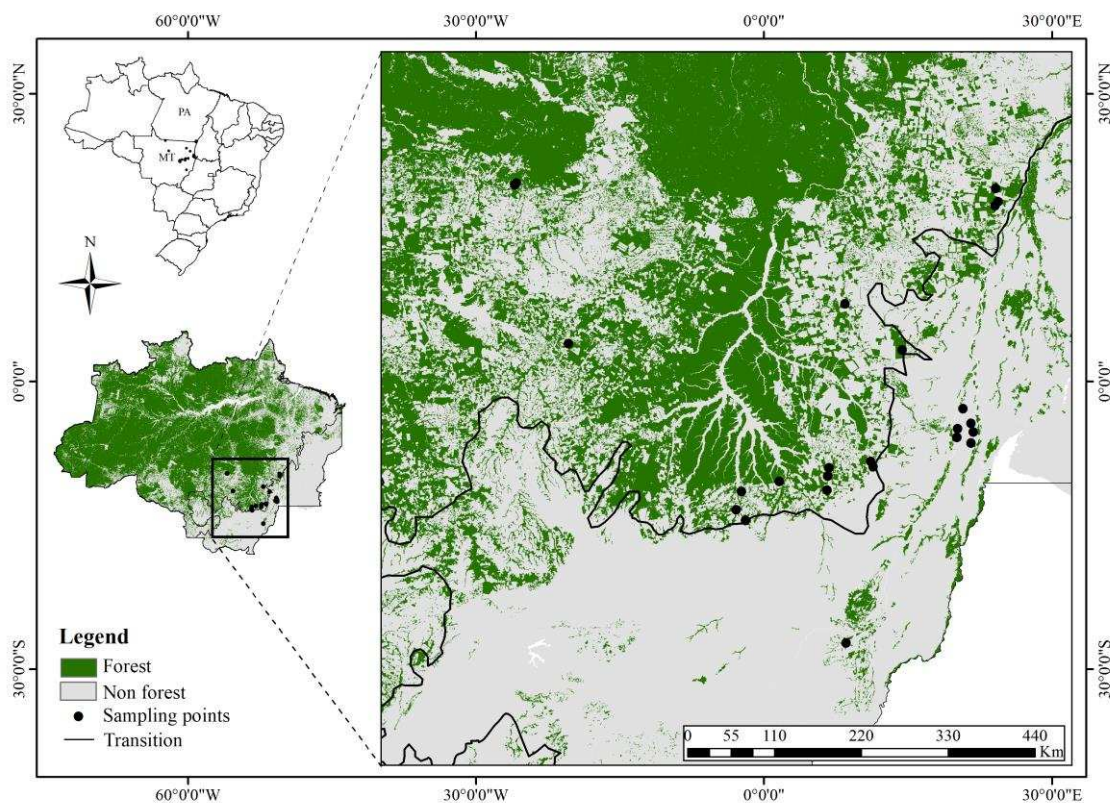
Figure 1. Location of the forests sampled in the southern Amazon border, between

eastern and northern Mato Grosso and southern Pará, Brazil, showing the approximate

biome boundaries based in IBGE (2004). The classification of forest and no forest was

based on the PRODES (Amazon Deforestation Monitoring Project) (INPE 2017). All

plots sampled lie within mature, evergreen or semi-evergreen forest fragments.

Table 1. Characteristics of plots sampled in different tropical forest ecosystems at the

southern Amazon border. FA, fragment area; DE, distance to the forest edge; Prec, total

mean annual precipitation; Temp, mean annual temperature; TB, total above-ground

biomass per hectare; PP, private properties; and CU, conservation unit. In this study, we

used codes ('Plot code') to represent the different types of vegetation: FEP, floresta

estacional perenifólia (seasonal evergreen forest), FTP, floresta estacional perenifólia

em terra preta de índio (seasonal evergreen forest on anthropogenic black earth); FES,

floresta estacional semidecidual (seasonal semi-deciduous forest); FOA, floresta

ombrófila aberta (open rainforest); and FSI, floresta sazonalmente inundável

(seasonally flooded forest). Equivalent forest plot codes are given to indicate

equivalency to those codes used in the ForestPlots.net database (Lopez-Gonzalez et al.

2011) where the data have been deposited.

Plot Code	Forest plot code	Geographical coordinate	Local	AF (ha)	DE (m)	Prec (mm)	Temp (°C)	TB (Mg)
FEP-01	FLO-01	-12.8S and -51.9W	PP	870	1,030	1613	25.5	111.1
FEP-02	FLO-02	-12.8S and -51.9W	PP	2,035	1,000	1621	25.6	144.7
FEP-03	TAN-02	-13.1S and -52.4W	PP	8,432	990	1625	24.9	143.5
FEP-04	TAN-03	-12.8S and -52.3W	PP	16,901	520	1679	25.1	127.4
FEP-05	TAN-04	-12.9S and -52.4W	PP	16,901	329	1662	25	138.3
FEP-06	FRP-01	-11.5S and -51.5W	PP	45,459	3,600	1634	26.9	135.1
FEP-07	POA-01	-11.0S and -52.2W	PP	9,789	1,180	1772	26.1	140.1
FES-01	VCR-02	-14.8S and -52.2W	PP	4,968	1,350	1511	25.2	196.8
FES-02	GAU-02	-13.4S and -53.3W	PP	3,499	160	1701	24.1	91.7
FES-03	SAT-01	-9.8S and -50.5W	PP	17,624	90	1821	26.7	121.8
FES-04	SAA-01	-9.8S and -50.4W	PP	13,039	860	1815	26.8	187.7
FES-05	SAA-02	-9.6S and -50.4W	PP	15,680	2,980	1778	26.6	166.3
FOA-01	SIP-01	-11.4S and -55.3W	PP	12,066	900	1848	25.1	79.2
FOA-02	ALF-01	-9.6S and -55.9W	CU	17,628	5,440	2350	25.5	98.8
FOA-03	ALF-02	-9.6S and -55.9W	CU	17,628	5,410	2353	25.6	160.5
FSI-01	PEA-01	-12.1S and -50.8W	CU	21	1	1631	27.3	133.7
FSI-02	PEA-02	-12.3S and -50.7W	CU	378	1	1637	27.2	154.7
FSI-03	PEA-03	-12.4S and -50.9W	CU	164	1	1621	27.1	131.4
FSI-04	PEA-04	-12.4S and -50.7W	CU	605	1	1637	27.1	210.4
FSI-05	PEA-07	-12.5S and -50.9W	CU	5	1	1621	27.1	226.8
FSI-06	PEA-08	-12.5S and -50.7W	CU	8	1	1632	27	222.5
FTP-01	GAU-04	-13.1S and -53.3W	PP	234	150	1795	24.7	145.8
FTP-02	GAU-05	-13.0S and -52.9W	PP	29,560	2,720	1757	24.9	250.2
FTP-03	GAU-06	-13.3S and -53.4W	PP	85	80	1729	24.7	176.9

Forest fragments

The largest and best preserved regional fragments of mature forests were

selected for the study, using Google Earth imagery in order to capture regional variation

in floristics and physiognomy, and with at least three plots for each forest type. All

forest fragments are surrounded by extensive cattle-ranching or soybean fields. The

fragments surveyed varied in size from 5 to 45,459 ha (Table 1).

