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**Reassembly of a tropical rainforest ecosystem: A new chronosequence in the**

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**Ecuadorian Chocó tested with the recovery of tree attributes**

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51 **Abstract**

52 From hunting and foraging to clearing land for agriculture, humans modify forest  
53 biodiversity, landscapes, and climate. Forests constantly undergo disturbance–recovery  
54 dynamics and understanding them is a major objective of ecologists and conservationists.  
55 Chronosequences are a useful tool for understanding global restoration efforts. They  
56 represent a space-for-time substitution approach suited for the quantification of the *resistance*  
57 of ecosystem properties to withstand disturbance and the *resilience* of these properties until  
58 reaching pre-disturbance levels. Here we introduce a newly established chronosequence with  
59 62 plots (50 × 50 m) in active cacao plantations and pastures, early and late regeneration, and  
60 mature old-growth forests, across a 200 km<sup>2</sup> area in the extremely wet Chocó rainforest. Our  
61 chronosequence covers by far the largest total area of plots compared to others in the  
62 Neotropics. Plots ranged from 159–615 masl in a forested landscape with  $74 \pm 2.8$  % forest  
63 cover within a 1-km radius including substantial old-growth forest cover. Land-use legacy  
64 and regeneration time were not confounded by elevation. We tested how six forest structure  
65 variables (maximum tree height and DBH, basal area, number of stems, vertical vegetation  
66 heterogeneity, and light availability), aboveground biomass (AGB), and rarefied tree species  
67 richness change along our chronosequence. Forest structure variables, AGB, and tree species  
68 richness increased with regeneration time and are predicted to reach similar levels to those in  
69 old-growth forests after ca. 30–116, 202, and 108 yrs, respectively. Compared to previous  
70 work in the Neotropics, old-growth forests in Canandé accumulate high AGB that takes one  
71 of the largest time spans reported until total recovery. Our chronosequence comprises one of  
72 the largest tree species pools, covers the largest total area of regenerating and old-growth  
73 forests, and has higher forest cover than other Neotropical chronosequences. Hence, our  
74 chronosequence can be used to determine the time for recovery and stability (resistance and  
75 resilience) of different taxa and ecosystem functions, including species interaction networks.

76 This integrative effort will ultimately help to understand how one of the most diverse forests  
77 on the planet recovers from large-scale disturbances.

78

## 79 **1. Introduction**

80 Tropical forests are highly threatened by deforestation, degradation, and climate change.

81 Every year, some 4.5 million ha of old-growth forests are logged and mostly replaced by

82 pastures and crops (FAO, 2020). This disturbance entails a reduction in species diversity and

83 in the complex interaction networks knitted between them (Poorter et al., 2021a, b). The

84 ongoing 'UN decade of ecosystem restoration' underscores a global urgency to mitigate losses

85 in ecosystems and bend the curve of biodiversity decline, but requires a sound understanding

86 of ecological dynamics (Cooke et al., 2019). Knowledge of the active restoration and natural

87 recovery potential of tropical forests is not only needed for ecological studies, but also for

88 conservation and restoration implementers. Studying chronosequences covering different

89 stages of natural forest regeneration allows for a deeper understanding of succession and the

90 extent to which secondary forests recover (Figure 1; Chazdon et al., 2007). Here, we

91 introduce the theoretical background and spatial design of a well-resolved chronosequence in

92 a Neotropical rainforest that allows us to study the recovery of plant and animal communities

93 and the interactions that maintain this highly diverse ecosystem. We analyse baseline data as

94 a case study to show how six measures of forest structure, aboveground biomass (AGB), and

95 tree species richness change with regeneration time along with other abiotic variables.

96 Tropical forests are key biomes because they harbour over half of the global

97 biodiversity (Pimm & Raven, 2000) and account for one-third of primary productivity on the

98 land surface by sequestering 1.7 Gt of carbon per year (Malhi, 2012; Harris et al., 2021).

99 Tropical ecosystems are inherently dynamic, going through stages of natural and

100 anthropogenic **disturbances** (see **Glossary** for definitions of key terms). Recovery occurs at

101 various spatial and temporal scales and provides services and renewable resources for more  
102 than one billion people (Ghazoul et al., 2015; Lewis et al., 2015; Ghazoul & Chazdon, 2017).  
103 Despite their importance, half of tropical forests had already disappeared by the beginning of  
104 the twentieth century (Wright, 2005), and deforestation continues at high levels today (FAO,  
105 2020).

106 Worldwide demand for palm oil, soybeans, rubber, timber, and other cash crop  
107 products contributed to unprecedented levels of conversion of tropical forests during the last  
108 decades (Watson et al., 2018). Shifting agriculture is one of the primary drivers of forest loss  
109 in tropical regions (Curtis et al., 2018). The disturbance of tropical forests for agricultural  
110 purposes can promote edge effects and the isolation of forest patches (Malhi et al. 2014;  
111 Wanyama et al. 2023). Distance to source populations and the amount of forest left at  
112 different radii are critical factors influencing how animals mediate key ecosystem functions  
113 such as seed dispersal, pollination, host plant defense, predation, parasitism, and  
114 decomposition (Lundberg & Moberg, 2003; Tylianakis & Morris, 2017). Therefore, forest  
115 conversion can lead to profound changes in forest structure and composition, affecting  
116 different dimensions of biodiversity and their functional consequences. These changes may  
117 include a strong decrease in biomass and species diversity losses in plant and animal  
118 communities (Longo et al., 2016; Baccini et al., 2017; Pfeifer et al., 2017; Matricardi et al.,  
119 2020). For instance, only 59 % of plant or animal species occur in disturbed and secondary  
120 habitats altogether when compared to old-growth forests (Alroy, 2017). Thus, for the  
121 conservation and long-term permanence of tropical forests, it is fundamental to understand  
122 how they recover after disturbance, which factors promote recovery, and at which speed they  
123 operate (Chazdon et al., 2009; Gardner et al., 2009).

