

Big trees drive forest structure patterns across a lowland Amazon regrowth gradient

Tassiana Maylla Fontoura Caron¹, Victor Juan Ulises Rodriguez Chuma², Alexander Arévalo Sandi³, Darren Norris^{1, 3, 4 *}

¹ Programa de Pós-Graduação em Biodiversidade Tropical, Universidade Federal do Amapá (UNIFAP), Rod. Juscelino Kubitschek, Km 02, 68902-280, Macapá, AP, Brazil.

² Facultad de Ciencias Forestales, Universidad Nacional de la Amazonia Peruana (UNAP), Pevas 5ta cuadra, Iquitos, Perú.

³ Programa de Pós-graduação em Ecologia, Instituto Nacional de Pesquisas da Amazônia (INPA), Av. André Araújo 2936, Petrópolis 69067-375 Manaus, AM, Brazil.

⁴ Coordenação de Ciências Ambientais, Universidade Federal do Amapá (UNIFAP), Rod. Juscelino Kubitschek Km 02, 68902-280 Macapá, AP, Brazil

** Corresponding author at: Coordenação de Ciências Ambientais, Universidade Federal do Amapá (UNIFAP), Rod. Juscelino Kubitschek Km 02, 68902-280 Macapá, AP, Brazil. Email address: dnorris75@gmail.com (D. Norris).

1 **Abstract**

2 Degraded Amazonian forests can take decades to recover and the ecological results of natural
3 regeneration are still uncertain. Here we use field data collected across 15 lowland Amazon
4 smallholder properties to examine the relationships between forest structure, mammal diversity,
5 regrowth type, regrowth age, topography and hydrography. Forest structure was quantified together
6 with mammal diversity in 30 paired regrowth-control plots. Forest regrowth stage was classified into
7 three groups: late second-regrowth, early second-regrowth and abandoned pasture. Basal area in
8 regrowth plots remained less than half that recorded in control plots even after 20-25 years. Although
9 basal area did increase in sequence from pasture, early to late-regrowth plots, there was a significant
10 decline in basal area of late-regrowth control plots associated with a decline in the proportion of
11 large trees. There was also contrasting support for different non-mutually exclusive hypotheses, with
12 proportion of small trees (DBH <20cm) most strongly supported by topography (altitude and slope)
13 whereas the proportion of large trees (DBH >60cm) supported by plot type and regrowth class. These
14 findings support calls for increased efforts to actively conserve large trees to avoid retrogressive
15 succession around edges of degraded Amazon forests.

16

17 **Introduction**

18 Healthy tropical forests provide goods and services to human populations. Yet tropical forests show
19 worrying rates of forest loss with an elevated loss / gain ratio and a statistically significant trend in
20 annual forest loss of 2101 km²/year¹. One option to revert tropical forest loss is the restoration of
21 degraded forests and deforested landscapes^{2,3}. Although the post-disturbance restoration of forest
22 ecosystems often involves passive restoration strategies (i.e. natural regeneration), the ecological
23 results of this type of restoration are still uncertain²⁻⁴.

24 Continuing widespread forest losses across Amazonia compromises vital ecosystem services
25 such as carbon storage, regulation of hydrological cycles and climate patterns⁵⁻⁷. Riverside forests
26 are particularly threatened and suffer losses due to the conversion of forest cover to pastures,
27 compromising the maintenance of water flows⁸. The recovery of degraded areas is necessary to
28 recuperate the standing forest value and the Amazon offers an excellent recovery opportunity due to
29 its natural potential for regeneration^{9,10}. Yet, the regrowth rate of degraded Amazon forests can be
30 slow, as abandoned areas are typically on compacted poor quality soils^{11,12} and due to the high
31 structural and biological diversity of the original forests¹³.

32 Separating the complex interactions driving recruitment and recovery patterns of highly diverse
33 Amazon forests is challenging^{2,3,14,15}, yet we know that different faunal groups can modulate and
34 generate key impacts¹⁶⁻¹⁹. Indeed, the successional trajectory of natural regeneration in degraded
35 forests can depend strongly on the concomitant recovery of faunal diversity and associated
36 ecosystem services (e.g. seed dispersal)²⁰⁻²². For example, seed predation by both vertebrates and
37 invertebrates^{23,24} can limit germination and subsequent recruitment²². Long-term experiments have
38 demonstrated the impact of vertebrates on recruitment, showing how this group contributes to the
39 maintenance of tropical forest species and structural diversity²⁵⁻²⁸.

40 Amazon mammals are important component of forest diversity^{25,29} including carbon²⁸ and
41 biomass cycles¹⁸. Mammals can also play an important role in the successional trajectory and
42 recovery of degraded areas as dispersers and predators of both seeds and seedlings²³. Mid- and
43 large-bodied mammals (weight > 1 kg) can disperse a large numbers of seeds over long distances
44^{23,30}. For example, lowland tapirs can travel over 4 kilometers in a day³¹ and disperse seeds of more
45 than 70 tree species³². The loss of mid- to large-bodied mammals may release some plant species
46 from herbivory and increase their dominance, which subsequently decreases tropical forest
47 biodiversity^{33,34}.

48 Given the need to understand the patterns of forest structure in Amazonian forests, here we aim
49 to identify how biotic and abiotic factors (Table 1) can explain patterns in forest structure across a
50 successional gradient.

51

52 Table 1. Explanatory variables.

Working Hypothesis/Model	Variables	Source	Description/Ecological relevance	Supporting references
	Plot type	In situ.	Categorical factor with two levels: control and regrowth. Included in all models.	
Topography	Altitude	SRTM-DEM	Altitude is a driver and modulator of species distribution patterns from microhabitat to biogeographic scales. Altitude not only affects soil, water availability, climate and a myriad of other abiotic and biotic variables	10,13,35,36
	Slope	SRTM-DEM	Slope affects soil, water availability and tree mortality rates.	
Hydrography	TWI	SRTM-DEM	Topographic wetness index accounts for the topographic control of water movement in sloped landscapes and the associated control on soil moisture.	37
	Drainage proximity	SRTM-DEM	Combination of water gravitational potential (Height above the nearest drainage (HAND)) and soil drainage (Horizontal distance from nearest drainage (HDND)).	38-40
Regrowth class	Regrowth class	In situ: interview	Land-use history has a strong influence on rates of forest recovery. Categorical factor with three levels of regrowth class derived from the land-use history: late second-regrowth forest, early second-regrowth and pasture.	10,16,41,42
Time	Years since last use	In situ: interview		
	Years since initial clearing	In situ: interview	Time is a major determinant of forest succession.	
Mammals	Species richness	In situ: camera-trap images		
	Functional diversity	In situ: camera-trap images	Mammal diversity is positively related to tree biomass.	17,18,28,43

53

54

55 Table 2. Summary of survey locations. Characteristics of 15 sites used to study forest structure.
 56 Values are means with ranges in parentheses.

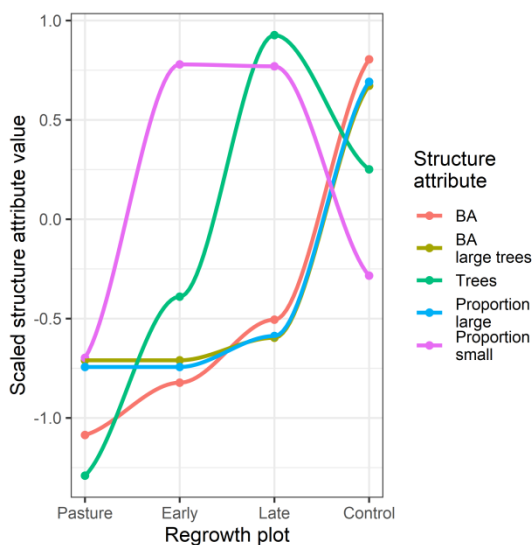
Regrowth class	Sites/ Plots	Size (ha)	Distance		Forest cover (%)		Tree BA (m ² /ha)		Large tree BA (m ² /ha)	
			River (m)	Town (km)	1 km	5 km	Reg.	Cont.	Reg.	Cont.
Late	5/10	5.8	288.15	40.4	91.5	97.4	17.6	35.8	1.5	11.4
		(2.0–12.0)	(110–554)	(35.0–45.0)	(87.4–95.4)	(96.4–98.5)	(11.5–25.3)	(21.7–47.4)	(0.0–7.7)	(0.0–22.3)
Early	5/10	2.4	348.5	38.2	90.7	97.2	11.1	49.3	0.0	20.8
		(1.0–4.5)	(150–554)	(30.0–43.7)	(87.1–96.0)	(96.4–98.4)	(4.7–19.2)	(34.6–76.5)	(0.0–0.0)	(0.0–44.2)
Pasture	5/10	8.7	266.8	32.8	88.8	96.9	5.8	47.1	0.0	24.0
		(6.8–9.9)	(170–461)	(26.8–40.8)	(85.7–91.1)	(95.5–98.5)	(0.0–15.9)	(32.6–64.8)	(0.0–0.0)	(12.0–41.2)
Totals	15/30	5.6	301.1	37.1	90.4	97.2	11.5	44.1	0.5	18.8
		(1.0–12.0)	(110–554)	(26.8–45.0)	(85.7–96.0)	(95.5–98.5)	(0.0–25.3)	(21.7–76.4)	(0.0–7.7)	(0.0–44.2)

57

58 Results

59 Variation in stand structure variables

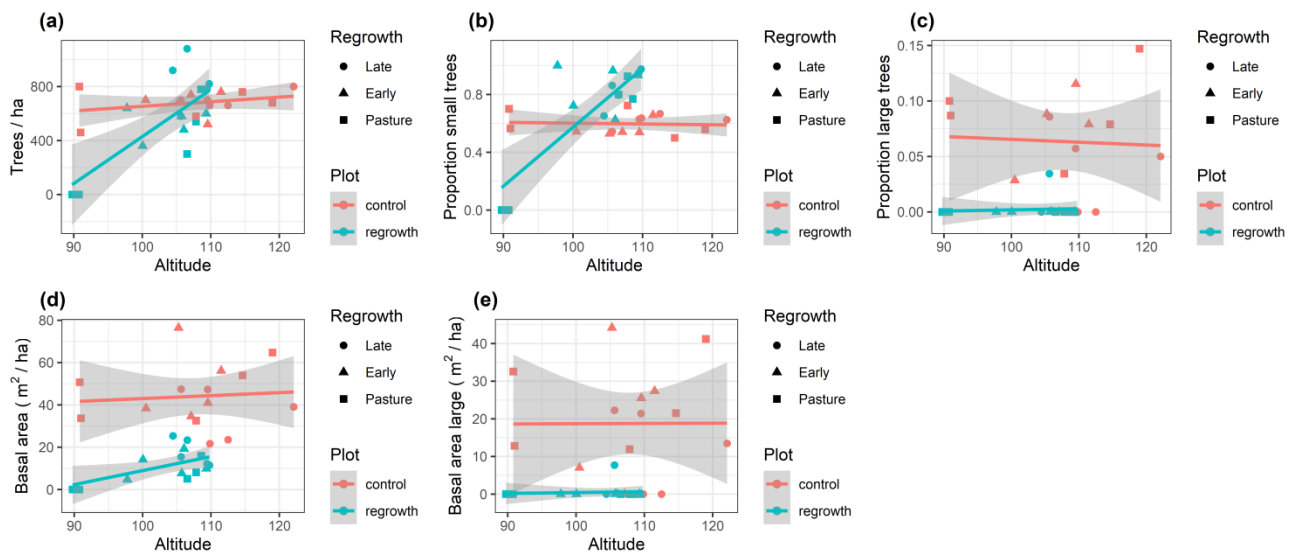
60 There were clear differences in forest structure between control and regrowth plots (Figure 1). On
 61 average control plots had increased basal area and increased proportion of large trees (Figure 1). In
 62 contrast regrowth plots tended to have increased proportion of small trees (<20 cm DBH).