169

170   Forest structure

171   In each fragment we established an inventory plot of 1 ha, which was

172   subdivided into 25 contiguous subplots of 20 m x 20 m. The forest plots were

173   established between 2008 and 2016 within the private properties and in conservation

174   units; locations varied between 1 and 5440 m from the nearest edge of the fragment. Six

175   plots were seasonally flooded (Table 1) and occasionally affected by fire; the others

176   have no recent record of fire and were either on anthropogenic black earth (terra preta

177   de índio), open rain forests, seasonal evergreen forests, or seasonal semi-deciduous

178   forests (Table 1). For this study, we used the latest available censuses between 2013 and

179   2016.

180   We identified and tagged all the woody individuals with a diameter at breast

181   height (1.3 m) of $\geq$ 10 cm, for a total of 14,185 (range = 338-1599; standard deviation =

182   31) trees and at least 410 (range = 9-135; standard deviation = 256) taxa identified to

183   species level. We identified species in the field or by comparison of collections with

184   herbarium (NX, UFMT, UB and IAN) material of known identity, and with the help of

185   specialists. After identification, the material was incorporated into Herbarium NX, Nova

186   Xavantina, Mato Grosso (Coleção Zoobotânica James Alexander Ratter). We

187   determined the classification of families based on APG III (Angiosperm Phylogeny

188   Group 2009) and reviewed and updated the nomenclature of the taxa using the Lista de

189   Espécies da Flora do Brasil (http://floradobrasil.jbrj.gov.br/2015).

190   We measured the diameter of each tree following standard protocols of the

191   RAINFOR network (http://www.rainfor.org/). We measured the total height using a

192   Leica DISTO laser measurement device. Data were deposited in the ForestPlots.net

193   forest monitoring database (Lopez-Gonzalez et al. 2011).

194

195   Habitat fragmentation

196   To evaluate the effect of habitat fragmentation on forest structure, we

197   measured distance from each plot to the forest edge, the size of each fragment and the

198   forest cover in surrounding landscapes. Whenever possible we measured the distance to

199   the nearest edge in the field. When this was not possible, we estimated this value using

200   Google Earth, which provided a spatial resolution of approximately 20 to 30 m

201   depending on available imagery, and based on our own detailed knowledge, having

202   explored the local context of each plot on foot. In our definition of forest habitat edge,

203   we included all other vegetation and land-use such as plantations, pastures, and roads at

204   least 25 m wide, as well as natural grasslands in the six floodplain forests.

205   We calculated the area of the fragment where each plot was located using

206   Google Earth and ZONUM software (http://zonums.com/online/kmlArea/). These edge

207   and fragment data were collected at the closest possible date to the field sampling and in

208   no case were they collected more than 2 years after the last forest census.

209   We calculated the percentage of forest cover surrounding each plot using

210   buffers of radius size of 1000 m (314 ha), following recommendations of Rocha-Santos

211   et al. (2016). For this we used the land-based metrics in the Fragstats software, that

212   computes descriptors of forest patch and landscape attributes (McGarigal and Cushman

213   2002).

214

215   Climate variables

216   To evaluate the climate effect on the forest structure, we obtained data on 19

217 bioclimatic variables (Table S1) from the WorldClim 1.4 database, with a horizontal

218 resolution of ca. 1 km (Hijmans et al. 2005). We also used data from the Tropical

219 Rainfall Monitoring Mission (TRMM) (NASA 2012) to derive the mean of the annual

220 maximum climatological water deficit (MCWD) (Aragão et al. 2007) between January

221 1999 and December 2011, including the droughts of 2005, 2007 and 2010 (Figure S2).

222 To estimate this, we first calculated MCWD for each year, and then took the mean of all

223 years. MCWD was defined as the most negative value of climatological water deficit

224 (precipitation lower than evapotranspiration) among all the months in each year.

225

226 Data analysis

227   In each plot, we calculated the minimum, maximum, median, and 95

228 percentile of tree diameter (D), height (H) and their allometric (H:D) relationship. We

229 also calculated the weighted Lorey's height values, based on basal area per subplot,

230 using the equation

231          $\sum AB_i * H_i / \sum AB_i$,

232 where AB_i is the basal area of an individual and H_i is its height (e.g. Saatchi et al. 2011).

233 To evaluate the H:D relationship, independently of disturbance, such as the damage

234 caused by recently-opened clearings, we excluded from the analyses all trees with

235 broken stems or those with more than 50% of the crown broken off.

236   We also calculated the mean, median, and total biomass of trees per plot.

237 We estimated the biomass (B) based on the Pantropical model revised by Chave et al.

238 (2014), which is derived from the equation in Chave et al. (2005), that is,

239          $B = 0.0673 \times (\rho D^2 H)^{0.976}$,

240 where D is the diameter in cm, H is the total height of the tree in m, and ρ is

241 the density of the wood. We obtained wood density values from the ForestPlots

242 database (https://www.ForestPlots.net/). We chose this equation to calculate the

243 biomass because it is the most robust approach, given that it takes into consideration the

244 diameter and height of each tree.

245 We developed a correlation matrix of the Kendall's tau values of the

246 environmental and forest structure variables mentioned above (Table S3). Multiple

247 variables share similar source data, leading to high correlation amongst them, so we

248 excluded those with greatest correlations ($r > 0.7$) to avoid repetition of largely

249 redundant forest structure and climate variables (Tables S3 and S4). For all variables,

250 the maximum values and the 95 percentiles were highly correlated; we included only

251 the 95 percentile in order to avoid the influence of outliers. Finally, we excluded

252 predictor variables that correlated poorly ($r < 0.1$) with the vegetation descriptors

253 (Tables S3 and S4).

254 To verify possible differences among all forest plots in the structural

255 variables (95 percentiles of the D, H and H:D, and mean B), we applied the Kruskal-

256 Wallis analysis of variance with the Dunnett post hoc test and a Bonferroni correction

257 (Zar 2010).

258 We evaluated the influence of habitat fragmentation and climatic variables

259 on forest structure using simple correlation and Generalised Linear Models (GLM). We

260 also included in the models the forest type for each forest plot. Simple correlation

261 showed that, six seasonally flooded plots and two plots on anthropogenic black earth

262 were unduly influential, with extreme structure and covarying extreme climatic and

263 fragmentation conditions. To avoid these outliers driving the regional results we

264 excluded them from the GLM and correlation analyses described above.

265   To build the GLM, we first standardised the data and removed the

266 collinearities on the basis of Variance Inflation Factors (VIFs) of less than 10 (Quinn

267 and Keough 2002). We conducted model selection using the Akaike's Information

268 Criterion (AIC), with a model considered to be the best if it had the lowest AIC value

269 (Barton 2016). To access the spatial autocorrelation in the residuals for each model we

270 used Moran's I. Here, no spatial dependence was detected among plots, indicating that

271 the data were not spatially structured (Figure S5). Thus, we considered the plots as

272 independent samples in our subsequent analyses.

273   We conducted the analyses using SAM 4.0 program (Rangel et al. 2010)

274 and R 2.15.1 (R Core Team 2012). The applied R packages were vegan (Oksanen et al.