124

## 125 **1.1 Resistance, resilience, and functional recovery of tropical forests**

126 Measuring forest recovery is challenging, and it often requires a definition of stability in time.  
127 Stability can be defined for either single properties of a system (e.g., population density,  
128 diversity, network metric, or a mean process rate) or for multiple dimensions, including  
129 variation within and among properties (Pimm, 1984; Donohue et al., 2013; Hillebrand et al.,  
130 2018). Several concepts of **stability** exist, including resistance and resilience (Van Meerbeek  
131 et al., 2021). **Resistance** (Figure 1) mirrors the proportion of species, communities,  
132 interactions, and processes that are sustained following **perturbation** and thus persist in an  
133 altered ecosystem – facilitated by properties to resist immediate impact (pulse perturbation)  
134 and/or changes in environmental conditions in continued disturbance regimes (press  
135 perturbation; Pimm, 1984; McCann, 2000; White et al., 2020). **Resilience** can be defined as  
136 the recovery rate of the amount lost to a stable level (often close to an 'equilibrium') found in  
137 an undisturbed reference ecosystem, i.e., the speed of **reassembly** or its completeness after a  
138 certain time span (Holling et al., 1973; Folke et al., 2004; Pimm et al., 2019). **Recovery time**  
139 describes the span from the time disturbance halts until recovery is reached; and can be  
140 accelerated both by high resistance and by high resilience independently (Pimm et al., 2019).

141 Species communities and their ecosystem functions are known to vary in their  
142 resistance and resilience. Some taxa such as trees almost completely vanish in agricultural  
143 landscapes while others, such as ants, become more abundant (Dunn et al., 2004; Rozendaal  
144 et al., 2019; Hoenle et al., 2022). Measurements of recovery time across different systems  
145 and taxa, including changes in trait rules and species interactions in time, thus contribute to  
146 our understanding of variation in resilience. However, these variables have seldom been  
147 measured *in situ*. A recent global meta-analysis (Biggs et al., 2020) found only 15 studies  
148 (none in tropical forests) that tested whether functional redundancy of traits improves  
149 stability in terrestrial ecosystems. Therefore, the resistance and resilience of ecological

150 communities to ecosystem conversion are still poorly assessed, hampering our understanding  
151 of their potential recovery.

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### 153 **1.2 Chronosequences substitute time with space**

154 Most contemporary assessments of tropical forest resilience, in terms of structure and species  
155 diversity, are based on **chronosequences**. Chronosequences allow the study of forest  
156 succession (Chazdon et al., 2007) while examining forest resilience and projecting its  
157 trajectory. Generally, the trajectory of the resilience of diversity (or network complexity) may  
158 either be linear (or saturating) or non-linear (hump-shaped, unimodal, etc.). The  
159 **'intermediate disturbance hypothesis'** (Grime, 1973; Connell, 1978) predicts a hump-  
160 shaped trajectory, with a diversity peak at intermediate stages of recovery after disturbance  
161 and/or at an intermediate disturbance level or frequency. Surprisingly, only a few tests of this  
162 influential hypothesis have proved its validity for tree diversity, for which it was proposed  
163 (Hubbell et al., 1999; Molino & Sabatier, 2001). In addition, non-linearity has not been  
164 explicitly considered and tested in published chronosequences of reassembling animal  
165 communities (e.g., Dunn, 2004; Crouzeilles et al., 2016 and references therein). The recovery  
166 of forests is generally expected – but rarely shown – to be fastest for species diversity,  
167 intermediate for species composition, and slowest for complex species interaction networks  
168 (Acevedo-Charry & Aide, 2019; Poorter et al., 2021a).

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### 170 **1.3 The Chocó rainforest**

171 The Chocó rainforest is a biodiversity hotspot that is highly threatened by deforestation. The  
172 Chocó is located in northwestern South America to the west of the Ecuadorian and  
173 Colombian Andes and in Panamá, occupying only 0.2% of Earth's land surface (Myers et al.,  
174 2000; Pérez-Escobar et al., 2019). However, it is considered the world's ninth most

175 biodiverse hotspot, harboring ~11,000 plant species (3% of the total), 25% of which are  
176 endemic (Gentry, 1982; Christenhusz et al., 2017). The Chocó is a highly threatened habitat  
177 that has been destroyed at very high rates. The particularly high deforestation rate in this area  
178 has been quantified by satellite images available since 1986 (Gonzalez-Jaramillo et al., 2016),  
179 and until 2015, only a small amount of the old-growth forest (<11%) was left (Fagua et al.  
180 2019).

181

## 182 **2. The Reassembly Project**

183 Given the extensive deforestation and natural regeneration in the Chocó, this ecosystem is  
184 ideal to understand resistance, resilience, and recovery of a tropical rainforest. Here we  
185 present the design of a collaborative German-Ecuadorian Research Unit “*Reassembly of*  
186 *species interaction networks*” aiming to unravel to what extent, and how fast a rainforest can  
187 re-establish itself after deforestation. This includes the diverse plant and animal communities,  
188 their complex interaction networks, and relevant ecosystem processes that characterize such  
189 forest ecosystems. The *Reassembly* Research Unit aims to understand the stability, in terms  
190 of resistance and resilience, of a tropical forest ecosystem and many of its components –  
191 based on space-for-time substitution of community and network recovery, underlying trait  
192 rules, and consequences for ecosystem processes (Figure 1). While *Reassembly* primarily  
193 targets fundamental research questions on network reassembly, its results inform nature  
194 conservation and policy makers on the recovery potential of tropical forests, potentially  
195 improving the restoration of tropical forests (Tylianakis et al., 2010; Pocock et al., 2012;  
196 Kaiser-Bunbury et al., 2015, 2017).