63

64 Figure 1: Forest structure changes across a lowland forest regrowth gradient. Showing mean values of five
 65 forest structure attributes recorded in 30 plots (15 control and 15 regrowth). Regrowth plot shows differences
 66 between control, late second-regrowth, early second-regrowth and pasture plots. Values are scaled (centered

67 and scaled by the standard deviation) to enable simultaneous visual comparison of the different attributes. The
68 lines are from LOESS smoothing as guides to aid visual interpretation.
69
70
71 The number and basal area of living trees tended to increase with altitude and this relationship was
72 stronger in regrowth areas (Figure 2). The relationship with altitude was strongly affected by low
73 lying (90 masl) pasture plots with no trees that generated significant leverage on the linear
74 relationship (Figure 2).

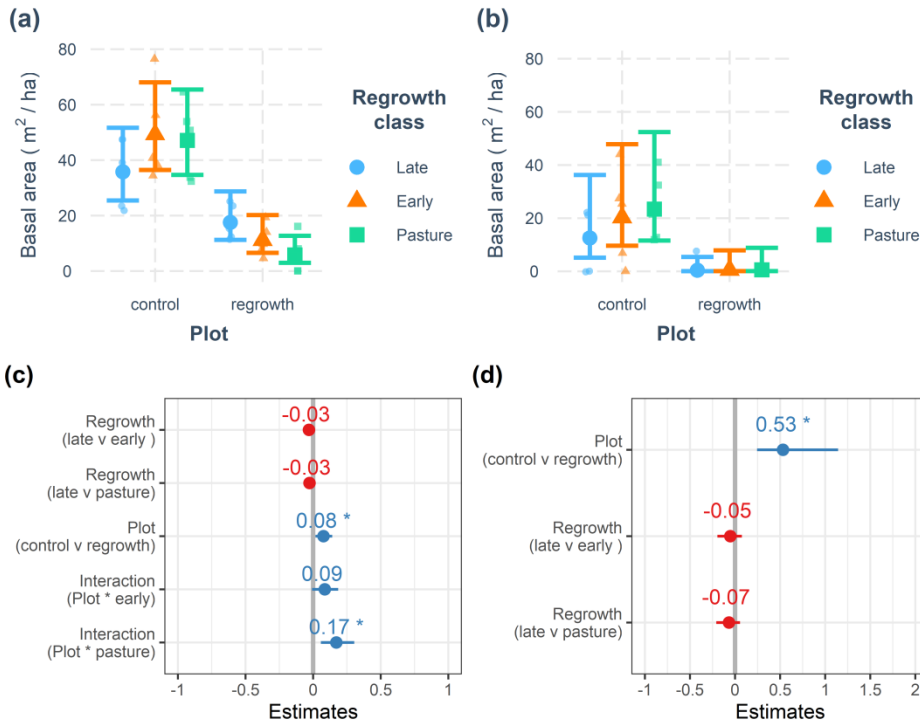


75

76 Figure 2: Forest structure along a lowland Amazon regrowth gradient. Showing trends in (a) number of trees
77 (> 10 cm DBH) per ha, (b) proportion of small trees (10 – 20 cm DBH), (c) proportion of large (> 60 cm
78 DBH) trees, (d) basal area and (e) basal area of large (>60 cm DBH) tree in 30 plots (15 control and 15
79 regrowth). Lines and shaded areas are mean values and 95% confidence intervals from linear models
80 illustrating trends in basal area with increasing altitude (masl). Points with different shapes represent different
81 regrowth classes.

82

83 Basal area ranged from 0 to 76.4 m²/ha across the 30 survey plots (Table 2), with control plots
84 showing an average fourfold increase in basal area compared with regrowth plots (mean basal area
85 44.1 and 11.5 m²/ha, control and regrowth respectively, Figure 2). The patterns in plot basal area also
86 differed between regrowth classes (Figure 3, Supplementary Table S1). There was a significant
87 interaction between plot type (control/regrowth) and regrowth stage, with basal area increasing
88 across pasture, early and late regrowth plots but control plots showing the opposite trend, with basal
89 area decreasing significantly in late-regrowth control plots (Figure 3).



90

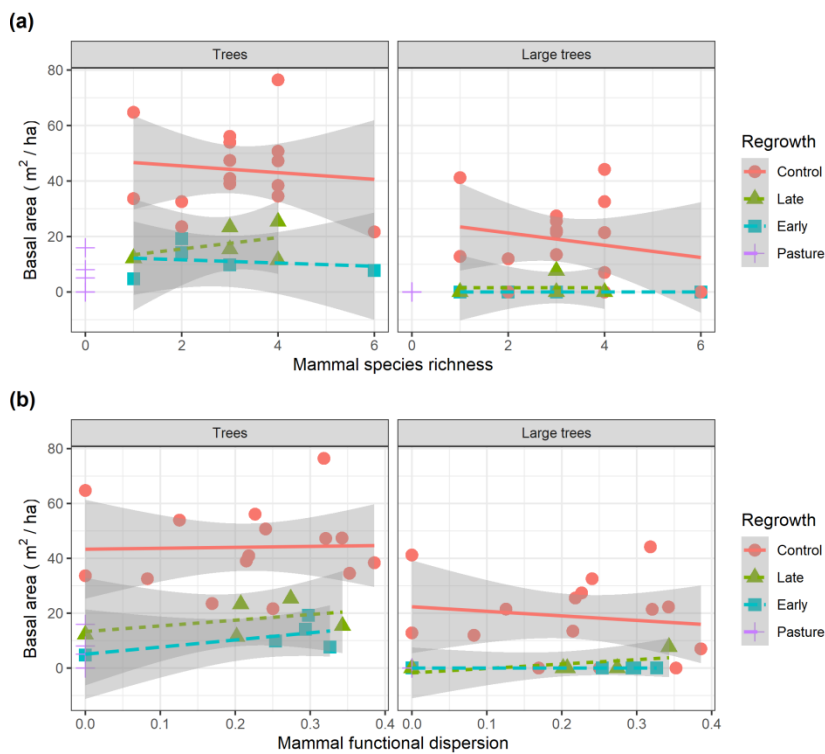
91 Figure 3: Basal area changes across a lowland forest regrowth gradient. The basal area of all (a and c) and
 92 large (b and d) living trees were recorded in 30 plots (15 control and 15 regrowth). Regrowth class shows
 93 differences between late second-regrowth, early second-regrowth and pasture plots contrasted with control
 94 forest plots. Top row shows Generalized Linear Model (GLM) predictions (mean and 95% confidence
 95 intervals) for basal area of (a) all and (b) large trees. Bottom row is the associated Forest-plot of the most
 96 parsimonious GLMs testing for interactions between regrowth class, plot type and years since last use in the
 97 basal area of (c) all and (d) large trees. Forest-plots show coefficient estimates and standard errors.

98

99 There was a highly significant positive linear relationship between overall basal area and large
 100 tree basal area ($F_{1,28} = 127.5$, $R^2 = 0.82$, $P < 0.0001$). The basal area of large trees decreased
 101 significantly in regrowth compared with control plots (Figure 3). On average large trees accounted
 102 for 42% of the basal area in control plots compared with only 4% in regrowth plots (Table 2). Indeed
 103 a single large tree (>60 cm DBH) was recorded only once in a late-regrowth plot. This relationship
 104 was also reflected in the decline in basal area of late-regrowth control plots (Figure 3), which was
 105 associated with a decline on the proportion of large trees that accounted for a reduced 31% of the
 106 basal area in late-regrowth control plots (Table 2).

107 **Relationships between forest structure, mammal diversity and environmental variables**

108 Mammal diversity varied considerably across the survey plots (Figure 4). There appeared to be
109 a tendency for basal area to increase with mammal diversity in Late-regrowth plots, yet basal area
110 was only weakly associated with mammal diversity within the different regrowth classes (Figure 4).
111 Indeed, the diversity of mammals was found to be only weakly informative for explaining the basal
112 area of trees across the 30 sample plots (Table 3).
113



114 Figure 4. Mammal diversity and basal area across a lowland forest regrowth gradient. The basal area of (a) all
115 and (b) large living trees were recorded together with the diversity (species richness and functional dispersion)
116 of terrestrial mammals in 30 plots (15 control and 15 regrowth). Lines and shaded areas are mean values and
117 95% confidence intervals from linear models illustrating trends in basal area with increasing mammal
118 diversity. Points with different shapes represent different regrowth plot types.
119

120
121 Comparison of the models representing the alternative hypotheses showed that plot type
122 (control v regrowth) and topography were the most important first ranked variables for the five forest
123 structure attributes (Table 3). The most simple model including only plot type explained more than
124 50% of model deviance for all forest structure attributes except for the number of trees (DBH>10
125 cm). Plot type and regrowth class were both included in the 95% confidence set of models for the

126 basal area of large trees (Table 3). In contrast Topography was the most important (first ranked)
127 model for the number of trees, proportion of small trees and tree basal area (Table 3). Mammal
128 diversity, Time and Hydrography models were not well supported and were not included in the 95%
129 confidence set of models for any of the forest structure attributes (Table 3).
130

131 Table 3. Summary of the Generalized Linear Models created to explain forest structure in 30 plots
 132 (15 control and 15 regrowth). Models ordered by decreasing AIC (Akaike Information Criterion)
 133 values.