275 2016), spdep (Bivand et al. 2013), spacemakeR (Dray 2013), MuMIn (Barton 2016) and

276 packfor (Dray et al. 2016). We adopted a 5% significance level for all analyses and used

277 999 randomisations for the permutation methods.

278

279 **Results**

280 Forest structure

281   In general, the three open rainforest plots (FOA-01-03), a forest plot on

282 anthropogenic black earth (FTP-01), were significantly taller than the six seasonally

283 flooded forest plots (FSI-01-06), three seasonal semi-deciduous forest (FES-01-02-05)

284 (Figure 2 and Table S6) and like the other 11 forest plots (FEP-01-07; FES-03-04 and

285 FTP-02-03). The H:D ratio varied in a similar fashion to tree height, with the lowest

286 ratios (i.e., the lowest heights for a given diameter) being recorded in two of the

287 seasonally flooded forest plots (FSI-05 and FSI-06). Tree diameter and biomass did not

288 vary systematically among the plots, except for FSI-03, which had lower diameter and

289 biomass than the most of others plots (Figure 2).
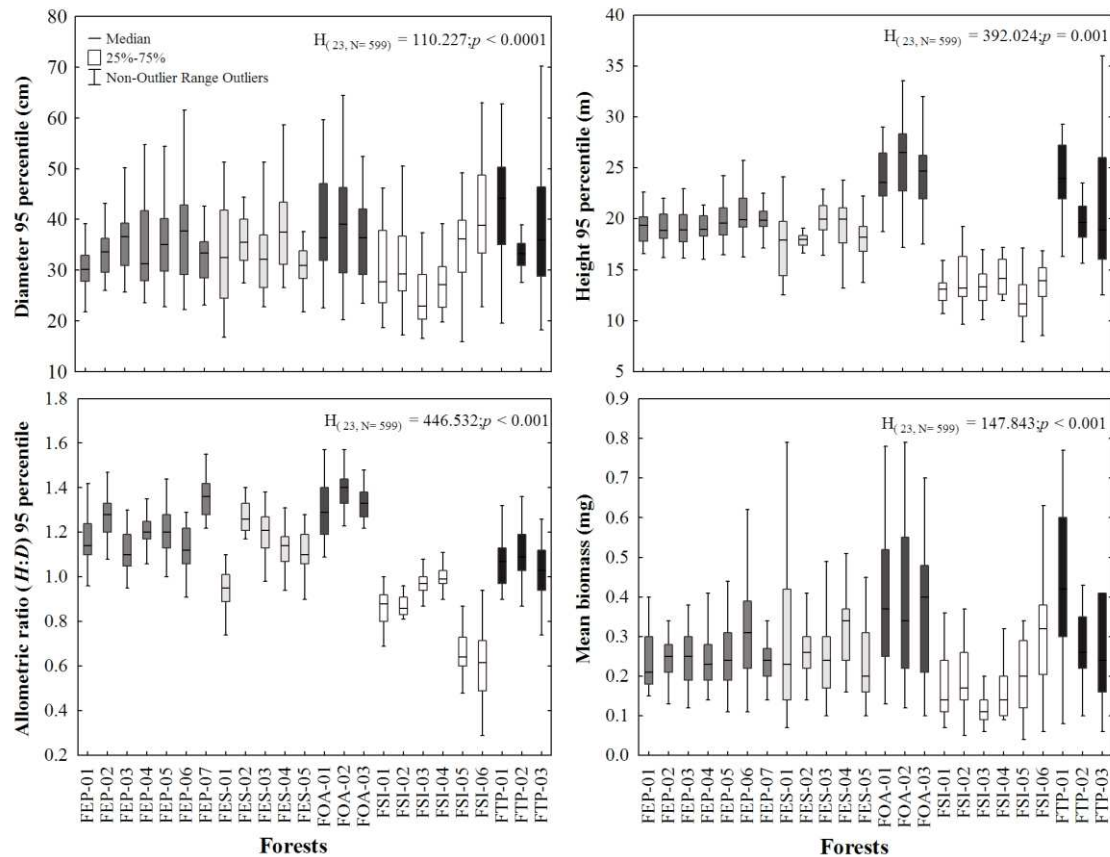
290

291 Figure 2. Variation in the vertical structure of forests at the southern Amazon border.

292 Box-plots show subplot-level values in each location, statistical comparisons are made

293 for among-forest analyses based on the non-parametric Kruskal-Wallis test (H). The

294 complementary pair-wise analysis of forest structure is provided in Table S7. ■ = FTP

295 (seasonal evergreen forest on anthropogenic black earth), ■ = FOA (open rainforest),

296 ■ = FEP (seasonal evergreen forest), □ = FES (seasonal semi-deciduous forest), □ =

297 FSI (seasonally flooded forest).

298

299 Relationship between forest structure, fragmentation and climate variables

300   The structural variables were associated with the precipitation and with

301   fragment area and distance from the edge (Figure 3 and Table 2). Tree height, allometry

302   (H:D) and biomass all correlated positively with precipitation and fragment area (Figure

303   3). Tree height also correlated with the MCWD (Figure 3). Tree diameter did not

304   correlate with any of the variables. Additionally, the precipitation and MCWD

305   correlated positively with the fragment area ($P < 0.05$; Kendal's $\tau = 0.44$ and 0.60,

306   respectively).

307

308   Table 2. The relationship between environmental variables and forest structure, using

309   generalised linear models, of the southern Amazonia forests, Brazil. DE, distance to the

310   edge; PrecWM, precipitation of wettest month; H:D, allometric H:D ratio; FES,

311   seasonal semi-deciduous forest-plots; FOA, open rainforest-plots. Significant effects (P

312   ≤ 0.05) are shown in bold type.

Factors	Estimate	Standard	t	P
Height 95 percentile				
Intercept	-0.276	0.109	-2.531	**0.003**
FES	-0.008	0.161	-0.050	0.961
FOA	1.392	0.328	4.249	**0.001**
PrecWM	0.431	0.140	3.082	**0.010**
Diameter 95 percentile				
Intercept	-0.356	0.290	-1.228	0.243
FES	0.039	0.445	0.089	0.931
FOA	1.715	0.530	3.237	**0.007**
H:D 95 percentile				
Intercept	<0.001	0.174	<0.001	1.000
DE	-0.785	0.302	-2.597	**0.023**
PrecWM	1.260	0.302	4.167	**0.001**
Mean biomass				
Intercept	-0.540	0.166	-3.249	**0.007**
FES	0.244	0.257	0.949	0.361
FOA	2.291	0.303	7.555	**<0.001**