197

### 198 **2.1 Study Site**



199 Our study is located within the Ecuadorian Chocó, in the lowland rainforest of the Canandé  
200 watershed, near the town of Hoja Blanca in the Esmeraldas Province. The climate is typical  
201 for equatorial lowland rainforest, with mean annual temperature of ca. 22–23°C and mean  
202 annual precipitation of 3000–6000 mm with a dry season between June–July and October–  
203 November (based on the nearest weather stations 20–50 km away: Santo Domingo: 2800  
204 mm, Luis Vargas Torres: 6200 mm). On-site weather data collection is currently undergoing  
205 with rain gauges and data loggers, and we expect to update this information in the future.  
206 Current land use in the area is relatively recent, occurring within the last 50 yrs according to  
207 local residents. A large portion of the old-growth forest in the area has been logged by timber  
208 industries and transformed by local people into cacao plantations, and less commonly into oil  
209 palms, as well as pastures for cows and horses to sustain their livelihood. Many plantations  
210 and pastures have been abandoned at different times, often to be included in private rainforest  
211 reserves, leaving a mosaic of habitats that differ in their time of recovery. The private  
212 reserves *Canandé* and *Tesoro Escondido* and surroundings (0.5°N 79.2°W, 130–540 masl)  
213 are 14,000 and 1,800 ha in size, respectively, and *Canandé* is still growing. Both harbor  
214 different habitats that range from old-growth forests to active plantations and pastures,  
215 including secondary forests of different regeneration ages. *Fundación Jocotoco* established  
216 *Canandé* Reserve in 2000 and immediately included pastures and plantations in their  
217 protected area. The *Tesoro Escondido* Reserve was established in 2016, with around 90% of  
218 primary forest, the rest being pastures of different ages and old cacao plantations. These  
219 active and old pastures and cacao plantations provide the basis for the chronosequence sites  
220 in this project.

221 We established 62 plots of 50 x 50 m (0.25 ha) along a chronosequence of forest  
222 regeneration in *Canandé* and *Tesoro Escondido* reserves during 2021–2022 (Figure 2a).  
223 Larger plots were difficult to implement owing to the topography of the area, which is hilly

224 with many streams. The large number of plots implemented increases the spatial resolution of  
225 the chronosequence allowing well-replicated spatial data. We included cacao plantations and  
226 pastures as the main land-use types in the area to understand whether different land-use  
227 legacies lead to differences in resilience (Guariguata & Ostertag, 2000; Jakovac et al., 2021).  
228 Six plots are located in active cacao plantations, six in active pastures, 17 in regenerating  
229 cacao plantations, 16 in regenerating pastures, and 17 in old-growth forests (Figure 2a; Table  
230 1).

231 A correct estimation of the regeneration time for the plots in our chronosequence was  
232 fundamental for a proper assessment of forest recovery. Regeneration time was determined as  
233 the number of years that have passed since abandonment until observational data were  
234 finished in 2023. Year of abandonment of the plots was established based on the  
235 documentation by *Fundación Jocotoco* when land purchasing began. Before this, the year  
236 was estimated based on interviews with previous landowners and park rangers. In addition,  
237 we refined these data using satellite-based (see Landcover context section). Regenerating  
238 cacao and pasture plots have been under natural regeneration without major human  
239 disturbance for 1–38 yrs (Table 1; Table S1). The mean ( $\pm$  SE) time of regeneration for cacao  
240 ( $15.8 \pm 2.6$  yrs) and pasture plots ( $15.1 \pm 3$  yrs) is similar ( $t = 0.18$ ,  $p = 0.86$ ). Active cacao  
241 plantations and pastures belong to farms located around regenerating plots and were assigned  
242 a regeneration time of zero because they are currently under use and represent typical  
243 conditions at which regeneration may start. Old-growth forests were not assigned a  
244 regeneration time because there was minimal evidence of human disturbance.

245 Active cacao plantations are monocultures of *Theobroma cacao* L. (varieties  
246 'National' and 'CCN51') usually below 5 m in height, however, trees can reach up to  
247 10 m in older plantations. Cacao trees, planted from seeds in the study area, are spaced out by  
248 2–4 m each, often as part of mixed polyculture including coffee or fruit trees. Plantations

249 generally lack shade trees, and herbicide application is common. Active pastures are mostly  
250 grazed by cattle and occasionally by horses. They consist mainly of pasture grasses such as  
251 *Brachiaria* (Trin.) Griseb. or *Axonopus scoparius* (Flüggé) Kuhlman., but some single large  
252 remnant shade trees or palms have been left. Smaller tree islands can be found along creeks.  
253 Old-growth forests contain large slow-growing trees of potential timber use and showed no  
254 signs of tree harvesting such as tree stumps or skid trails seen elsewhere. Besides, several  
255 species with high wood density values that are usual members of old-growth forests such as  
256 *Licania* spp., *Pouteria* spp., *Eschweilera* spp., or *Lecythis* spp. have been recorded in the old-  
257 growth forest plots.