Forest structure	Model ^a	Dev. Exp	Loglik	BIC	AIC	Δ AIC	W_i AIC ^b
Trees							
	Topography	60.5	-100.6	231.9	219.3	0.0	0.91
	Regrowth class	49.8	-105.1	234.1	224.3	5.0	0.07
	Mammals	51.3	-104.7	240.0	227.3	8.1	0.02
	Hydrography	45.8	-106.3	243.1	230.5	11.3	0.00
	Time	42.9	-107.0	244.7	232.1	12.8	0.00
	Plot	6.5	-114.4	239.1	234.9	15.6	0.00
Proportion small trees							
	Topography	75.2	-64.3	155.7	144.5	0.0	0.96
	Hydrography	70.4	-67.5	162.2	151.0	6.5	0.04
	Mammals	66.4	-70.2	167.6	156.4	11.9	0.00
	Time	65.1	-71.0	169.3	158.1	13.6	0.00
	Plot	53.7	-78.7	164.3	161.5	17.0	0.00
	Regrowth class	55.4	-77.6	175.6	167.2	22.7	0.00
Proportion large trees							
	Plot	54.9	-30.5	67.8	65.0	0.0	0.88
	Regrowth class	60.6	-28.9	78.2	69.8	4.8	0.08
	Mammals	62.9	-28.3	83.7	72.5	2.8	0.02
	Time	62.8	-28.3	83.8	72.5	2.8	0.02
	Topography	54.7	-30.6	88.5	77.3	7.5	0.00
	Hydrography	52.8	-31.1	89.5	78.3	8.5	0.00
BA all trees							
	Topography	80.7	-100.9	232.3	220.4	0.0	0.89
	Regrowth class	72.9	-105.8	235.4	225.6	5.2	0.06
	Plot	62.9	-110.4	231.0	226.8	6.4	0.04
	Mammals	71.7	-106.4	243.4	230.8	10.4	0.01
	Time	69.3	-107.6	245.8	233.2	12.8	0.00
	Hydrography	66.0	-109.1	248.8	236.2	15.9	0.00
BA large trees							
	Plot	58.3	-61.5	133.1	128.9	0.0	0.87
	Regrowth class	67.1	-59.9	143.5	133.7	4.8	0.08
	Hydrography	70.9	-59.0	148.7	136.1	7.1	0.02
	Mammals	70.4	-59.2	148.9	136.3	7.4	0.02
	Time	65.7	-60.1	150.9	138.3	9.4	0.01
	Topography	60.8	-61.0	152.7	140.1	11.2	0.00

134 ^a Models used to explain forest structure. All models contained plot type (control/regrowth) as categorical
 135 factor. Variables and associated estimates in the different models can be found as Supplementary Information
 136 Table S2 online.

137 ^b Akaike weights (W_i) from largest to smallest.

138 Discussion

139 We integrate field and remotely sensed data to establish support for multiple non-mutually exclusive
140 hypotheses explaining patterns in forest structure across a lowland Amazon regrowth gradient. We
141 establish that different hypotheses are supported for different structure attributes. Here we discuss
142 these findings in terms of prospects for the passive restoration of degraded Amazon forests.

143 The mean basal area value from our 15 control plots ($44.1 \text{ m}^2/\text{ha}$) was close to the mean from
144 42 Guyana Shield forest plots ($43.4 \text{ m}^2/\text{ha}$, range $10 - 65 \text{ m}^2/\text{ha}$) in French Guiana³⁵. The results
145 from Molto, et al.³⁵ were obtained from an extensive survey of $0.5 - 1 \text{ ha}$ plots. Although our plot
146 size was smaller compared to Molto, et al.³⁵, the similarity in mean values suggests that our plots do
147 provide a representative sample of forest structure in the regrowth areas. The basal areas obtained
148 from our regrowth plots followed a similar trajectory to those reported from abandoned pasture in
149 Costa Rica¹⁰, where the most recently abandoned pasture plots (<14 year) had mean basal area of
150 $13.5 \text{ m}^2/\text{ha}$, with basal area increasing to $26.1 \text{ m}^2/\text{ha}$ after 21 – 30 year¹⁰. compared with 11.1 and
151 $17.6 \text{ m}^2/\text{ha}$ respectively in our Early (1-5 year) and Late (20 – 25 year) regrowth plots. This also
152 follows a similar pattern to values reported from 370 successional forest plots in the Brazilian
153 Amazon, with basal area values typically $< 10 \text{ m}^2/\text{ha}$ in early stages (< 5 year) and reaching $25 \text{ m}^2/\text{ha}$
154 after 15 years⁴⁴.

155 Although results from lowland forest sites in Costa Rica suggest rapid recovery of pasture
156 areas¹⁰ this could be related to the substantially lower basal area in the seven old growth reference
157 plots ($26.1 \text{ m}^2/\text{ha}$, range $19.3 - 32.2 \text{ m}^2/\text{ha}$) compared with those in our study area. Our results are
158 similar to those reported from the central Amazon, where 25 y of regrowth restored half of the
159 mature-forest biomass⁴¹. A recent analysis of 45 Neotropical secondary forest study sites found that
160 secondary forests in the lowland tropics reach 90 percent of old growth biomass in a median time of
161 66 yr¹³. Our findings do suggest nuanced difference in successional trajectories. Basal area increased
162 rapidly in early regrowth stages and this could be explained by the less intensive land use (i.e. lack of

163 pasture) and the proximity to large areas of intact forest. In contrast basal area of late-regrowth areas
164 was less than those reported from other areas^{10,44}. This could be related to soil productivity, as
165 previous studies show that highly diverse Guyana Shield wet forests can take longer to establish¹³.
166 With basal area of control plots dominated by large trees it seems likely that many decades will be
167 necessary for forest structure (total basal area, proportion of large trees) to return to pre-disturbance
168 values.

169 The success of active and passive restoration can depend on ecological conditions⁴⁵. We found
170 topography was the most informative model for explaining patterns in number of trees, tree basal
171 area and proportion of small trees (Table 3, Figure 2). Differences in altitude and slope have been
172 shown to affect floristic structure of tropical forests from local to regional scales^{16,35,42,46-48}. Indeed,
173 even relatively small variations in topography can generate changes in local-scale soil chemistry,
174 hydrology and microclimate^{46,49}. The effects of topography do not operate in isolation from
175 hydrology and the increased numbers of small trees and tree biomass with increasing altitude (Figure
176 3) agree with previous studies that show trees grow more slowly in more low lying (and often more
177 waterlogged) terrain⁴².

178 We found a weak association between mammal diversity and regrowth forest structure.
179 Previous studies in a nearby protected area show that this group of mammals (mid- to large-bodied
180 Artiodactyla, Perissodactyla and Rodentia) are more strongly associated with factors such as access
181 to water⁵⁰ and altitude^{50,51}. A recent study also showed that mammal abundances were more
182 strongly associated with phenology (fruit fall) than basal area along 10 km of forest in the western
183 Guyana Shield⁵². Additionally, regrowth class was found to be the primary driver of mammal species
184 encountered independent of forest cover⁵³. For example the number of species detected in control
185 and regrowth plots (all with forest cover >87%) varied between 1 and 6 (Figure 4). Mid- to large-
186 bodied seed dispersers are a critical component of Amazon forests^{18,19,25} and are also widespread
187 and ubiquitous across myriad Amazonian forest types⁵⁴⁻⁵⁶. The eight species are therefore not

188 strictly dependent on the quality of forest habitat compared with other more specialist groups such as
189 primates⁵⁷. The lack of a strong relationship between diversity of these eight mammal seed
190 dispersers and forest structure attributes (i.e. overall basal area and proportion of small trees) is
191 therefore to be expected.

192 Decades of research show that myriad edge effects can extend up to 150 m in fragmented
193 Amazon forests^{58,59}. Considering the range of expected edge-effects it is highly probable that the
194 natural regeneration and/or restoration of regrowth habitats in Amazon small-holdings (typically <
195 100 ha) will strongly depend on species ecological responses to habitat edges⁶⁰. Previous studies
196 show that edge effects increase mortality of large trees, which in turn has major impacts on forest
197 ecosystems⁶¹. In highly fragmented areas edge-effects can drive tree communities through a process
198 of “retrogressive succession”⁶² and toward an early successional state that may persist indefinitely.
199 This early successional state can be characterized by functional and structural differences in that
200 larger slower-growing tree species with high wood density tend to decline whereas faster-growing
201 tree and liana species with lower wood density increase^{62,63}. The decline in the number and basal
202 area of large trees from our control plots along 20-25 year old edges suggest that retrogressive
203 succession may establish even in relatively un-fragmented areas surrounded by extensive forest
204 cover.

205 Our findings provide an early warning that even under a best case scenario there is potential for
206 “retrogressive succession”. We found not only a lack of large trees in regrowth plots but also that
207 large tree basal declined in older late-regrowth control plots. We suggest that this decline in large
208 trees may be the primary driver of differences between regrowth and old growth forest and as such
209 represent an unquantified component of resilience and time to recovery of Neotropical secondary
210 forests. We also suggest that the continued presence of mid- and large bodied mammal seed
211 dispersers in the study area are likely to be vital in order to avoid such “retrogressive succession”.

212 **Methods**

213 **Ethics Statement**

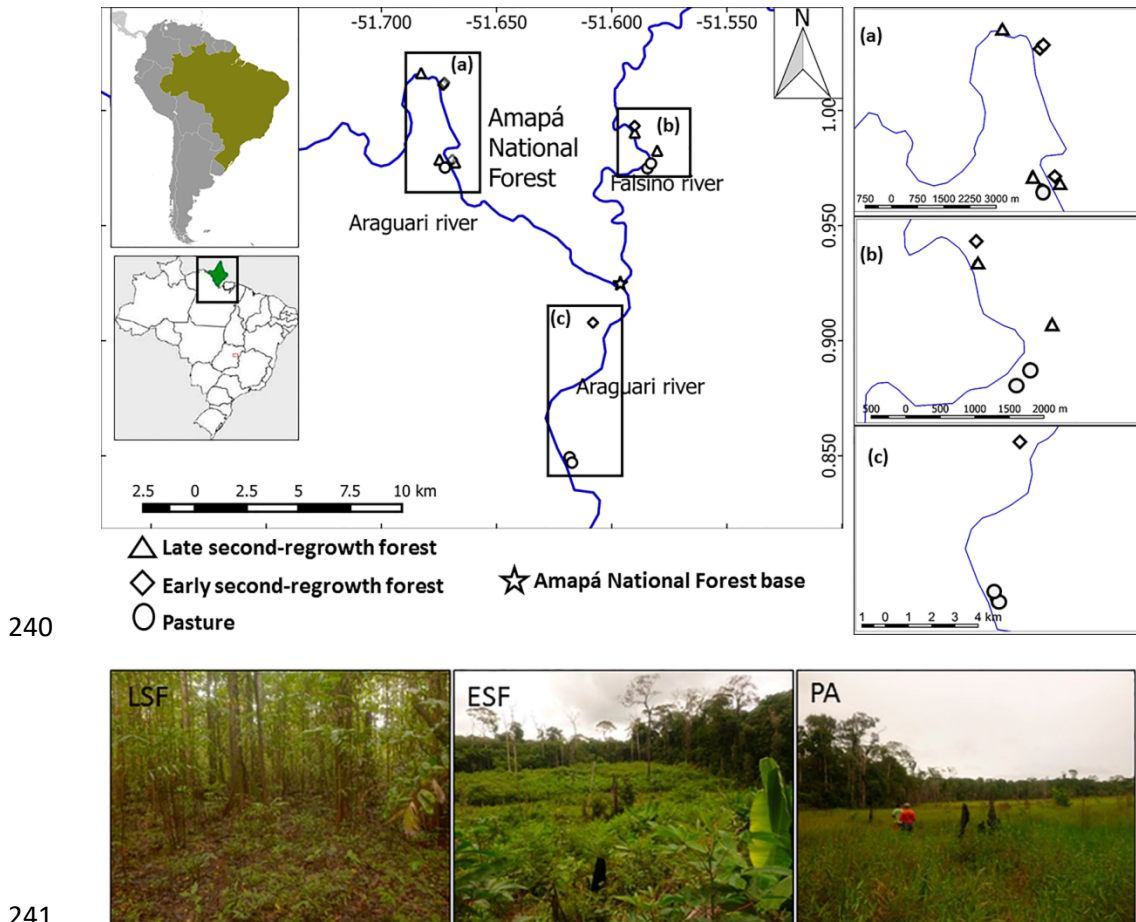
214 All methods were carried out in accordance with relevant guidelines and regulations. Fieldwork and
215 data collection was conducted under research permit numbers SISBIO 40355–1, 47859-1 and 47859-
216 2 to DN, issued by the Brazilian Ministério do Meio Ambiente (“MMA”). Data collection used non-
217 invasive, remotely activated camera traps and did not involve direct contact or interaction with
218 animals, thus no ethical approval was required. Interviews with local residents were approved by
219 Brazilian Ministério do Meio Ambiente (SISBIO permits 45034-1, 45034-2, 45034-3) and the Ethics
220 Committee in Research from the Federal University of Amapá (UNIFAP) (CAAE
221 42064815.5.0000.0003, Permit number 1.013.843). Interviews were conducted with residents that
222 were both (1) willing to be interviewed (written informed consent was obtained from all interviewees) and (2)
223 aware of the site history.