313
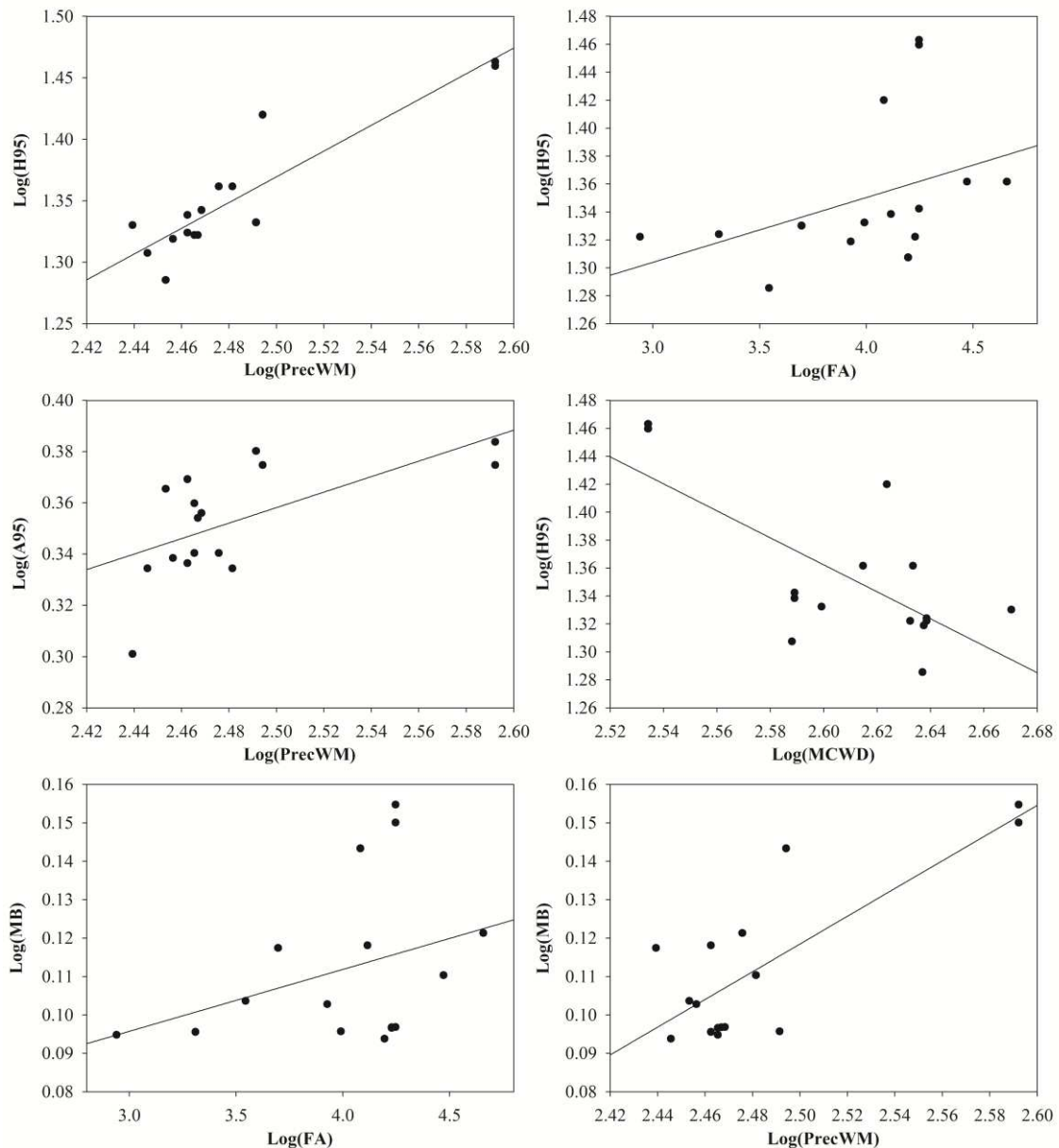
314

315 Figure 3. Significant (P ≤ 0.05) relationships between forest structure and climatic and

316 fragmentation variables of the southern Amazon border forest plots. H95 = height 95

317 percentile, A95 = allometric ratio (H:D) 95 percentile, MB = mean biomass (Mg), FA =

318 fragment area (ha), PrecWM = precipitation of wettest month (mm), MCWD =

319 maximum climatological water deficit (mm).

320

321   Based on the best GLM models for each forest structure variable, forest type

322   and precipitation were most related to tree height (Table 2). Forest type was also a

323   strongly related to tree diameter and biomass. Annual mean precipitation and distance

324   from the edge were important factors for mean plot H:D (Table 2). The percentage of

325   forest cover around each plot was not selected in the best models and was not correlated

326   with any forest structure variables. All plots presented more than 50% forest cover in

327   surrounding landscapes.

328   Precipitation and MCWD were not selected in the same model, indicating

329   that each had similar (but inverse) effects on forest structure. Thus, all structural

330   parameters affected positively by precipitation (Table 2) are affected negatively by

331   moisture stress (MCWD) (Table S7).

332

333   **Discussion**

334   Our results show that the forests of the southern border zone of Amazonia

335   vary remarkably in their structure, principally in terms of their tree height and tree

336   height:diameter ratio. Most of the structural variation in these forests was statistically

337   related to fragment area and precipitation, supporting our overall expectations and

338   largely consistent with our hypotheses. Here we briefly first discuss this overall

339   variability and its potential ultimate drivers, before proceeding to discuss the results in

340   more detail.

341

342   Structural variability of the forests of the southern Amazon border zone

343   Our general expectation was that climatic variation in the region would be a

344   fundamental determinant of the variability in forest structure here, principally because

345 drought events and seasonality may be more intense at the southern border in relation to

346 the core area of the Amazonas basin with evergreen non-seasonal rain forests (Lewis et

347 al. 2011). In particular, water deficit may kill large trees (McIntyre et al. 2015), taller

348 trees tend to be most affected by these conditions (Rowland et al. 2015). As these trees

349 die and break-up or fall, large clearings are opened, favouring the establishment of

350 species of different ecological groups (Lawton and Putz 1988). The frequent formation

351 of clearings in these hyperdynamic transitional forests, as documented by Marimon et

352 al. (2014), may thus also contribute to the structural variability found here. Finally, the

353 forests of the southern border of the Amazon are located within a mosaic of vegetation

354 types with many species typical of the adjacent biomes (Ratter et al. 1973), which may

355 have a direct influence on the structural diversity of these forests.

356

357 Seasonally flooded forest plots

358   The lowest height and H:D allometric ratio in the seasonally flooded forest

359 plots may be explained by their smaller fragment size and proximity to edges. These

360 factors as well as higher temperatures and lower precipitation (Table 1) may intensify

361 the fire effects. Fires in the wider grassland matrix can penetrate into forest fragments

362 and increase tree mortality, as observed in a recent study in these forest plots

363 (Maracahipes et al. 2014). It therefore appears likely that the combined effects of

364 reduced fragment area and precipitation and higher temperatures, together with fire and

365 its potential interactions with droughts (Brando et al. 2014), contribute to forest

366 structure here.

367

368 Response of the forest structure to the fragmentation and climate variables

369   Temperature appears to be an important factor determining the height of the

370   trees worldwide, including potentially in tropical forests (Koch et al. 2004; Way and

371   Oren 2010; Feldpausch et al. 2011; Lines et al. 2012; Pan et al. 2013), but here the

372   absence of a clear statistical relationship between structure and temperature (P > 0.05,

373   Kendal's τ = 0.31) suggests it is not critical at the southern Amazon transition zone.

374   Rather, in our study the greater forest heights, H:D ratio and biomass that were

375   observed with increasing precipitation suggest water supply is the dominant climate

376   control on forest structure, and is consistent with some work elsewhere in the tropics

377   (e.g. Alvarez et al. 2017), given especially that tropical plants tend to grow faster and

378   taller as water is more available (Vlam et al. 2014; Givnish et al. 2014). In addition to

379   apparent effects of annual rainfall, we also found that climatological water deficit was

380   associated with reduced investment by the trees in height growth, consistent with the

381   hypothesis that tree height is constrained by the availability of water (Ryan et al. 2004;

382   Givnish et al. 2014). A significant positive correlation was also found between

383   precipitation and tree height along a precipitation gradient in Australia, which Givnish

384   et al. (2014) related to the increase in leaf area and rates of photosynthesis with

385   increasing precipitation.