258

## 259 **2.2 Suitability of the spatial design**

260 We aimed to select plots suitable for statistical analysis of forest recovery based on  
261 regeneration time and land-use legacy (cacao or pasture). Importantly, we selected plots from  
262 potential candidates that increased the spatial independence (e.g., large distances between  
263 plots of the same category and similar age) while avoiding an elevation bias or other  
264 confounding variables like amount of forest around the plots or land-use duration. The  
265 minimum distance between plots of the same type was 184 m between active cacao plots  
266 (Figure 2b). For the other legacies, the minimum distance between two plots was over 250 m.  
267 Mean distances between plots were  $5.2 \pm 1.1$  km for active cacao plantations,  $5.5 \pm 0.7$  km  
268 for regenerating cacao plots,  $4.6 \pm 1.1$  km for active pastures,  $5.3 \pm 0.7$  km for regenerating  
269 pastures, and  $5.5 \pm 0.77$  km for old-growth forest plots.

270 Elevation bias within different reachable parts of the study area was reduced to the  
271 best extent possible within the limitations of the land-use history. Elevation (altitude in m  
272 above sea level) in our study plots ranges between 159–615 masl (Figure 2c; Table 1; Table  
273 S1). Mean elevation in active and regenerating cacao plots ( $350.5 \pm 25.4$  masl) was similar to

274 active and regenerating pasture plots ( $368.2 \pm 27.9$  masl;  $t = -0.47$ ,  $p = 0.64$ ). Mean elevation  
275 in old-growth forests ( $360 \pm 28.8$  masl) was also similar to active and regenerating cacao  
276 plots ( $t = -0.25$ ,  $p = 0.8$ ) and to active and regenerating pastures ( $t = 0.2$ ,  $p = 0.84$ ). There was  
277 no correlation between elevation and (square-root transformed) regeneration time across all  
278 the active and regenerating cacao and pasture plots (Pearson's  $r = 0.17$ ,  $p = 0.26$ ; Figure 2d),  
279 and neither for cacao (Pearson's  $r = 0.22$ ,  $p = 0.31$ ) nor for pasture alone (Pearson's  $r = 0.13$ ,  
280  $p = 0.56$ ). Note that square-root transformation of regeneration time is a common method for  
281 obtaining annual growth rates because it approximately linearizes the slope (Hoenle et al.,  
282 2022).

283 The estimated duration of land use, either as pasture or cacao plantation, of the plots  
284 before abandonment was determined also through interviews, and by examining historic  
285 Landsat data (see Landcover context section below). We acknowledge that the accuracy of  
286 these estimates are low for many plots. Land-use duration of all active plots ranges between  
287 6–23 yrs with a mean of  $15.5 \pm 2.1$  yrs (Table S1). Regenerating plots had a larger range of  
288 duration use (1–30 yrs;  $11.4 \pm 1.2$ ). There were no differences in duration use between cacao  
289 and pasture plots ( $t = -0.59$ – $0.09$ ,  $p = 0.56$ – $0.93$ ).

290 Even though there was no bias in elevation and land-use, analyses along  
291 chronosequences benefit from the consideration of biotic and abiotic environmental variation  
292 as potential explanatory variables besides the time of regeneration, including variables that  
293 are independent of recovery. The lack of such variables has led to criticism of  
294 chronosequence approaches (Johnson & Miyanishi, 2008; Elsy et al., 2023; but see Walker et  
295 al., 2010). Our abiotic baseline data across our chronosequence included temperature,  
296 humidity, soil composition, canopy cover, precipitation, and landscape composition derived  
297 from satellite data (Table S1).

298

299 **2.3 Landcover context**

300 To understand landscape effects across the chronosequence plots, we developed a landcover  
301 map (Figure S1) for active-to-regenerating agriculture and forest within 1 km of each study  
302 plot, a scale relevant for many organisms including invertebrates to smaller vertebrates.  
303 Obtaining precise maps over time is a challenge in lowland rainforests dominated by high  
304 cloud cover. Thus, we used an integrated approach combining different satellite imagery  
305 sources, ground in-situ verification, and local knowledge. Boundaries of current and historic  
306 land use were manually digitized from combined sources in ArcGis (Esri, 2023). As the  
307 primary base layer, we used 0.5 m resolution optical imagery with an accuracy of 5 m from  
308 World Imagery with cloud-free images for 22 Jan 2017 (north, 12 plots) and 13 Nov 2020  
309 (south, 50 plots; Esri 2023). Shrub-successional and canopy-closure phases  $\leq 15$  yrs were  
310 generally visible on optical imagery with differences between early and late regeneration. To  
311 quantify the borders of older regeneration, we used time-series analysis of historic Landsat  
312 images from the European Commission's Joint Research Centre for change in forest cover of  
313 Tropical Moist Forest (JRC TMF) which attempts to quantify the year when forest  
314 degradation occurred since 1990 (Vancutsem et al., 2021), as well as Global Forest Watch  
315 (GFW) analysis of changes in forest height since 2000 (Hansen et al., 2013). Remote sensing  
316 data were ground-truthed in the field by mapping changes in forest structure and composition  
317 along trails (e.g., areas dominated by *Cecropia* spp.). For regenerating cacao and the oldest  
318 regenerating pasture plots which were not visible on optical imagery or historic Landsat, we  
319 quantified approximate disturbance areas in the field. Fieldwork was conducted with original  
320 settlers or their families who had lived on the land and knew the timing of agricultural  
321 abandonment. Local knowledge and structural changes observed on the ground matched  
322 historic Landsat ~2000 when extensive forest clearing occurred after the completion of the  
323 main logging road. Landsat maps also supported field data and local knowledge indicating

324 minimal disturbance at old-growth plots. However, without complete Lidar coverage it was  
325 not possible to identify disturbances > 20-30 yrs including selective logging known to have  
326 occurred in easily accessible areas < 1 km from the main road since the 1960s (e.g. Endesa  
327 Botrosa concession). Thus, current forest cover maps represent areas dominated by mature  
328 trees but do not differentiate changes in forest structure associated with old-growth.