224

225 **Study area**

226 Our study took place in 15 areas of regrowth on small holder properties^{53,64} in the center of the State
227 of Amapá (Figure 5). The regional climate is classified by Koppen-Geiger as Am (Equatorial
228 monsoon)⁶⁵, with annual rainfall greater than 2500 mm⁶⁶. The driest months are September to
229 November (total monthly rainfall < 150 mm) and the wettest months from February to April (total
230 monthly rainfall > 300 mm)⁶⁶. The State of Amapá has the lowest deforestation rate in Brazil and >
231 70% of the Amapá receives some form of legal protection. There is no large scale agricultural
232 developments or monocultures along the waterways and properties retain typically small (< 1000 ha)
233 areas of opened land, which are cleared for small scale family agriculture, which focuses on acai,
234 small scale production of fruits and vegetables for sustenance and limited commercial sale of
235 regional produce (e.g. manioc flour) in local markets. There are some 54 properties upstream of the
236 nearest town (Porto Grande⁶⁷). There has never been any expansive clearcutting in the region and

237 there are no monocultures (e.g. soy) or cattle production. All sites were at least 26 km from the
238 nearest town by river, and all sites are surrounded by matrix of continuous closed canopy forest
239 cover (Table 2). Pesticides and/or herbicides had never been used at any of the sites.



242 **Fig. 5.** Map of the study area in the eastern Amazon. Showing the location of 15 study sites,
243 grouped into three regrowth stages in the small holder properties close to rivers (solid blue lines): late
244 second-regrowth forest (LSF, triangles), early second-regrowth forest (ESF, squares) and pasture
245 (PA, circles).
246

247 A described previously^{53,64} the 15 small-holder properties were selected based on differences in
248 land-use histories and forest succession/regrowth stage. All sites were close (110 – 554m, Table 2) to
249 100 – 200 m wide rivers that are navigable by motorized boats, but due to riverbank formation the
250 sites are never flooded. These 15 sites were grouped into three regrowth classes based on the land-
251 use history: late second-regrowth forest (N = 5, most recent human disturbance between 20 and 25
252 years), early second-regrowth (N = 5, most recent human disturbance between 1 and 5 years), and

253 pasture (N = 5, recently cleared and abandoned pasture areas dominated by grasses/herbs but that had
254 never been used to raise livestock, with the most recent disturbance between 1 and 17 years). Each of
255 the 15 regrowth sites was paired with a nearby (60 to 150 m) control site i.e. 20 – 30 m tall *terra-*
256 *firme* forest site without a history of mechanized timber extraction. To reduce the possible
257 confounding influence of edge effects that are known to strongly influence the distribution of trees in
258 Neotropical forests, all regrowth and control sites were established at a standardized distance
259 (approximately 30 m) from the nearest control-regrowth habitat edge.

260

261 **Forest structure**

262 Data were collected from May to August 2016. Forest structure data (i.e., number of trees and
263 basal area) were obtained from plots measuring 50 x 10 m (500 m²), at each of the 30 points, totaling
264 1.5 hectare. This plot size was selected as it has been widely used to examine structural changes in
265 tropical forests^{19,41,42,68} and several of the regrowth areas were too small (Table 2) to enable the
266 establishment of larger spatially independent plots. We obtained five measures (responses) to
267 characterize the forest structure in each plot. These were selected based on previous studies that
268 show their appropriateness to distinguish attributes of regrowth/successional stages related to
269 biodiversity of Amazon forests^{13,44,69,70}. The number of all trees ≥ 10 cm DBH (diameter at breast
270 height at a standard 1.3 m above ground, or above tallest root buttress) was used to quantify the
271 number of trees per area in each plot (m²). This count included all trees which had at least half of
272 their basal trunk inside the plot. The proportion of small (10 – 20 cm DBH) trees was calculated to
273 represent the expected increase of younger trees in regrowth areas. The proportion of large (>60 cm
274 DBH) trees was calculated as this is known as an important characteristic of mature/late succession
275 areas^{44,70}. We also calculated the basal area of all and large trees as this is known to be strongly
276 correlated with tree biomass⁷¹. For example basal area and biomass were > 99% correlated in 23
277 plots from lowland Costa Rica¹⁰.

278

279 **Explanatory variables**

280 We investigated predictions from multiple non-mutually exclusive hypotheses to explain
281 patterns in basal area (Table 1). A total of 10 variables were used to form models to represent 5
282 working hypotheses (topography, hydrography, regrowth class, time and mammal diversity) that
283 based on the findings from previous studies were likely to explain the observed patterns^{10,13,28,40,41,72}.
284 We chose to work with mainstream, widely available environmental variables. Four of these (the
285 topographic and hydrographic model variables) were computed from remotely sensed digital terrain
286 model (SRTM-DTM): altitude (masl), slope, TWI (Topographic wetness index), DND (Distance to
287 Network Drainage) calculated from the interaction between HAND (Height above network drainage)
288 and HDND (Horizontal distance to network drainage). The time model included years since the
289 regrowth site was opened and years since last use, both of which were obtained from interviews with
290 local landowners.

291 Mammal functional diversity was obtained from a camera-trap survey conducted at the same
292 time (May to September 2016) and in the same plots as forest structure was sampled⁵³. Camera traps
293 equipped with infrared triggers (Bushnell Trophy Cam, 8MP, Overland Park, KS, USA) were
294 installed in each of the 30 plots following standardized protocols^{50,51,73}. This camera trap survey [full
295 details provide in⁵³] including a sampling effort of 827 camera-trap days (450 and 377 camera-trap
296 days, control and regrowth sites respectively) was used to estimate functional diversity of eight
297 terrestrial mammal seed dispersers (*Cuniculus paca*, *Dasyprocta leporina*, *Myoprocta acouchy*,
298 *Mazama americana*, *M. nemorivaga*, *Pecari tajacu*, *Tayassu pecari* and *Tapirus terrestris*).

299

300 **Data analysis**

301 Tree Basal Area in each plot was obtained as the sum of the basal area value for each
302 individual tree derived from the DBH of each tree following the formula BA (basal area in m^2)=

303 0.00007854 X DBH² (constant obtained by solving the following equation to obtain BA in m² from
304 the DBH measured in cm ⁶⁹):

$$BA = \frac{\pi \times (DBH/2)^2}{10000}$$

305 We calculated basal area of all and large (>60cm DBH) living trees ^{69,70,72}. We also calculated the
306 proportion of small stems (10 – 20 cm DBH trees) as this has been shown to be an important
307 measure of stand structure in forest regrowth areas ^{35,69}.

308 To represent diversity of terrestrial mammal seed dispersers we calculated a richness and
309 functional diversity (FD) value for each of the 30 plots ⁵³. Richness was calculated as the observed
310 number of species (hereafter “species richness”) at each plot. Although there are many diversity
311 metrics, we chose species richness as it is widely used and clearly interpretable ^{74,75} and with
312 relatively few (eight) species and 30 plots there were strong correlations between species richness
313 values and alternative diversity metrics such as Shannon and Simpson diversity (Spearman rho >
314 0.89). We used Functional Dispersion (FDis) ⁷⁶ as an index of functional diversity as it is not
315 strongly influenced by outliers, accounts for relative abundances, is unaffected by species richness
316 and can be calculated from any distance/dissimilarity measure ^{76,77}. Functional Dispersion was
317 estimated with the dbFD function ⁷⁷ using default settings.

318 To examine patterns in forest structure attributes we used Generalized Linear Models. We used
319 an information theoretic model averaging framework ⁷⁸ to examine the support for five models
320 representing the five non-mutually exclusive hypotheses – topography, hydrography, regrowth class,
321 time and mammal diversity (see Table 1 for variable description and ecological relevance). We
322 evaluated models based on their information content, as measured by AIC – Akaike Information
323 Criterion. The relative importance of the models was measured by the models Akaike weights
324 (Burnham & Anderson 2002 pp. 75-77, 167-172), which is a scaled measure of the likelihood ratio

325 that ranges between 0 (least important) and 1 (most important). None of the unexplained variation
326 (model residuals) was related to the geographic distance among plots so we did not need to control
327 for spatial dependence. All analysis were conducted using the R language and environment for
328 statistical computing ⁷⁹, with base functions and functions available in the following packages: vegan
329 ⁸⁰, ggplot2 ⁸¹, MuMIn ⁸², and tweedie ⁸³.

330

331

332 Data Availability Statement

333 The raw forest structure and environmental data used in the analysis of this study have been
334 deposited in the OSF - Center for Open Science at DOI: 10.17605/OSF.IO/MC27U.