386   The negative correlation between the cumulative water deficit and tree

387   height may be related to the mortality of the largest individuals during extreme drought

388   events (Phillips et al. 2010). Such droughts have been directly observed in the study

389   region in 2005, 2007, and 2010 (e.g. Brando et al. 2014), and these have indeed tended

390   to kill larger trees (Phillips et al. 2010; Feldpausch et al. 2016), as is often the case with

391   droughts in other tropical forests (Bennett et al. 2015). In Amazonia, recent strong

392   droughts appear also to be a major cause of the recent basin-wide increase in tree

393   mortality rates (Phillips et al. 2009; Brienen et al. 2015). In the near future, more

394   frequent extreme droughts, especially if combined with warming of the Amazon region

395   and thermal peaks in El Niño events such as in 2015-16 (Jiménez-Muñoz et al. 2016),

396   may therefore have profound implications for the forest structure of the southern

397   Amazon border, located as they are in a region that is already naturally close to their

398   distributional and hydraulic limits. In this scenario, large trees are more susceptible to

399   damage to the xylem, which can ultimately result in the death of the plant (e.g. McIntyre

400   et al. 2015) and eventually lead to forests of lower stature (McDowell et al. 2008;

401   Rowland et al. 2015). Trees being smaller in drier areas with greater water deficiency is

402   directly be related to conservative modifications in the hydraulic structure of the plants

403   under hydrological stress to avoid embolism (e.g. Lines et al. 2012, Claeys and Inzé

404   2013). Thus, as have recently argued in both tropical and temperate zone contexts (e.g.

405   Stegen et al. 2011; Banin et al. 2012; McIntyre et al. 2015) it is likely that trees in

406   forests subject either to more extreme climatic events, or to more disturbance (including

407   seasonally flooded habitats), or both, will in general tend to be shorter at a given

408   diameter in order to avoid risks of hydraulic and/or mechanical failure, whereas trees in

409   forests with high rainfall, such as our FOA-01 and FOA-02, will have greater heights

410   and hence greater biomass.

411   Besides the correlation with the climatic variables, both height and the

412   biomass of trees were positively correlated with fragment area. This result may be

413   related to the incidence of wind in smaller fragments which have a higher proportion of

414   forest edge (D'Angelo et al. 2004; Laurance 2004; Haddad et al. 2015). These

415   disturbances are known to be able to generate high mortality, especially of the tallest

416   trees (Laurance et al. 2000; Laurance 2004), and consequently in our dataset such edge-

417   generated disturbances may have affected the height and biomass of trees. Elsewhere,

418   local climatic changes as a result of fragmentation can reduce the density and diversity

419   of species (Mantyka-Pringle et al. 2012). Such effects can also increase the

420   susceptibility of fragmented forest structure and their biota to fire (Laurance and

421   Williamson 2001; Laurance 2004). In the southern Amazon region, these different

422   effects are all likely to be relevant, but clearly further analysis is needed, including long-

423   term monitoring evaluation of the climatic and dynamic processes in these forests.

424

425   **Conclusions**

426   Our analysis across different locations, spanning a large part of the southern

427   Amazon zone, suggests climate sensitivity in forest structure here. Climate change, and

428   especially any reduction in annual or seasonal precipitation, is thus likely to have a

429   significant effect on the forest structure in the southern border of the Amazon.

430   Secondly, our results also suggest that the effects of reduction in the annual

431   precipitation may be further exacerbated in smaller fragments. This suggests that habitat

432   fragmentation may intensify the negative effects of climate change and burning in

433   forests in the southern Amazon border, resulting in a substantial risk of increases in tree

434   mortality. Given the likely susceptibility of the remaining southern Amazon border

435   forests to environmental change, strong conservation strategies are urgently needed to

436   guarantee the persistence of these habitats, especially for the smaller fragments and

437   those close to agricultural frontiers.

438

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455

456   **Notes on contributors**

457   Simone Matias Reis is a Ph.D. student focussing on the effects of biotic and abiotic

458   factors on forest structure in the southern border of Amazonia.

459   Beatriz Schwantes Marimon is professor and an ecologist with experience in working

460   with structure and dynamics of different vegetation types of the southern Amazon

461   boundaries in Central Brazil. Her current work focuses on ecology and management of

462   forests in the transition zone between Cerrado and Amazon forest biomes, and currently

463   has great interest in understanding changes in tropical plants communities as a result of

464   climate change.

465   Ben Hur Marimon-Junior is professor. He has experience in forest ecology, studying

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467   biodiversity and ecosystem functions. His current work focuses on understanding

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469   Paulo S. Morandi is a Ph.D. student focussing on structural patterns, diversity and

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476   and landscape on the resilience of secondary forests in the Amazon.

477   Eder Carvalho das Neves is M.Sc. student focussing on the physiology of forests in the

478   Amazon-Cerrado transition.

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490

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Supplementary material

Table S1. Environmental predictors and vegetation descriptors used in the analyses.

Variable abbreviation	Environmental predictors	Variable abbreviation	Vegetation descriptors
FA	Fragment area (ha)	MIH	Minimum height (m)
DE	Distance to the forest edge (m)	MAH	Maximum height (m)
FC	Forest cover (%)	MH	Median height (m)
MCWD	Maximum climatological water deficit (mm)	H95	Height 95 percentile (m)
Temp	Mean annual temperature (°C)	LH	Weighted Lorey's height
TempMDR	Mean diurnal range (°C)	MD	Median diameter (cm)
Isoter	Isothermality (°C)	MAD	Maximum diameter (cm)
TempSaz	Temperature seasonality (standard deviation *100) (°C)	D95	Diameter 95 percentile (cm)
TempWM	Max temperature of warmest month (°C)	MIA	Minimum allometric ratio (H:D)
TempCM	Min temperature of coldest month (°C)	MAA	Maximum allometric ratio (H:D)
TempAR	Temperature annual range (°C) TempWM - TempCM	MA	Median allometric ratio (H:D)
TempWeQ	Mean temperature of wettest quarter (°C)	A95	Allometric ratio (H:D) 95 percentile
TempDQ	Mean temperature of driest quarter (°C)	MB	Mean biomass (Mg ha)
TempWaQ	Mean temperature of warmest quarter (°C)	MEB	Median biomass (Mg ha)
TempCQ	Mean temperature of coldest quarter (°C)	TB	Total biomass (Mg ha)
Prec	Total annual precipitation (mm)		
PrecWM	Precipitation of wettest month (mm)	-	-
PrecDM	Precipitation of driest month (mm)	-	-
PrecSaz	Precipitation seasonality (Coefficient of Variation) (mm)	-	-
PrecWeQ	Precipitation of wettest quarter (mm)	-	-
PrecDQ	Precipitation of driest quarter (mm)	-	-
PrecWaQ	Precipitation of warmest quarter (mm)	-	-
PrecCQ	Precipitation of coldest quarter (mm)	-	-

Figure S2. Mean of the maximum climatological water deficit (MCWD) (mm) in the Amazon basin between 1999 and 2011, in the context of the rest of Amazonia. Circles show the forest plots localization.