329 We quantified the location of *Reassembly* plots in a mosaic of agricultural land uses  
330 and forest including distance to the forest edge, patch size, and forest cover at three spatial  
331 scales, defined as the percentage classified as forest divided by the total area within three  
332 different radii (1 km, 500 m and 100 m; Table S2). Plots were located in a relatively intact  
333 landscape and overall forest cover averaged  $74 \pm 2.8$  % within a 1 km radius, including  
334 regenerating forest. Most plots were near forest, and on average cacao and pasture plots were  
335 located  $58 \pm 7$  m from the nearest forest edge within  $10.1 \pm 2.1$  ha patches except for  
336 regenerating cacao plots which were in smaller  $2.0 \pm 0.4$  ha cacao plantations (Figure 3a).  
337 Old-growth plots averaged  $388 \pm 72$  m to the nearest forest edge and at the larger spatial  
338 scale (within a 1 km radius), were surrounded by  $87 \pm 4$  % forest, an 18% increase compared  
339 to agricultural plots at  $69 \pm 3$  % ( $F = 10.6, p = 0.001$ ). (Figure S2a, S2b). At smaller spatial  
340 scales (within a 100 m radius), old-growth forest plots were exclusively surrounded by  $98 \pm 2$   
341 % forest whereas forest cover was minimal around cacao and pasture plots (active:  $11 \pm 1$  % ,  
342 regenerating:  $36 \pm 1$  %). There was no significant correlation between elevation and land  
343 cover within 100 m (Pearson's  $r = 0.11, p = 0.37$ ; Figure S2c; Table S3). Because large tracts  
344 of intact forest tend to be retained in inaccessible areas such as ridgetops, there were weak  
345 significant correlations between elevation and landcover at 500 m and 1 km scales (Pearson's  
346  $r = 0.26-0.43, p = 0.0003-0.03$ ) (Figure S2d, S2e). Additionally, land cover at different  
347 scales was significantly correlated with (square-root transformed) regeneration time  
348 (Pearson's  $r = 0.38-0.8, p = 1.583e-15-0.002$ ; Figure S2f, S2g, S2h; Table S2).

349

### 350 **3. Recovery of forest structure, AGB, and tree species richness**

351 Trees are key components of tropical ecosystems because they provide the structure and  
352 microclimate of the forest where animals can coexist and also form the base for key  
353 ecological interactions such as pollination, seed dispersal, and herbivory. We therefore  
354 estimate the time for total recovery, and the stability in terms of resistance and resilience of  
355 forest structure variables, aboveground biomass (AGB), and tree species richness in the 62  
356 plots. We included as forest structure variables maximum tree height, maximum tree diameter  
357 at breast height (DBH), basal area, number of stems, vertical vegetation heterogeneity, and  
358 light availability. For analysis, AGB was estimated using species-specific wood density,  
359 while tree species richness was estimated using rarefaction with a representative coverage  
360 level of 0.75. We run these calculations for wild trees with  $DBH \geq 7.95$  cm.

361

#### 362 **3.1 Methods**

##### 363 3.1.1 Tree survey and identification

364 Our assessment of trees on each plot occurred from February 2022 to July 2023 and  
365 comprised tree species identification and labeling of all individuals  $\geq 25$  cm of circumference  
366 at 1.3 m above the ground ( $\geq 7.95$  cm diameter at breast height, DBH), including palms and  
367 lianas. We also measured the height of each tree using a laser rangefinder/hypsometer  
368 (Forestry Pro II, Nikon). Given that the terrain in the area is irregular and some height  
369 measures could be overestimated, any tree  $> 50$  m was limited to this height. If a tree had  
370 more than one stem  $\geq 25$  cm circumference, we counted and measured up to the four thickest  
371 stems. A botanical collection and silica samples for further genetic and chemical analyses  
372 were obtained from each tree species. Tree identification was performed at the Herbario  
373 Nacional del Ecuador–INABIO using the collections deposited there as references.

374 A total 7921 stems of 7494 cultivated and non-cultivated trees were surveyed,  
375 representing the actual forest structure,. Among these, we identified 859 tree species and  
376 morphospecies, with 39% of them identified up to species level and 25 % to genus level. In  
377 total, 1393 trees were identified to morphospecies level because they did not present leaves  
378 during the survey or because additional work at the herbarium is required. We built a second  
379 dataset of only wild trees, including remnant ones in active agricultural plots. We used this  
380 second dataset for statistical analyses, by removing 779 stems of 639 cultivated trees of 11  
381 species, including cacao, coffee, coconut, and lemon trees common in the area. Finally, we  
382 built a third dataset considering only wild trees with  $DBH \geq 10$  cm to make our results on  
383 AGB and species richness comparable with others. In this dataset, 5393 stems of 5216 trees,  
384 and 733 wild tree species and morphospecies remained.