335

336 References

- 337 1 Hansen, M. C. *et al.* High-Resolution Global Maps of 21st-Century Forest Cover Change. *Science* **342**,
338 850-853, doi:<https://doi.org/10.1126/science.1244693> (2013).
- 339 2 Chazdon, R. L. & Guariguata, M. R. Natural regeneration as a tool for large-scale forest restoration in
340 the tropics: prospects and challenges. *Biotropica* **48**, 716–730,
341 doi:<https://doi.org/10.1111/btp.12381> (2016).
- 342 3 Holl, K. D. Restoring tropical forests from the bottom up. *Science* **355**, 455-456,
343 doi:<https://doi.org/10.1126/science.aam5432> (2017).
- 344 4 Brancalion, P. H. S. *et al.* Balancing economic costs and ecological outcomes of passive and active
345 restoration in agricultural landscapes: the case of Brazil. *Biotropica* **48**, 856-867,
346 doi:<https://doi.org/10.1111/btp.12383> (2016).
- 347 5 Foley, J. A. *et al.* Amazonia revealed: forest degradation and loss of ecosystem goods and services in
348 the Amazon Basin. *Frontiers in Ecology and the Environment* **5**, 25-32, doi:[https://doi.org/10.1890/1540-9295\(2007\)5\[25:ARFDAL\]2.0.CO;2](https://doi.org/10.1890/1540-9295(2007)5[25:ARFDAL]2.0.CO;2) (2007).
- 350 6 Montibeller, B., Kmoch, A., Virro, H., Mander, Ü. & Uuemaa, E. Increasing fragmentation of forest
351 cover in Brazil's Legal Amazon from 2001 to 2017. *Scientific Reports* **10**, 5803,
352 doi:<https://doi.org/10.1038/s41598-020-62591-x> (2020).
- 353 7 Csillik, O., Kumar, P., Mascaro, J., O'Shea, T. & Asner, G. P. Monitoring tropical forest carbon stocks
354 and emissions using Planet satellite data. *Scientific Reports* **9**, 17831, doi:<https://doi.org/10.1038/s41598-019-54386-6> (2019).
- 356 8 Nunes, S. *et al.* Uncertainties in assessing the extent and legal compliance status of riparian forests
357 in the eastern Brazilian Amazon. *Land Use Policy* **82**, 37-47,
358 doi:<https://doi.org/10.1016/j.landusepol.2018.11.051> (2019).
- 359 9 Rocha, G. P. E., Vieira, D. L. M. & Simon, M. F. Fast natural regeneration in abandoned pastures in
360 southern Amazonia. *Forest Ecology and Management* **370**, 93-101,
361 doi:<https://doi.org/10.1016/j.foreco.2016.03.057> (2016).
- 362 10 Letcher, S. G. & Chazdon, R. L. Rapid Recovery of Biomass, Species Richness, and Species
363 Composition in a Forest Chronosequence in Northeastern Costa Rica. *Biotropica* **41**, 608-617,
364 doi:<https://doi.org/10.1111/j.1744-7429.2009.00517.x> (2009).
- 365 11 Fearnside, P. M. Deforestation in Brazilian Amazonia: History, Rates, and Consequences.
366 *Conservation Biology* **19**, 680-688, doi:<https://doi.org/10.1111/j.1523-1739.2005.00697.x> (2005).
- 367 12 Laurance, W. F. *et al.* Rain forest fragmentation and the proliferation of successional trees. *Ecology*
368 **87**, 469-482, doi:<https://doi.org/10.1890/05-0064> (2006).
- 369 13 Poorter, L. *et al.* Biomass resilience of Neotropical secondary forests. *Nature* **530**, 211-214,
370 doi:<https://doi.org/10.1038/nature16512> (2016).
- 371 14 Camargo, J. L. C., Ferraz, I. D. K. & Imakawa, A. M. Rehabilitation of degraded areas of central
372 Amazonia using direct sowing of forest tree seeds. *Restoration Ecology* **10**, 636-644,
373 doi:<https://doi.org/10.1046/j.1526-100X.2002.01044.x> (2002).
- 374 15 Guariguata, M. R. & Ostertag, R. Neotropical secondary forest succession: changes in structural and
375 functional characteristics. *Forest ecology and management* **148**, 185-206,
376 doi:[https://doi.org/10.1016/S0378-1127\(00\)00535-1](https://doi.org/10.1016/S0378-1127(00)00535-1) (2001).

- 377 16 Crouzeilles, R. *et al.* A global meta-analysis on the ecological drivers of forest restoration success.
378 *Nature Communications* **7**, 11666, doi:[https://10.1038/ncomms11666](https://doi.org/10.1038/ncomms11666) (2016).
- 379 17 Chazdon, R. L. *et al.* The Potential for Species Conservation in Tropical Secondary Forests.
380 *Conservation Biology* **23**, 1406-1417, doi:[https://10.1111/j.1523-1739.2009.01338.x](https://doi.org/10.1111/j.1523-1739.2009.01338.x) (2009).
- 381 18 Peres, C. A., Emilio, T., Schiatti, J., Desmouliere, S. J. & Levi, T. Dispersal limitation induces long-term
382 biomass collapse in overhunted Amazonian forests. *Proceedings of the National Academy of Sciences*
383 *of the United States of America* **113**, 892-897, doi:[https://10.1073/pnas.1516525113](https://doi.org/10.1073/pnas.1516525113) (2016).
- 384 19 Pessoa, M. S. *et al.* Deforestation drives functional diversity and fruit quality changes in a tropical
385 tree assemblage. *Perspectives in Plant Ecology, Evolution and Systematics* **28**, 78-86,
386 doi:<https://doi.org/10.1016/j.ppees.2017.09.001> (2017).
- 387 20 Bowen, M. E., McAlpine, C. A., House, A. P. & Smith, G. C. Regrowth forests on abandoned
388 agricultural land: a review of their habitat values for recovering forest fauna. *Biological Conservation*
389 **140**, 273-296, doi:<http://dx.doi.org/10.1016/j.biocon.2007.08.012> (2007).
- 390 21 Chazdon, R. L. & Uriarte, M. Natural regeneration in the context of large-scale forest and landscape
391 restoration in the tropics. *Biotropica* **48**, 709-715, doi:<http://dx.doi.org/10.1111/btp.12409> (2016).
- 392 22 Neuschulz, E. L., Mueller, T., Schleuning, M. & Böhning-Gaese, K. Pollination and seed dispersal are
393 the most threatened processes of plant regeneration. *Scientific Reports* **6**, 29839,
394 doi:10.1038/srep29839 (2016).
- 395 23 Stoner, K. E., Riba-Hernández, P., Vulinec, K. & Lambert, J. E. The role of mammals in creating and
396 modifying seedshadows in tropical forests and some possible consequences of their elimination.
397 *Biotropica* **39**, 316-327, doi:<http://dx.doi.org/10.1111/j.1744-7429.2007.00292.x> (2007).
- 398 24 Griffiths, H. M., Bardgett, R. D., Louzada, J. & Barlow, J. The value of trophic interactions for
399 ecosystem function: dung beetle communities influence seed burial and seedling recruitment in
400 tropical forests. *Proceedings of the Royal Society B* **283**, 20161634,
401 doi:<http://dx.doi.org/10.1098/rspb.2016.1634> (2016).
- 402 25 Asquith, N. M. & Mejía-Chang, M. Mammals, edge effects, and the loss of tropical forest diversity.
403 *Ecology* **86**, 379-390, doi:<http://dx.doi.org/10.1890/03-0575> (2005).
- 404 26 Beck, H., Snodgrass, J. W. & Thebpanya, P. Long-term exclosure of large terrestrial vertebrates:
405 Implications of defaunation for seedling demographics in the Amazon rainforest. *Biological*
406 *Conservation* **163**, 115-121, doi:<http://dx.doi.org/10.1016/j.biocon.2013.03.012> (2013).
- 407 27 Paine, C. E., Beck, H. & Terborgh, J. How mammalian predation contributes to tropical tree
408 community structure. *Ecology* **97**, 3326-3336, doi:<https://doi.org/10.1002/ecy.1586> (2016).
- 409 28 Sobral, M. *et al.* Mammal diversity influences the carbon cycle through trophic interactions in the
410 Amazon. *Nature Ecology & Evolution* **1**, 1670-1676, doi:[https://10.1038/s41559-017-0334-0](https://doi.org/10.1038/s41559-017-0334-0) (2017).
- 411 29 Bascompte, J. & Jordano, P. Plant-animal mutualistic networks: the architecture of biodiversity.
412 *Annual Review of Ecology, Evolution, and Systematics* **38**, 567-593,
413 doi:<http://dx.doi.org/10.1146/annurev.ecolsys.38.091206.095818> (2007).
- 414 30 Wunderle, J. M. The role of animal seed dispersal in accelerating native forest regeneration on
415 degraded tropical lands. *Forest Ecology and Management* **99**, 223-235,
416 doi:[http://dx.doi.org/10.1016/S0378-1127\(97\)00208-9](http://dx.doi.org/10.1016/S0378-1127(97)00208-9) (1997).
- 417 31 Fragoso, J. M. V. Tapir-Generated Seed Shadows: Scale-Dependent Patchiness in the Amazon Rain
418 Forest. *Journal of Ecology* **85**, 519-529, doi:<http://dx.doi.org/10.2307/2960574> (1997).
- 419 32 Hibert, F. *et al.* Unveiling the diet of elusive rainforest herbivores in next generation sequencing era?
420 The tapir as a case study. *PLoS One* **8**, e60799, doi:<http://dx.doi.org/10.1371/journal.pone.0060799>
421 (2013).
- 422 33 Terborgh, J. *et al.* Tree recruitment in an empty forest. *Ecology* **89**, 1757-1768,
423 doi:<http://dx.doi.org/10.1890/07-0479.1> (2008).
- 424 34 Wright, S. J. *et al.* The plight of large animals in tropical forests and the consequences for plant
425 regeneration. *Biotropica* **39**, 289-291, doi:<http://dx.doi.org/10.1111/j.1744-7429.2007.00293.x>
426 (2007).