Table S3. Kendall tau correlations of the all 37 environmental and forest structure variables obtained to the forests of the southern Amazon border. FA = fragment area (ha), DE = distance to the edge (m), MCWD= maximum climatological water deficit (mm), Temp = mean annual temperature (℃), TempMDR = Mean diurnal range (℃), Isoter = Isothermality (℃), TempSaz = Temperature seasonality (standard deviation *100) (℃), TempWM = Max temperature of warmest month (℃), TempCM = Min temperature of coldest month (℃), TempAR = Temperature annual range (℃) TempWM – TempCM, TempWeQ = Mean temperature of wettest quarter (℃), TempDQ = Mean temperature of driest quarter (℃), TempWaQ = Mean temperature of warmest quarter (℃), TempCQ = Mean temperature of coldest quarter (℃), Prec = Total annual precipitation (mm), PrecWM = Precipitation of wettest month (mm), PrecDM = Precipitation of driest month (mm), PrecSaz = Precipitation seasonality (Coefficient of Variation) (mm), PrecWeQ = Precipitation of wettest quarter (mm), PrecDQ = Precipitation of driest quarter (mm), PrecWaQ = Precipitation of warmest quarter (mm), PrecCQ = Precipitation of coldest quarter (mm), MIH = Minimum height (m), MAH = Maximum height (m), MH = Median height (m), H95 = Height 95 percentile (m), LH = Weighted Lorey's height, MD = Median diameter (cm), MAD = Maximum diameter (cm), D95 = Diameter 95 percentile (cm), MIA = Minimum allometric ratio (H:D), MAA = Maximum allometric ratio (H:D), MA = Median allometric ratio (H:D), A95 = Allometric ratio (H:D) 95 percentile, MB = Mean biomass (Mg ha), MEB = Median biomass (Mg ha), TB = Total biomass (Mg ha). Significant correlations (p ≤ 0.05) are shown in bold type.

	FA	DE	MCWD	Temp	TempMDR	Isoter	TempSaz	TempWM	TempCM	TempAR	TempWeQ	TempDQ	TempWaQ	TempCQ	Prec	PrecWM	PrecDM
FA		**0.55**	**0.61**	-0.18	0.28	-0.17	-0.19	0.17	-0.16	**0.29**	-0.26	0.02	-0.20	-0.01	**0.37**	**0.48**	0.00
DE			**0.51**	-0.21	**0.30**	-0.25	-0.13	0.12	-0.20	**0.32**	**-0.33**	-0.09	-0.23	-0.10	0.23	**0.44**	0.00
MCWD				-0.13	0.24	-0.12	**-0.35**	**0.34**	-0.05	0.24	-0.27	0.13	-0.16	0.08	**0.62**	**0.56**	0.05
Temp					**-0.82**	0.51	**-0.42**	0.32	**0.92**	**-0.80**	**0.85**	**0.75**	**0.96**	**0.78**	-0.16	**-0.37**	0.17
TempMDR						**-0.64**	0.29	-0.17	**-0.79**	**0.96**	**-0.89**	**-0.59**	**-0.85**	**-0.62**	0.29	**0.50**	-0.12
Isoter							**-0.41**	0.20	**0.56**	**-0.69**	**0.66**	**0.62**	**0.54**	**0.63**	-0.18	**-0.46**	0.15
TempSaz								**-0.72**	**-0.51**	0.29	-0.27	**-0.68**	**-0.39**	**-0.67**	**-0.30**	-0.09	-0.24
TempWM									**0.39**	-0.15	0.18	**0.57**	0.30	**0.55**	**0.43**	0.22	**0.34**
TempCM										**-0.80**	**0.78**	**0.82**	**0.89**	**0.85**	-0.10	**-0.35**	0.22
TempAR											**-0.88**	**-0.59**	**-0.82**	**-0.63**	**0.30**	**0.54**	-0.13
TempWeQ												**0.60**	**0.88**	**0.63**	**-0.30**	**-0.51**	0.09
TempDQ													**0.71**	**0.97**	0.09	-0.17	**0.31**
TempWaQ														**0.74**	-0.19	**-0.40**	0.14
TempCQ															0.05	-0.21	**0.30**
Prec																**0.56**	0.25
PrecWM																	-0.12
PrecDM																	
PrecSaz																	
PrecWeQ																	
PrecDQ																	
PrecWaQ																	
PrecCQ																	
MAH																	
MIH																	
MH																	
H95																	
LH																	
MAD																	
MD																	
D95																	
MAA																	
MIA																	
MA																	
A95																	
MB																	
MEB																	
TB																	