385

### 386 3.1.2 Forest structure

387 The forest structure variables analysed using the second dataset were maximum tree height  
388 (average of the five tallest stems), maximum tree DBH (average of the five widest stems),  
389 basal area, number of stems, vertical vegetation heterogeneity, and light availability. We  
390 multiplied basal area results by 4 to provide estimated data per 1-ha. Here, light availability is  
391 measured as the total site factor (TSF; i.e., the total solar radiation including direct and  
392 diffuse light, relative to open conditions) at a height of 2 m at three random selected locations  
393 within each plot using hemispherical photography (Solariscope SOL 300B). The instrument  
394 measures the shading of forest canopies relative to open field conditions. Accounting for the  
395 theoretical path of the sun (depending on geographical orientation and latitude), it obtains an  
396 accurate sub-canopy light estimate (Canham et al., 1990). Light measurements were taken at  
397 the four corners of a 10 × 10 m square centered around the plot center. For analysis, we  
398 averaged these measurements per plot. We also reversed TSF values (1-TSF) to understand it



399 as the change of canopy shading due to an increased appearance of leaves and stems along  
400 the chronosequence. To assess vertical vegetation heterogeneity, we estimated vegetation  
401 cover at heights of 0.5, 1, 2, 4, 8, 16, 32, and 64 meters within each plot. To do so, we  
402 established five circular sampling areas with a 20-meter radius per plot, one at the center and  
403 four others at the corners. For each circle, the Shannon-Wiener diversity index was  
404 calculated, and the mean value of these indices from all five circles was then averaged to  
405 determine the overall vertical vegetation heterogeneity for each plot (Bibby et al. 2000).

406

### 407 3.1.3 Aboveground biomass (AGB)

408 For analysis, AGB was computed for the second dataset with wild trees with  $DBH \geq 7.95$  cm.  
409 We calculated AGB for each individual using species-specific wood density along with DBH  
410 and height data with the Chave et al. (2014) equation #4 in the package BIOMASS (Réjou-  
411 Mechain et al., 2017) and then added per plot. If the species identity was not available for  
412 different reasons, BIOMASS obtains a wood density value for the plant genus or the family,  
413 because wood density is phylogenetically conserved (Chave et al., 2005). For those stems  
414 with no identification to family level, BIOMASS obtains a mean wood density value per plot  
415 and uses it in the AGB calculations. We also provide as supplementary material a measure of  
416 AGB calculated with a different equation from Chave et al. (2005) using DBH, height, and  
417 assuming a constant wood density of  $0.5 \text{ g/cm}^3$  (Table S4). In Table S4 we also included  
418 AGB calculated for the first (all trees with  $DBH \geq 7.95$  cm) and for the third datasets (wild  
419 trees with  $DBH \geq 10$  cm). In addition, we calculated aboveground carbon (AGC) stocks per  
420 plot for the three datasets in BIOMASS without accounting for potential errors in height and  
421 DBH because that data was provided and not estimated (Table S4). We multiplied AGB and  
422 AGC results by 4 to provide data for 1-ha.

423

424 3.1.3 Tree species richness

425 As with AGB, we used estimations of tree species richness for wild trees with DBH  $\geq$  7.95  
426 cm for analyses. We estimated coverage-based rarefied species richness using the package  
427 iNEXT (Hsieh et al., 2016) based on a representative coverage level of 0.75. In addition, we  
428 provide raw species richness obtained with vegan (Oksanen et al., 2022) as supplementary  
429 material (Table S4).

430

431 3.1.4 Statistical analysis

432 First, we performed linear models to determine how and by which factors the six measures of  
433 forest structure (maximum tree height, maximum tree DBH, basal area, number of stems,  
434 vertical vegetation heterogeneity, and reversed TSF), AGB, and tree species richness change  
435 along the chronosequence. Elevation, the percentage of forest at 100 m around the plots, and  
436 land-use legacy (cacao or pasture) interacting with (square-root transformed) regeneration  
437 time, and duration use were set as fixed effects. The interaction between legacy and (square-  
438 root transformed) regeneration time was included because we wanted to determine whether  
439 former cacao and pasture plots can regenerate at a similar time. Old-growth forests were not  
440 included in the models because of their unknown age and extent of their last natural or human  
441 disturbance (if any). We performed an ANOVA test on the resulting models to look for type  
442 II errors. For model validation, we first checked visually that the residuals of the models  
443 follow a normal distribution. To test the lack of variance homogeneity, we performed a  
444 Bartlett test and examined the Pearson correlation between fitted and residual values of the  
445 models. We determined the lack of spatial autocorrelation of the model residuals with a  
446 Moran's I test from the package ape (Paradis & Schliep et al., 2019).

447 Second, we estimated the time for total recovery of the response variables for all  
448 active and regenerating plots and for each legacy type (cacao and pasture) based on a

449 simplified linear model without elevation, the interaction between legacy and (square-root  
450 transformed) regeneration time, the percentage of forest 100 m around the plots, and duration  
451 of use. We defined the following linear trend for the response variable ( $y$ ) over time ( $t$ ):

$$452 \quad y = a + b*t. \quad (1)$$

454 Here,  $a$  is the starting point (intercept) at time 0 and  $b$  is the slope of (square-root  
455 transformed) regeneration time. We used the median value of each response variable in old-  
456 growth forests ( $OG_{\text{median}}$ ) as reference to reveal the estimated recovery time as

$$457 \quad T_{\text{full}} = (OG_{\text{median}} - a) / b. \quad (2)$$

459 We then back-transformed ( $\wedge^2$ ) the time for total recovery because we used square-root  
460 transformed regeneration time in the models.