- 427 35 Molto, Q. *et al.* Predicting tree heights for biomass estimates in tropical forests; a test from French
428 Guiana. *Biogeosciences* **11**, 3121-3130, doi:[https://10.5194/bg-11-3121-2014](https://doi.org/10.5194/bg-11-3121-2014) (2014).
- 429 36 Körner, C. The use of 'altitude' in ecological research. *Trends in Ecology & Evolution* **22**, 569-574,
430 doi:<https://doi.org/10.1016/j.tree.2007.09.006> (2007).
- 431 37 Beven, K. & Kirkby, M. J. A physically based, variable contributing area model of basin hydrology.
432 *Hydrological Sciences Journal* **24**, 43-69, doi:<https://doi.org/10.1080/02626667909491834> (1979).
- 433 38 Campling, P., Gobin, A. & Feyen, J. Logistic Modeling to Spatially Predict the Probability of Soil
434 Drainage Classes. *Soil Science Society of America Journal* **66**, 1390-1401,
435 doi:<https://10.2136/sssaj2002.1390> (2002).
- 436 39 Nobre, A. D. *et al.* Height Above the Nearest Drainage – a hydrologically relevant new terrain model.
437 *Journal of Hydrology* **404**, 13-29, doi:<http://dx.doi.org/10.1016/j.jhydrol.2011.03.051> (2011).
- 438 40 Schiatti, J. *et al.* Vertical distance from drainage drives floristic composition changes in an Amazonian
439 rainforest. *Plant Ecology & Diversity* **7**, 241-253, doi:<https://10.1080/17550874.2013.783642> (2014).
- 440 41 Gehring, C., Denich, M. & Vlek, P. L. G. Resilience of secondary forest regrowth after slash-and-burn
441 agriculture in central Amazonia. *Journal of Tropical Ecology* **21**, 519-527,
442 doi:<https://10.1017/S0266467405002543> (2005).
- 443 42 Feldpausch, T. R., Riha, S. J., Fernandes, E. C. M. & Wandelli, E. V. Development of Forest Structure
444 and Leaf Area in Secondary Forests Regenerating on Abandoned Pastures in Central Amazônia. *Earth*
445 *Interactions* **9**, 1-22, doi:<https://10.1175/EI140.1> (2005).
- 446 43 Luskin, M. S., Ickes, K., Yao, T. L. & Davies, S. J. Wildlife differentially affect tree and liana
447 regeneration in a tropical forest: An 18-year study of experimental terrestrial defaunation versus
448 artificially abundant herbivores. *Journal of Applied Ecology* **56**, 1379-1388,
449 doi:<https://10.1111/1365-2664.13378> (2019).
- 450 44 Lu, D., Mausel, P., Brondízio, E. & Moran, E. Classification of successional forest stages in the
451 Brazilian Amazon basin. *Forest Ecology and Management* **181**, 301-312,
452 doi:[https://doi.org/10.1016/S0378-1127\(03\)00003-3](https://doi.org/10.1016/S0378-1127(03)00003-3) (2003).
- 453 45 Crouzeilles, R. *et al.* Ecological restoration success is higher for natural regeneration than for active
454 restoration in tropical forests. *Science Advances* **3**, e1701345, doi:<https://10.1126/sciadv.1701345>
455 (2017).
- 456 46 de Castilho, C. V. *et al.* Variation in aboveground tree live biomass in a central Amazonian Forest:
457 Effects of soil and topography. *Forest Ecology and Management* **234**, 85-96,
458 doi:<https://doi.org/10.1016/j.foreco.2006.06.024> (2006).
- 459 47 Jucker, T. *et al.* Topography shapes the structure, composition and function of tropical forest
460 landscapes. *Ecology Letters* **21**, 989-1000, doi:<https://10.1111/ele.12964> (2018).
- 461 48 Fortunel, C. *et al.* Topography and neighborhood crowding can interact to shape species growth and
462 distribution in a diverse Amazonian forest. *Ecology* **99**, 2272-2283, doi:<https://10.1002/ecy.2441>
463 (2018).
- 464 49 Tiessen, H., Chacon, P. & Cuevas, E. Phosphorus and nitrogen status in soils and vegetation along a
465 toposequence of dystrophic rainforests on the upper Rio Negro. *Oecologia* **99**, 145-150,
466 doi:<https://10.1007/BF00317095> (1994).
- 467 50 Paredes, O. S. L., Norris, D., Oliveira, T. G. d. & Michalski, F. Water availability not fruitfall modulates
468 the dry season distribution of frugivorous terrestrial vertebrates in a lowland Amazon forest. *PLOS*
469 *ONE* **12**, e0174049, doi:<http://dx.doi.org/10.1371/journal.pone.0174049> (2017).
- 470 51 Michalski, L. J., Norris, D., de Oliveira, T. G. & Michalski, F. Ecological relationships of meso-scale
471 distribution in 25 neotropical vertebrate species. *PLoS One* **10**, e0126114,
472 doi:<http://dx.doi.org/10.1371/journal.pone.0126114> (2015).
- 473 52 Mendes Pontes, A. R. Tree reproductive phenology determines the abundance of medium-sized and
474 large mammalian assemblages in the Guyana shield of the Brazilian Amazonia. *Animal Biodiversity*
475 *and Conservation* **43.1**, 9-26, doi:<https://doi.org/10.32800/abc.2020.43.0009> (2020).

- 476 53 Arévalo-Sandi, A., Bobrowiec, P. E. D., Rodriguez Chuma, V. J. U. & Norris, D. Diversity of terrestrial
477 mammal seed dispersers along a lowland Amazon forest regrowth gradient. *PLOS ONE* **13**,
478 e0193752, doi:<https://10.1371/journal.pone.0193752> (2018).
- 479 54 Arita, H. T., Robinson, J. G. & H REDFORD, K. Rarity in Neotropical forest mammals and its ecological
480 correlates. *Conservation Biology* **4**, 181-192, doi:[http://dx.doi.org/10.1111/j.1523-](http://dx.doi.org/10.1111/j.1523-1739.1990.tb00107.x)
481 [1739.1990.tb00107.x](http://dx.doi.org/10.1111/j.1523-1739.1990.tb00107.x) (1990).
- 482 55 Peres, C. A. & Palacios, E. Basin-wide effects of game harvest on vertebrate population densities in
483 Amazonian forests: Implications for animal-mediated seed dispersal. *Biotropica* **39**, 304-315,
484 doi:<http://dx.doi.org/10.1111/j.1744-7429.2007.00272.x> (2007).
- 485 56 Emmons, L. H. & Feer, F. *Neotropical rainforest mammals: A field guide.*, (The University of Chicago
486 Press, 1997).
- 487 57 Michalski, F., Michalski, L. J. & Barnett, A. A. Environmental determinants and use of space by six
488 Neotropical primates in the northern Brazilian Amazon. *Studies on Neotropical Fauna and*
489 *Environment* **52**, 187-197, doi:<https://10.1080/01650521.2017.1335276> (2017).
- 490 58 Laurance, W. F. *et al.* Ecosystem Decay of Amazonian Forest Fragments: a 22-Year Investigation.
491 *Conservation Biology* **16**, 605-618, doi:<http://dx.doi.org/10.1046/j.1523-1739.2002.01025.x> (2002).
- 492 59 Norris, D., Peres, C. A., Michalski, F. & Hinchsliffe, K. Terrestrial mammal responses to edges in
493 Amazonian forest patches: a study based on track stations. *Mammalia* **72**, 15-23,
494 doi:<https://10.1515/mamm.2008.002> (2008).
- 495 60 Martínez-Ramos, M. *et al.* Natural forest regeneration and ecological restoration in human-modified
496 tropical landscapes. *Biotropica* **48**, 745-757, doi:<http://dx.doi.org/10.1111/btp.12382> (2016).
- 497 61 Laurance, W. F., Delamônica, P., Laurance, S. G., Vasconcelos, H. L. & Lovejoy, T. E. Rainforest
498 fragmentation kills big trees. *Nature* **404**, 836-836, doi:<https://10.1038/35009032> (2000).
- 499 62 Tabarelli, M., Lopes, A. V. & Peres, C. A. Edge-effects Drive Tropical Forest Fragments Towards an
500 Early-Successional System. *Biotropica* **40**, 657-661, doi:<https://10.1111/j.1744-7429.2008.00454.x>
501 (2008).
- 502 63 Santos, B. A. *et al.* Drastic erosion in functional attributes of tree assemblages in Atlantic forest
503 fragments of northeastern Brazil. *Biological Conservation* **141**, 249-260,
504 doi:<https://doi.org/10.1016/j.biocon.2007.09.018> (2008).
- 505 64 Arévalo-Sandi, A. R. & Norris, D. Short term patterns of germination in response to litter clearing and
506 exclosure of large terrestrial vertebrates along an Amazon forest regrowth gradient. *Global Ecology*
507 *and Conservation* **13**, e00371, doi:<https://doi.org/10.1016/j.gecco.2017.e00371> (2018).
- 508 65 Kottek, M., Grieser, J., Beck, C., Rudolf, B. & Rubel, F. World map of the Köppen-Geiger climate
509 classification updated. *Meteorologische Zeitschrift* **15**, 259-263, doi:[http://dx.doi.org/10.1127/0941-](http://dx.doi.org/10.1127/0941-2948/2006/0130)
510 [2948/2006/0130](http://dx.doi.org/10.1127/0941-2948/2006/0130) (2006).
- 511 66 ANA. *Sistema de Monitoramento Hidrológico (Hydrological Monitoring System)*. Agência Nacional de
512 Águas[*[nl]*]National Water Agency, Available at <http://www.hidroweb.ana.gov.br>, (2017).
- 513 67 Norris, D. & Michalski, F. Socio-economic and spatial determinants of anthropogenic predation on
514 Yellow-spotted River Turtle, *Podocnemis unifilis* (Testudines: Pelomedusidae), nests in the Brazilian
515 Amazon: Implications for sustainable conservation and management. *Zoologia (Curitiba)* **30**, 482-
516 490, doi:<http://dx.doi.org/10.1590/S1984-46702013000500003> (2013).
- 517 68 Yirdaw, E., Monge Monge, A., Austin, D. & Toure, I. Recovery of floristic diversity, composition and
518 structure of regrowth forests on fallow lands: implications for conservation and restoration of
519 degraded forest lands in Laos. *New Forests* **50**, 1007-1026, doi:<https://10.1007/s11056-019-09711-2>
520 (2019).
- 521 69 McElhinny, C., Gibbons, P., Brack, C. & Bauhus, J. Forest and woodland stand structural complexity:
522 Its definition and measurement. *Forest Ecology and Management* **218**, 1-24,
523 doi:<https://doi.org/10.1016/j.foreco.2005.08.034> (2005).
- 524 70 Sist, P., Mazzei, L., Blanc, L. & Rutishauser, E. Large trees as key elements of carbon storage and
525 dynamics after selective logging in the Eastern Amazon. *Forest Ecology and Management* **318**, 103-
526 109, doi:<https://doi.org/10.1016/j.foreco.2014.01.005> (2014).