Continuation…

	PrecSaz	PrecWeQ	PrecDQ	PrecWaQ	PrecCQ	MAH	MIH	MH	H95	LH	MAD	MD	D95	MAA	MIA	MA	A95	MB	MEB	TB
FA	**-0.32**	**0.36**	0.16	-0.07	**0.31**	**0.35**	0.23	**0.55**	**0.51**	**0.44**	**0.30**	0.20	0.05	**0.32**	**0.42**	**0.47**	**0.51**	**0.31**	**0.43**	0.20
DE	**-0.30**	**0.29**	0.23	**-0.29**	**0.33**	**0.49**	**0.34**	**0.50**	**0.49**	**0.38**	**0.38**	0.18	0.14	**0.30**	**0.33**	**0.49**	**0.50**	**0.31**	**0.35**	**0.29**
MCWD	**-0.64**	**0.50**	**0.38**	-0.04	**0.48**	**0.39**	**0.42**	**0.57**	**0.46**	**0.51**	**0.46**	0.18	0.12	**0.33**	0.23	**0.56**	**0.56**	**0.32**	**0.44**	0.28
Temp	0.19	**-0.49**	-0.05	0.15	-0.19	**-0.35**	-0.21	**-0.31**	**-0.36**	**-0.35**	-0.23	0.04	**-0.33**	-0.11	**-0.46**	-0.27	**-0.29**	**-0.45**	-0.10	-0.22
TempMDR	**-0.36**	**0.60**	0.16	-0.17	**0.30**	**0.48**	0.22	**0.43**	**0.49**	**0.46**	**0.29**	-0.04	**0.42**	0.17	**0.54**	**0.42**	**0.42**	**0.55**	0.19	**0.37**
Isoter	0.23	**-0.56**	-0.07	**0.37**	-0.27	**-0.43**	-0.08	**-0.34**	**-0.46**	**-0.36**	-0.16	0.03	**-0.39**	-0.21	**-0.46**	**-0.30**	**-0.33**	**-0.45**	-0.12	**-0.33**
TempSaz	0.28	0.03	-0.16	-0.02	-0.18	-0.02	-0.25	-0.19	-0.04	-0.11	-0.17	-0.08	0.16	-0.19	0.14	-0.25	-0.20	0.13	-0.20	-0.13
TempWM	**-0.33**	0.10	0.28	0.03	0.27	0.06	0.18	0.17	0.14	0.14	0.11	0.05	-0.05	0.19	-0.06	0.20	0.16	0.00	0.25	0.22
TempCM	0.15	**-0.43**	0.00	0.11	-0.13	**-0.34**	-0.14	-0.27	**-0.34**	**-0.32**	-0.19	0.04	**-0.36**	-0.06	**-0.43**	-0.23	-0.26	**-0.44**	-0.07	-0.21
TempAR	**-0.35**	**0.63**	0.13	-0.19	**0.31**	**0.50**	0.24	**0.42**	**0.53**	**0.48**	**0.31**	-0.03	**0.45**	0.18	**0.53**	**0.41**	**0.43**	**0.59**	0.22	**0.39**
TempWeQ	**0.36**	**-0.62**	-0.16	**0.29**	**-0.34**	**-0.49**	**-0.29**	**-0.46**	**-0.48**	**-0.47**	**-0.34**	-0.04	**-0.40**	-0.23	**-0.49**	**-0.42**	**-0.43**	**-0.53**	-0.23	**-0.36**
TempDQ	-0.05	**-0.30**	0.10	0.12	0.00	-0.22	-0.01	-0.08	-0.19	-0.14	-0.04	0.09	-0.29	0.01	**-0.35**	-0.04	-0.08	**-0.34**	0.06	-0.12
TempWaQ	0.25	**-0.51**	-0.08	0.19	-0.22	**-0.40**	-0.21	**-0.36**	**-0.38**	**-0.38**	-0.25	0.04	**-0.35**	-0.13	**-0.49**	**-0.31**	**-0.33**	**-0.46**	-0.11	-0.26
TempCQ	-0.02	**-0.34**	0.07	0.15	-0.05	-0.25	-0.02	-0.10	-0.23	-0.18	-0.07	0.06	**-0.33**	-0.01	**-0.37**	-0.05	-0.09	**-0.39**	0.01	-0.13
Prec	**-0.63**	**0.59**	**0.44**	0.11	**0.55**	**0.36**	0.24	**0.44**	**0.50**	**0.50**	0.26	0.11	0.19	**0.30**	**0.31**	**0.39**	**0.40**	**0.39**	**0.36**	**0.39**
PrecWM	**-0.45**	**0.81**	0.20	-0.06	**0.40**	**0.52**	**0.36**	**0.51**	**0.78**	**0.75**	**0.35**	0.03	**0.29**	**0.44**	**0.34**	**0.52**	**0.55**	**0.53**	**0.35**	**0.57**
PrecDM	**-0.29**	-0.12	**0.64**	0.07	**0.44**	0.06	-0.06	-0.03	0.06	-0.01	0.02	0.18	0.05	-0.08	0.02	-0.15	-0.12	0.09	0.08	0.02
PrecSaz		**-0.50**	**-0.62**	0.12	**-0.78**	**-0.38**	**-0.36**	**-0.52**	**-0.43**	**-0.46**	**-0.37**	-0.11	-0.12	-0.26	-0.19	**-0.51**	**-0.47**	-0.27	-0.28	**-0.38**
PrecWeQ			0.22	-0.09	**0.41**	**0.42**	0.28	**0.54**	**0.65**	**0.61**	0.26	0.04	**0.29**	**0.37**	**0.42**	**0.52**	**0.52**	**0.50**	**0.29**	**0.51**
PrecDQ				0.11	**0.75**	0.25	0.11	0.27	**0.32**	0.28	0.16	0.15	0.05	0.13	0.12	0.20	0.20	0.21	0.21	0.21
PrecWaQ					-0.13	-0.20	-0.32	-0.23	-0.06	-0.07	**-0.30**	-0.09	-0.08	-0.16	-0.06	-0.25	-0.20	-0.02	-0.12	-0.04
PrecCQ						**0.40**	**0.33**	**0.42**	**0.49**	**0.41**	**0.30**	0.22	0.15	0.25	0.18	**0.35**	**0.35**	**0.34**	**0.34**	**0.29**
MAH							0.25	**0.42**	**0.68**	**0.68**	**0.55**	-0.03	**0.44**	0.24	**0.33**	**0.38**	**0.40**	**0.54**	0.14	**0.48**
MIH								**0.44**	**0.32**	**0.42**	**0.52**	0.08	0.18	**0.34**	-0.06	**0.47**	**0.51**	0.26	0.29	0.18
MH									**0.46**	**0.56**	0.31	0.16	0.15	**0.56**	**0.38**	**0.80**	**0.89**	0.27	**0.47**	**0.38**
H95										**0.78**	**0.42**	0.07	**0.39**	**0.29**	**0.35**	**0.40**	**0.45**	**0.66**	**0.31**	**0.55**
LH											**0.52**	-0.01	**0.41**	**0.45**	0.28	**0.51**	**0.58**	**0.57**	0.27	**0.60**
MAD												0.01	**0.31**	0.19	0.00	**0.35**	**0.35**	**0.39**	0.16	0.25
MD													0.27	0.06	0.09	-0.04	0.06	0.19	**0.58**	-0.22
D95														-0.01	0.15	0.03	0.14	**0.70**	0.28	0.16
MAA															0.18	**0.55**	**0.61**	0.06	0.28	0.28
MIA																**0.30**	**0.31**	**0.32**	0.24	**0.31**
MA																	**0.87**	0.12	**0.30**	**0.32**
A95																		0.24	**0.39**	**0.37**
MB																			**0.35**	**0.39**
MEB																				0.07
TB																				

Table S4. Pre-selected environmental and forest structure variables used in the analyses of the forest-plots of the southern Amazon border. FA = fragment area (ha), DE = distance to the edge (m), FC = forest cover (%), Temp = mean annual temperature (°C), PrecWM = precipitation of wettest month (mm), MCWD= maximum climatological water deficit (mm), MH= median height and H95 = 95 percentile, MD = median diameter and D95 = 95 percentile, MA = median allometric ratio (H:D) and A95 = 95 percentile, MB = mean biomass (Mg), and TB = total biomass.