461 Third, we calculated the resistance and resilience of the six variables of forest  
462 structure, AGB, and tree species richness across the regeneration gradient. Resistance and  
463 resilience were determined for cacao and pasture plots together and separately. Resistance  
464 was calculated as

$$465 \quad \text{Resistance} = (AC_{\text{median}}) / (OG_{\text{median}}), \quad (3)$$

467 Where  $AC_{\text{median}}$  represents the median value of the response variables in active plots when  
468 regeneration time ( $t$ ) is zero. Resilience was determined as a percentage/year using the  
469 following formula:

$$470 \quad \text{Resilience} = ((OG_{\text{median}} - AC_{\text{median}}) / OG_{\text{median}}) / T_{\text{full}}. \quad (4)$$

472

473 **3.2 Results**

474 Mean values of the forest structure variables in old-growth forests were always higher  
475 compared to those in regenerating and active plots (Table 1; Table S4). For instance,  
476 maximum tree height in old-growth forests was  $35.54 \pm 0.92$  m while in regenerating plots it  
477 was  $22.96 \pm 1.47$  m and in active plots it was  $12.96 \pm 1.74$  m. Maximum tree DBH was  $93.99$   
478  $\pm 4.9$  cm in old-growth forests,  $42.53 \pm 2.76$  cm for regenerating plots and  $20.65 \pm 3.3$  cm for  
479 active plots. Basal area peaked at old-growth forests with a mean of  $39.54 \pm 2.58$  m<sup>2</sup>/ha,  
480 while it was  $13.31 \pm 1.43$  m<sup>2</sup>/ha and  $1.18 \pm 0.29$  m<sup>2</sup>/ha in regenerating and active plots,  
481 respectively. Number of stems varied the most with an average of  $178.35 \pm 7.32$  stems in old-  
482 growth forests,  $112.7 \pm 9.71$  stems in regenerating plots and  $8 \pm 1.47$  stems in active plots.  
483 Vertical vegetation heterogeneity in old-growth forests had a mean value of  $1.86 \pm 0.01$ ,  
484 while in regenerating and active plots it was  $1.57 \pm 0.04$  and  $0.77 \pm 0.15$ , respectively.  
485 Reversed TSF was  $82.29 \pm 1.38$  % in old-growth forests,  $76.47 \pm 1.56$  % in regenerating  
486 plots, and  $57.49 \pm 6.31$  % in active plots.

487         The six forest structure variables increased significantly with (square-root  
488 transformed) regeneration time (Figure 3; Table S5). Vertical vegetation heterogeneity was  
489 influenced by land-use legacy and by its interaction with (square-root transformed)  
490 regeneration time (Figure 3e; Table S5). Neither elevation nor the percentage of forest at 100  
491 m around the plots nor duration use influenced the change of the response variables along the  
492 chronosequence. Assumptions of the models including normal distribution of residuals and  
493 variance homogeneity were met except for vertical vegetation heterogeneity (Figure S3,  
494 Supplementary Results). Spatial autocorrelation was determined only for maximum tree DBH  
495 and vertical vegetation heterogeneity (Figure S3, Supplementary Results). Number of stems,  
496 vertical vegetation heterogeneity and reversed TSF were the parameters that recovered fastest  
497 at around 30 yrs, whereas maximum tree DBH was slowest to recover at 120 yrs (Table 2).

498 Total above-ground biomass (AGB) calculated with species-specific wood density for  
499 all trees, including cultivated ones, was 2181.4 t within all the plots of the chronosequence,  
500 while AGB only of non-cultivated wild trees was 2163.55 t. This means that the AGB from  
501 cultivated trees in our chronosequence, mostly cacao trees, represents 17.85 t (0.8 %). When  
502 considering only wild trees with DBH  $\geq$  10 cm, total AGB was 2124.15 t. AGB was highly  
503 correlated with basal area (Pearson's  $r = 0.97$ ,  $p < 0.001$ ). Within old-growth forests, the  
504 mean of AGB was  $358.54 \pm 30.72$  t/ha (Table 1; Table S4) while in active plots it was  $7.03 \pm$   
505  $2.34$  t/ha and in regenerating plots it was  $74.99 \pm 9.9$  t/ha. AGB increased significantly with  
506 (square-root transformed) regeneration time based on the ANOVA test ( $p < 0.001$ ; Table 3).  
507 A normal distribution of residuals, variance homogeneity, and lack of spatial autocorrelation  
508 were determined (Figure S4, Supplementary Results). The model estimated a complete  
509 recovery of AGB at 202.6 yrs for all active and regenerating plots, 199.6 yrs when  
510 regenerating from cacao, and 208.6 yrs when regenerating from pasture (Table 2; Figure 4a).

511 Rarefied tree species richness reached an average of  $71.8 \pm 6.45$  in old-growth forests  
512 (Table 1; Table S4),  $22.5 \pm 3.13$  in regenerating plots and  $4.88 \pm 1.2$  in active plots. Tree  
513 species richness also increased significantly only with (square-root transformed) regeneration  
514 time based on the ANOVA test ( $p < 0.001$ ; Table 3). Assumptions of the model were also  
515 met (Figure S4, Supplementary Results). Complete recovery of tree species richness was  
516 estimated at 107.9 yrs for all active and regenerating plots, 104.9 yrs for cacao, and 112 yrs  
517 for pasture (Table 2; Figure 4b).

518 Resistance was generally slightly higher for pasture plots, potentially because of the  
519 remnant trees left standing in pastures, except for vertical vegetation heterogeneity and  
520 reversed TSF (Table 2; Figure 5). The highest resistance was detected for reversed TSF  
521 (77.89 %) while the lowest was for AGB (0.92 %). Resilience, on the other hand, was slightly  
522 higher for cacao plots except for vertical vegetation heterogeneity and reversed TSF (Table 2;

523 Figure 5). The highest resilience was recorded for the number of stems (3.1 %/y) while the  
524 lowest was determined for AGB (0.49 %/y) too.