- 527 71 Phillips, O. L. *et al.* Species Matter: Wood Density Influences Tropical Forest Biomass at Multiple
528 Scales. *Surveys in Geophysics* **40**, 913-935, doi:<https://10.1007/s10712-019-09540-0> (2019).
- 529 72 Bastin, J.-F. *et al.* Pan-tropical prediction of forest structure from the largest trees. *Global Ecology*
530 *and Biogeography* **27**, 1366-1383, doi:<https://10.1111/geb.12803> (2018).
- 531 73 TEAM Network. 69 (Tropical Ecology, Assessment and Monitoring Network, Center for Applied
532 Biodiversity Science, Conservation International., Arlington, VA, USA., 2011).
- 533 74 Hortal, J., Borges, P. A. & Gaspar, C. Evaluating the performance of species richness estimators:
534 sensitivity to sample grain size. *Journal of Animal Ecology* **75**, 274-287,
535 doi:<http://dx.doi.org/10.1111/j.1365-2656.2006.01048.x> (2006).
- 536 75 Magurran, A. E. & McGill, B. J. in *Biological diversity: frontiers in measurement and assessment* (eds
537 A. E. Magurran & B. J. McGill) Ch. 1, 1-7 (Oxford University Press, 2011).
- 538 76 Laliberté, E. & Legendre, P. A distance-based framework for measuring functional diversity from
539 multiple traits. *Ecology* **91**, 299-305, doi:<http://dx.doi.org/10.1890/08-2244.1> (2010).
- 540 77 FD: measuring functional diversity from multiple traits, and other tools for functional ecology. R
541 package version 1.0-12. (2014).
- 542 78 Burnham, K. P. & Anderson, D. R. *Model selection and multi-model inference: A practical*
543 *information-theoretic approach*. (Springer, 2002).
- 544 79 R: A language and environment for statistical computing. 3.6.3 (R Foundation for Statistical
545 Computing, Vienna, Austria, 2020).
- 546 80 vegan: Community Ecology Package. R package version 2.4-0. [https://CRAN.R-](https://CRAN.R-project.org/package=vegan)
547 [project.org/package=vegan](https://CRAN.R-project.org/package=vegan) (2016).
- 548 81 Wickham, H. *ggplot2: elegant graphics for data analysis*. (Springer, 2009).
- 549 82 MuMIn: multi-model inference. R package version 1.15.6. [https://CRAN.R-](https://CRAN.R-project.org/package=MuumIn)
550 [project.org/package=MuumIn](https://CRAN.R-project.org/package=MuumIn) (2016).
- 551 83 Tweedie: Tweedie exponential family models. R package version 2.2.1. [https://cran.r-](https://cran.r-project.org/web/packages/tweedie)
552 [project.org/web/packages/tweedie](https://cran.r-project.org/web/packages/tweedie) (2014).
- 553

554 **Acknowledgements**

555 The Instituto Chico Mendes de Conservação da Biodiversidade (ICMBio) and the Amapá National
556 Forest staff (Érico Emed Kauano and Sueli Gomes Pontes dos Santos) and the Federal University of
557 Amapá (UNIFAP) provided logistical support. We thank the Brazilian Ministério do Meio Ambiente
558 (“MMA”) for authorizing data collection (SISBIO permits 40355–1, 47859-1 and 47859-2). We also
559 thank the local landowners who gave permission for data collection at their properties. We are deeply
560 indebted to Cremilson and Cledinaldo Alves Marques and family for their dedication, commitment
561 and assistance during the fieldwork.

562

563 **Author Contributions**

564 D.N. conceived of the project; V.J.U.CR and A.A.S collected data. V.J.U.CR, A.A.S and D.N.
565 performed data analysis and interpretation. A.A.S prepared figure 1. D.N. prepared figures 2-5.
566 T.M.F.C and D.N. wrote the main manuscript text. All authors reviewed and revised the manuscript.

567

568 **Competing Interests Statement**

569 The authors declare that they have no competing interests as defined by Nature Research, or other
570 interests that might be perceived to influence the results and/or discussion reported in this paper.

571

Supporting Information

Big trees drive forest structure patterns across a lowland Amazon regrowth gradient

Tassiana Maylla Fontoura Caron¹, Victor Juan Ulises Rodriguez Chuma^{1,2}, Alexander Arévalo-Sandi^{3,4}, Darren Norris^{1,3,5} *

¹ Programa de Pós-Graduação em Biodiversidade Tropical, Universidade Federal do Amapá (UNIFAP), Rod. Juscelino Kubitschek, Km 02, 68902-280, Macapá, AP, Brazil.

² Facultad de Ciencias Forestales, Universidad Nacional de la Amazonia Peruana (UNAP), Pevas 5ta cuadra, Iquitos, Perú.

³ Programa de Pós-graduação em Ecologia, Instituto Nacional de Pesquisas da Amazônia (INPA), Av. André Araújo 2936, Petrópolis 69067-375 Manaus, AM, Brazil.

⁴ Amazonian Mammals Research Group. Av. André Araújo 2936, Petrópolis 69067-375 Manaus, AM, Brazil.

⁵ Coordenação de Ciências Ambientais, Universidade Federal do Amapá (UNIFAP), Rod. Juscelino Kubitschek Km 02, 68902-280 Macapá, AP, Brazil

* Corresponding author at: Coordenação de Ciências Ambientais, Universidade Federal do Amapá (UNIFAP), Macapá, Brazil. Email address: dnorris75@gmail.com (D. Norris).

Supplementary Table S1

Table S1: Generalized Linear Model values.

<i>Predictors</i>	All trees						Large trees					
	Full			Best			Full			Best		
	<i>Est.</i>	<i>CI</i>	<i>p</i>	<i>Est.</i>	<i>CI</i>	<i>p</i>	<i>Est.</i>	<i>CI</i>	<i>p</i>	<i>Est.</i>	<i>CI</i>	<i>p</i>
(Intercept)	0.46	0.40–0.53	<0.001	0.46	0.43–0.50	<0.001	0.66	0.50–0.81	<0.001	0.64	0.55–0.76	<0.001
Last use	0.00	-0.00–0.00	0.933				-0.00	-0.01–0.01	0.896			
Regrowth (late v early)	-0.01	-0.10–0.07	0.826	-0.03	-0.08–0.01	0.176	-0.04	-0.24–0.17	0.710	-0.05	-0.20–0.08	0.441
Regrowth (late v pasture)	-0.00	-0.09–0.08	0.966	-0.03	-0.07–0.02	0.245	-0.08	-0.27–0.10	0.355	-0.07	-0.21–0.06	0.311
Plot (control v regrowth)	0.14	0.02–0.28	0.050	0.08	0.02–0.14	0.022	0.02	-0.40–0.44	0.917	0.53	0.24–1.14	0.016
Interaction (Last use * early)	-0.01	-0.02–0.01	0.467				-0.01	-0.05–0.03	0.525			
Interaction (Last use * pasture)	-0.00	-0.01–0.00	0.386				-0.00	-0.01–0.01	0.986			
Interaction (Last use * Plot)	-0.00	-0.01–0.00	0.280				0.03	-0.04–0.10	0.339			
Interaction (Plot * early)	0.03	-0.11–0.17	0.624	0.09	-0.01–0.19	0.090	46.73		0.999			
Interaction (Plot * pasture)	0.15	0.02–0.29	0.040	0.17	0.06–0.30	0.011	42.03		0.999			
Observations	30			30			30			30		
R ² Nagelkerke	0.83			0.80			0.81			0.74		
AIC	229.40			225.62			140.45			132.02		

1 Supplementary Table S2

2 Generalized Linear Models created to explain forest structure in 30 plots (15 control and 15 regrowth). Model summaries for responses of five structural attributes: (a) The
 3 number of all trees ≥ 10 cm DBH ; (b) Proportion of small (DBH 10 – 20 cm) trees; (c) Proportion of large (DBH >60 cm) trees; (d) Basal area of all trees; (e) Basal area of
 4 large trees. Showing slope estimates (“Est”) for 10 variables in 5 models (time, regrowth class, mammal diversity, hydrography and topography). “NE” denotes cases
 5 when values could not be reliably estimated. The time model included years since the regrowth site was opened and years since last use. The regrowth model included sites
 6 grouped into three regrowth classes (pasture, early-regrowth and late-regrowth). Topography included altitude (masl) and slope. Hydrography was modelled with TWI
 7 (Topographic wetness index), DND (Distance to Network Drainage) calculated from the interaction between HAND (Height above network drainage) and HDND (Horizontal
 8 distance to network drainage). Mammal diversity was obtained from camera-traps and quantified as species richness and Functional Dispersion (FDis).

9 (a) The number of all trees DBH ≥ 10 cm

<i>Predictors</i>	Time			Regrowth class			Mammals			Hydrography			Topography		
	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>
(Intercept)	35.4	15.9–55.0	0.002	35.2	27.3–43.1	<0.001	25.04	5.31–44.77	0.021	20.04	-25.26–65.35	0.395	33.66	-67.46–134.78	0.521
Plot type	-5.2	-32.9–22.4	0.714	6.6	-4.6–17.8	0.257	-8.48	-29.64–12.67	0.440	-95.06	-201.10–10.97	0.093	-402.66	-1045.25–239.93	0.232
Since first open	-0.2	-1.2–0.9	0.764												
Since last use	0.1	-2.7–2.8	0.954												
Plot *First	-0.4	-1.9–1.1	0.573												
Plot *Last	-0.8	-4.7–3.1	0.703												
First *Last	0.0	-0.1–0.1	0.963												
Plot *First *Last	0.1	-0.1–0.3	0.388												
Regrowth (Late v Early)				-1.2	-12.4–10.0	0.835									
Regrowth (Late v Pasture)				-2.4	-13.6–8.8	0.677									
Plot *Early				-14.0	-29.8–1.8	0.095									
Plot*Pasture				-23.2	-39.0–7.4	0.008									
Species.richness							2.9	-7.0–12.8	0.571						
Functional.dispersion							41.7	-78.5–162.0	0.503						
Plot *Species.richness							15.1	1.6–28.5	0.039						
Plot *FDis							-37.0	-174.8–100.75	0.604						

	Time	Regrowth class	Mammals	Hydrography	Topography
Richness *FDis			-11.6 -52.86–29.65 0.587		
Plot *Richness *FDis			-36.2 -88.80–16.5 0.192		
TWI				1.8 -4.2–7.8 0.561	
DND				-6.6 -42.1–29.0 0.722	
Plot * TWI				12.7 -2.0–27.3 0.105	
Plot * DND				-155.8 -285.0–25.3 0.029	
TWI * DND				1.6 -4.6–7.7 0.623	
Plot * TWI* DND				20.4 0.9–39.9 0.053	
Altitude					-0.0 -1.0–1.0 0.980
Slope					-1.7 -16.9–13.8 0.845
Plot *Altitude					3.7 -2.4–9.7 0.250
Plot * Slope					37.9 -70.3–146.2 0.499
Altitude * Slope					0.0 -0.1–0.2 0.813
Plot *Altitude* Slope					-0.34 -1.4–0.7 0.522
Observations	30	30	30	30	30
R ² Nagelkerke	1.000	1.000	1.000	1.000	1.000
Deviance	2206.473	1942.400	1883.533	2095.536	1438.744
AIC	232.075	224.251	227.328	230.527	219.246
log-Likelihood	-107.037	-105.125	-104.664	-106.264	-100.623