| Forest plots	Environmental predictors						Vegetation descriptors			
	FA	DE	FC	Temp	PrecWM	MCWD	H95	D95	A95	MB
FEP-01	870	1,030	99	25.5	291	-435.02	20.0	33.6	1.19	0.24
FEP-02	2,035	1,000	100	25.6	289	-435.02	20.1	36.6	1.34	0.25
FEP-03	8,432	990	98	24.9	285	-434.01	19.8	40	1.18	0.27
FEP-04	16,901	520	74	25.1	292	-428.93	20.0	37.8	1.26	0.25
FEP-05	16,901	329	100	25.0	291	-428.93	20.0	37.8	1.29	0.25
FEP-06	45,459	3,600	100	26.9	298	-411.82	22.0	41.4	1.19	0.32
FEP-07	9,789	1,180	100	26.1	309	-397.35	20.5	35.4	1.40	0.25
FES-01	4,968	1,350	78	25.2	274	-468.04	20.4	40.4	1.00	0.31
FES-02	3,499	160	69	24.1	283	-433.5	18.3	39.4	1.32	0.27
FES-03	17,624	90	58	26.7	293	-388.22	21.0	35.4	1.27	0.25
FES-04	13,039	860	88	26.8	289	-388.22	20.8	39.3	1.17	0.31
FES-05	15,680	2,980	100	26.6	278	-387.33	19.3	33.8	1.16	0.24
FOA-01	12,066	900	98	25.1	311	-420.38	25.3	44.8	1.37	0.39
FOA-02	17,628	5,440	100	25.5	390	-342.12	27.8	42.6	1.42	0.43
FOA-03	17,628	5,410	50	25.6	390	-342.12	28.1	42.3	1.37	0.41
FSI -01	21	1	-	27.3	273	-440.57	13.6	32.3	0.93	0.14
FSI -02	378	1	-	27.2	277	-454.52	15.0	35.2	0.92	0.19
FSI -03	164	1	-	27.1	273	-457.47	14.0	24.4	0.99	0.12
FSI -04	605	1	-	27.1	278	-454.52	15.7	28.1	1.02	0.15
FSI -06	5	1	-	27.1	274	-457.47	13.9	40.3	0.75	0.19
FSI -07	8	1	-	27.0	278	-444.82	15.6	45.0	0.77	0.3
FTP-01	234	150	38	24.7	308	-436.02	26.8	51.9	1.14	0.48
FTP-02	29,560	2,720	71	24.9	302	-429.99	22.0	34.7	1.16	0.29
FTP-03	85	80	30	24.7	294	-433.5	24.0	45.3	1.09	0.52
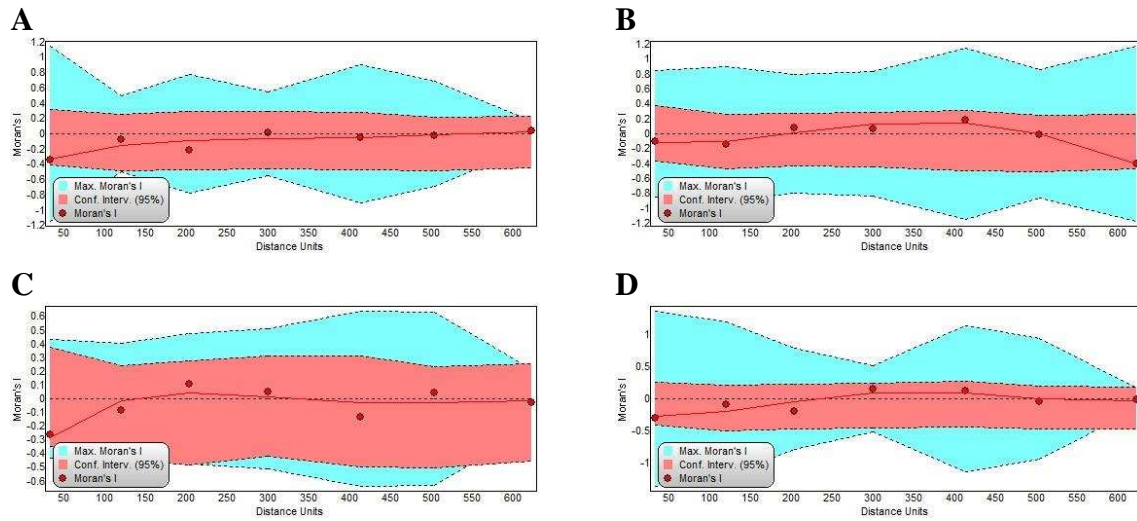
Figure S5. Spatial autocorrelation of the residuals of each model, based in Moran's I index for: A = height, B = diameter, C= allometric ratio (H:D), and D = biomass of the forests plots in the southern Amazon border.

Table S6. Comparison of the forest structure variables of the forests in the southern Amazon border, based on the Kruskal-Wallis nonparametric analysis of variance (H). MH= median height and H95 = 95 percentile, MD = median diameter and D95 = 95 percentile, MA = median allometric ratio (H:D) and A95 = 95 percentile, MB = mean biomass (Mg), and TB = total biomass. Values on different lines within the same column followed by different letters are significantly different based on Dunnett's post hoc test with the Bonferroni correction.

Forests	H95		D95		A95		MB	
FEP-01	19.3	afg	31.5	acd	1.17	aefghi	0.25	adef
FEP-02	19.3	afg	33.1	abcd	1.27	afg	0.25	abdef
FEP-03	19.0	fg	37.5	ab	1.10	deghi	0.27	abdef
FEP-04	19.0	fg	33.9	abcd	1.20	afghi	0.25	abdef
FEP-05	19.7	afg	35.1	abd	1.21	afgh	0.26	abdef
FEP-06	20.4	afg	38.1	ab	1.12	defghi	0.33	abf
FEP-07	19.9	afg	32.8	abcd	1.36	a	0.25	abdef
FES-01	17.6	def	33.8	abcd	0.94	bcd	0.30	abdef
FES-02	18.0	cdef	35.9	ab	1.26	afg	0.28	abef
FES-03	20.1	afg	34.4	abcd	1.19	afghi	0.26	abdef
FES-04	19.6	afg	38.2	ab	1.13	defghi	0.32	abf
FES-05	18.2	ef	31.9	acd	1.12	defghi	0.25	adef
FOA-01	24.0	a	38.8	ab	1.29	afg	0.39	ab
FOA-02	25.7	a	39.4	ab	1.39	a	0.44	ab
FOA-03	24.8	ag	38.3	ab	1.31	af	0.41	ab
FSI-01	13.1	bc	30.5	acd	0.84	bc	0.18	cde
FSI-02	14.2	bcde	31.6	acd	0.85	bc	0.20	cdef
FSI-03	13.1	bcd	24.5	c	0.97	bcd	0.12	c
FSI-04	14.3	bcde	27.0	cd	0.98	bcde	0.16	cd
FSI-05	11.9	b	35.2	ab	0.66	b	0.23	acdef
FSI-06	13.4	bcd	40.5	ab	0.61	b	0.32	abf
FTP-01	23.5	ag	43.2	b	1.06	cdehi	0.47	b
FTP-02	19.7	afg	33.1	abd	1.11	deghi	0.29	abef
FTP-03	21.1	afg	42.8	ab	1.02	bcdei	0.52	abdef

Table S7. Generalized linear models of the factors that influence forest structure of the vegetation in forest plots of the southern Amazon border. Temp = mean annual temperature, MCWD = maximum climatological water deficit, H:D = allometric H:D ratio, FES = seasonal semi-deciduous forest-plots, FOA = open rainforest-plots. Significant effects ($p \leq 0.05$) are shown in bold type.

Factors	Estimate	Standard	t	P
Height 95 percentile				
Intercept	2.462	1.229	2.003	0.070
FES	-0.206	0.177	-1.161	0.270
FOA	1.848	0.262	7.060	**0.000**
MCWD	0.007	0.003	2.340	**0.039**
H:D 95 percentile				
Intercept	8.630	2.679	3.221	**0.007**
MCWD	0.021	0.007	3.230	**0.007**
Temp	-0.497	0.230	-2.159	0.052