(b) Proportion of small (DBH 10 – 20 cm) trees

<i>Predictors</i>	Time			Regrowth class			Mammals			Hydrography			Topography		
	<i>OddsRatios</i>	<i>CI</i>	<i>p</i>	<i>OddsRatios</i>	<i>CI</i>	<i>p</i>	<i>OddsRatios</i>	<i>CI</i>	<i>p</i>	<i>OddsRatios</i>	<i>CI</i>	<i>p</i>	<i>OddsRatios</i>	<i>CI</i>	<i>p</i>
(Intercept)	1.1	0.6–2.2	0.704	1.6	1.2–2.2	0.002	1.1	0.5–2.5	0.778	0.5	0.11–2.7	0.461	5.2	0.1–515.3	0.475
Plot type	38.4	8.6–236.0	<0.001	3.7	2.0–5.1	<0.001	5.7	2.1–15.3	0.001	0.4	0.00–66.8	0.713	126.0	0.0–313.0	0.713
Since first open	1.0	1.0–1.1	0.336												
Since last use	1.0	0.9–1.1	0.834												
Plot *First	0.9	0.8–1.0	0.015												
Plot *Last	0.8	0.7–1.0	0.017												
First *Last	1.0	1.0–1.0	0.841												
Plot *First *Last	1.0	1.0–1.0	0.029												
Regrowth (Late v Early)				0.8	0.5–1.2	0.302									
Regrowth (Late v Pasture)				1.0	0.6–1.5	0.856									
Plot *Early				1.7	0.8–3.6	0.190									
Plot*Pasture				1.0	0.4–2.2	0.938									
Species.richness							1.2	0.8–1.8	0.284						
Functional.dispersion							2.3	0.0–256.3	0.724						
Plot *Species.richness							1.3	0.7–2.5	0.415						
Plot *FDis							0.0	0.0–0.3	0.020						
Richness *FDis							0.5	0.1–2.4	0.389						
Plot *Richness *FDis							1.7	0.1–22.2	0.702						
TWI										1.1	1.0–1.4	0.210			
DND										0.4	0.1–1.5	0.200			
Plot * TWI										1.3	0.6–2.6	0.464			
Plot * DND										0.0	0.0–12.4	0.211			
TWI * DND										1.2	0.9–1.5	0.169			

	Time	Regrowth class	Mammals	Hydrography			Topography		
Plot * TWI* DND				1.5	0.6–3.8	0.411			
Altitude							1.0	1.0–1.0	0.775
Slope							0.6	0.3–1.2	0.153
Plot *Altitude							0.9	0.6–1.5	0.803
Plot * Slope							0.7	0.0–160.2	0.920
Altitude * Slope							1.0	1.0–1.0	0.190
Plot *Altitude* Slope							1.0	0.9–1.1	0.965

Observations	30	30	30	30	30	30	30	30	30
R ² Tjur	0.133	0.111	0.113	0.113	0.188	0.188	0.192	0.192	0.192
Deviance	46.970	60.088	45.259	45.259	39.873	39.873	33.410	33.410	33.410
AIC	158.064	167.181	156.352	156.352	150.967	150.967	144.503	144.503	144.503
log-Likelihood	-71.032	-77.591	-70.176	-70.176	-67.483	-67.483	-64.252	-64.252	-64.252

12

13

14

<i>Predictors</i>	Time			Regrowth class			Mammals			Hydrography			Topography		
	<i>Odds Ratios</i>	<i>CI</i>	<i>p</i>	<i>Odds Ratios</i>	<i>CI</i>	<i>p</i>	<i>Odds Ratios</i>	<i>CI</i>	<i>p</i>	<i>Odds Ratios</i>	<i>CI</i>	<i>p</i>	<i>Odds Ratios</i>	<i>CI</i>	<i>p</i>
(Intercept)	0.0	0.0–0.1	<0.001	0.0	0.0–0.1	<0.001	0.1	0.0–0.5	0.004	0.0	0.0–0.8	0.032	6.43	NE	0.647
Plot type	0.4	0.0–54.2	0.714	0.2	0.0–0.9	0.042	0.0	0.0–0.9	0.040	3.9	NE	0.887	NE	NE	0.830
Since first open	1.1	1.0–1.2	0.071												
Since last use	1.2	1.0–1.5	0.088												
Plot *First	1.0	0.7–1.2	0.685												
Plot *Last	0.8	0.4–1.6	0.492												
First *Last	1.0	1.0–1.0	0.060												
Plot *First *Last	1.0	1.0–1.0	0.505												
Regrowth (Late v Early)				1.5	0.6–3.8	0.417									
Regrowth (Late v Pasture)				2.4	1.0–5.8	0.060									
Plot *Early				0.3	0.0–9.6	0.526									
Plot*Pasture				0.3	0.0–9.7	0.530									
Species.richness							0.9	0.4–1.8	0.653						
Functional.dispersion							0.6	NE	0.921						
Plot *Species.richness							0.1	0.0–5.3	0.271						
Plot *FDis							165.7	NE	0.540						
Richness *FDis							0.9	0.0–17.0	0.936						
Plot *Richness *FDis							1114.3	NE	0.253						
TWI										1.1	0.7–1.6	0.659			
DND										0.6	0.1–4.9	0.593			
Plot * TWI										0.6	0.0–7.5	0.669			
Plot * DND										NE	NE	0.393			

	Time	Regrowth class	Mammals	Hydrography	Topography
TWI * DND				1.1 0.8–1.6 0.562	
Plot * TWI* DND				0.3 0.0–10.4 0.462	
Altitude					0.96 0.89–1.03 0.265
Slope					0.63 0.18–2.16 0.464
Plot *Altitude					0.81 0.13–5.00 0.821
Plot * Slope					0.12 NE 0.897
Altitude * Slope					1.00 0.99–1.01 0.406
Plot *Altitude* Slope					1.02 0.75–1.38 0.906
Observations	30	30	30	30	30
R ² Tjur	0.027	0.008	0.001	0.028	0.026
Deviance	22.439	23.760	22.382	28.416	27.326
AIC	72.554	69.778	72.532	78.313	77.272
log-Likelihood	-28.277	-28.889	-28.266	-31.156	-30.636

16

17

18

<i>Predictors</i>	Time			Regrowth class			Mammals			Hydrography			Topography		
	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>
(Intercept)	0.45	0.39–0.52	<0.001	0.46	0.43–0.50	<0.001	0.43	0.36–0.51	<0.001	0.48	0.31–0.65	<0.001	0.48	0.10–0.90	0.028
Plot type	0.21	0.03–0.43	0.046	0.08	0.02–0.14	0.022	0.25	0.13–0.38	0.001	0.35	-0.59–1.47	0.500	4.85	-0.94–10.40	0.106
Since first open	-0.00	-0.00–0.00	0.956												
Since last use	-0.00	-0.01–0.00	0.341												
Plot *First	-0.00	-0.01–0.01	0.790												
Plot *Last	0.00	-0.02–0.02	0.998												
First *Last	0.00	-0.00–0.00	0.331												
Plot *First *Last	-0.00	-0.00–0.00	0.719												
Regrowth (Late v Early)				-0.03	-0.08–0.01	0.176									
Regrowth (Late v Pasture)				-0.03	-0.07–0.02	0.245									
Plot *Early				0.09	-0.01–0.19	0.090									
Plot*Pasture				0.17	0.06–0.30	0.011									
Species.richness							0.01	-0.03–0.05	0.624						
Functional.dispersion							-0.04	-0.53–0.42	0.855						
Plot *Species.richness							-0.06	-0.15–0.03	0.192						
Plot *FDis							-0.53	-1.28–0.22	0.178						
Richness *FDis							-0.01	-0.17–0.15	0.900						
Plot *Richness *FDis							0.22	-0.11–0.57	0.210						
TWI										-0.01	-0.03–0.02	0.634			
DND										0.05	-0.09–0.17	0.499			
Plot * TWI										-0.03	-0.18–0.10	0.679			
Plot * DND										0.46	-0.74–1.86	0.486			

TWI * DND				-0.01	-0.03	-0.02	0.468
Plot * TWI* DND				-0.06	-0.26	-0.12	0.553
Altitude							0.00 -0.00-0.00 0.994
Slope							-0.02 -0.08-0.03 0.412
Plot *Altitude							-0.04 -0.10-0.01 0.126
Plot * Slope							-0.48 -1.33-0.43 0.297
Altitude * Slope							0.00 -0.00-0.00 0.473
Plot *Altitude* Slope							0.00 -0.00-0.01 0.316

Observations	30	30	30	30	30
R ² Nagelkerke	0.997	0.998	0.997	0.996	0.999
Deviance	74.501	65.927	68.631	82.585	47.959
AIC	233.200	225.615	230.793	236.227	220.376

20

21

<i>Predictors</i>	Time			Regrowth class			Mammals			Hydrography			Topography		
	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>	<i>Est</i>	<i>CI</i>	<i>p</i>
(Intercept)	0.7	0.5–0.9	<0.001	0.7	0.6–0.7	<0.001	0.7	0.4–0.7	<0.001	0.7	0.4–1.0	<0.001	0.5	-0.53–1.71	0.406
Plot type	535.7	NE	0.998	0.3	-0.0–0.6	0.051	62.9	NE	0.999	NE	NE	0.823	-31.5	-204.4–39.4	0.552
Since first open	-0.0	-0.0–0.0	0.303												
Since last use	-0.0	-0.0–0.0	0.186												
Plot *First	-26.7	NE	0.998												
Plot *Last	-27.4	NE	0.998												
First *Last	0.0	-0.0–0.0	0.159												
Plot *First *Last	1.4	-1447.1–1449.9	0.998												
Regrowth (Late v Early)				-0.1	-0.2–0.0	0.189									
Regrowth (Late v Pasture)				-0.1	-0.2–0.0	0.103									
Plot *Early				39.5	NE	0.998									
Plot*Pasture				39.5	NE	0.998									
Species.richness							0.0	-0.1–0.1	0.762						
Functional.dispersion							-0.1	-1.0–0.8	0.811						
Plot *Species.richness							89.8	NE	0.999						
Plot *FDis							-212.3	NE	0.999						
Richness *FDis							0.0	-0.3–0.3	0.859						
Plot *Richness *FDis							-252.1	NE	0.999						
TWI										-0.02	-0.1–0.0	0.392			
DND										0.11	-0.1–0.3	0.273			
Plot * TWI										-63173.6	NE	0.823			
Plot * DND										976655.4	NE	0.823			

	Time	Regrowth class	Mammals	Hydrography	Topography
TWI * DND				-0.02 -0.1-0.0 0.239	
Plot * TWI* DND				-96752.4 NE 0.823	
Altitude					0.0 -0.0-0.0 0.765
Slope					-0.0 -0.2-0.1 0.861
Plot *Altitude					0.3 -0.4-1.9 0.550
Plot * Slope					5.7 -6.2-36.0 0.536
Altitude * Slope					0.0 -0.0-0.0 0.913
Plot *Altitude* Slope					-0.1 -0.3-0.1 0.538
Observations	30	30	30	30	30
R ² Nagelkerke	1.000	1.000	1.000	1.000	1.000
Deviance	140.557	134.847	121.384	119.165	161.409
AIC	138.281	133.726	136.303	136.050	140.145

23

24

25

26

27