525

### 526 **3.3 Discussion**

527 In an era of ongoing deforestation in the tropics with huge losses of biodiversity and  
528 ecosystem functioning, an important challenge today is to protect and restore such vulnerable  
529 ecosystems. This includes the urgent need to better understand drivers of successful natural  
530 recovery and community assembly of ecosystems, and natural disturbance–recovery  
531 dynamics. Chronosequences like the one presented in our study in the Ecuadorian Chocó  
532 have proven particularly useful to understand forest recovery over several decades, and to  
533 reveal successional patterns of various ecosystem attributes. One important goal in this  
534 endeavor is to distinguish resistance of properties to withstand disturbance and their  
535 resilience – both describing different dimensions of ecosystem stability. Here we presented  
536 the recovery of forest structure, aboveground biomass (AGB), and tree species richness, and  
537 showed how our spatial design of 62 plots was suitable to understand recovery trends over  
538 time without confounding effects of elevation, land-use duration, landscape composition, and  
539 spatial proximity. For example, we estimated a relatively slow recovery of tree richness (101  
540 yrs), both due to low resistance (7%) and resilience (0.9 %/y). Studies addressing recovery  
541 within the Reassembly project will estimate resistance and resilience of various taxa,  
542 interactions and ecosystem components, providing a broader quantitative assessment of this  
543 complex rainforest system. For ants, Hoenle et al. (2022) determined that species richness  
544 recovered much faster (7-8 yrs), with high resistance (72%) and resilience (10.6 %/y), than  
545 community composition (21-29 yrs) across many of the plots of our chronosequence. This  
546 finding is consistent with other chronosequences elsewhere, where species diversity of  
547 animals generally recovered faster than their species composition (Curran et al., 2014). We

548 predict that species interaction networks may recover even slower than species diversity and  
549 composition, particularly meta-community networks that have more complex features  
550 (Moreno-Mateos et al., 2020).

551 Overall, forest structure variables are less affected by deforestation than AGB and tree  
552 species richness based on the values of resistance. The forest structure variables directly  
553 related with tree size or density (maximum tree height and DBH, basal area, and number of  
554 stems) show relatively low resistance compared to variables indirectly related with trees such  
555 as vertical vegetation heterogeneity and reversed total site factor (TSF). These last two  
556 variables behave heterogeneously after disturbance among the active plots because not all  
557 plots have standing trees which influence vegetation complexity and the amount of light  
558 present. Particularly, reversed TSF showed the highest resistance suggesting that after forest  
559 disturbance there are still areas where light does not penetrate because of the remnant trees  
560 that could have been left standing or were planted during cultivation. Reversed TSF also  
561 shows relatively high resilience, which added to high resistance allows one of the shortest  
562 times for full recovery analyzed here. The low resistance and resilience for AGB results in  
563 the longest time for full recovery. The number of stems showed the highest resilience and  
564 also one of the shortest times for total recovery despite its low resistance, suggesting that  
565 recovery time is not limited by low resistance when resilience is high. Resistance and  
566 resilience for the variables analyzed here were in general similar for cacao and pasture plots,  
567 showing that land-use legacy does not influence recovery dynamics in our chronosequence as  
568 also revealed by our linear models.

569 The total recovery of the variables analyzed occurred at different times. The number  
570 of stems, vertical vegetation heterogeneity, and reversed TSF are the fastest variables to  
571 recover just after 30–32 yrs. On the other hand, maximum tree DBH and basal area recover  
572 after 120 and 104 yrs, respectively. Basal area recovers at a similar time compared to a

573 tropical forest in Panamá whose recovery was estimated after 90 yrs (Elsy et al., 2023). This  
574 shows that despite regenerating forests having a similar number of stems compared to old-  
575 growth forests in a relatively short period of time, the full recovery of their structure still  
576 takes 4x that time. The full recovery of AGB estimated at 202 yrs takes almost 2x the time  
577 required for the recovery of forest structure potentially because of differences in species  
578 composition and wood density. In old-growth forests it is common to find tree species with  
579 high wood density compared to species found during early forest regeneration stages (Poorter  
580 et al., 2021a, b). This increases AGB in old-growth forests compared to regenerating forests  
581 with trees of the same size. Based on the variables analyzed here, the recovery of forest in  
582 Canandé would take at least 200 years. Nevertheless, the recovery of species composition and  
583 interaction networks is expected to take even a longer time (Crouzeilles et al., 2017; Moreno-  
584 Mateos et al., 2020).

585

#### 586 **4. Generalization to other chronosequences**

587 A summary of many chronosequences from Neotropical forest sites has been published by  
588 the “2ndFOR” network (AGB from 41 sites: Poorter et al., 2016, tree species richness from  
589 56 sites: Rozendaal et al., 2019, several forest attributes from 77 sites including 8 from West  
590 Africa: Poorter et al., 2021b). This allowed us to place our study design, environmental  
591 conditions, tree diversity, and recovery into a broader context.

592 Our chronosequence covers by far the largest total area of plots (4.25 ha of old-  
593 growth forests, 8.25 ha of regenerating forests, 3 ha of active agriculture) among all studies  
594 reviewed by Rozendaal et al. (2019) and thus involves a particularly large number of tree  
595 morphospecies (Figure 6a), even if the latter may become slightly lower once all trees have  
596 been identified to species. Hence, the chronosequence in the Chocó is particularly well  
597 resolved due to a large number of relatively large plots; only five other studies reviewed by



598 Rozendaal et al. (2019) had more plots than ours but these were much smaller. Note,  
599 however, that most tree inventories used a lower threshold of DBH (5 cm) than ours (7.95  
600 cm). Our study area is particularly wet (precipitation > 3000 mm) and its forest cover is  
601 relatively high (74%) compared to the majority of other chronosequences (Figure 6b). Our  
602 chronosequence is representative regarding its regeneration time range (0–38 yrs) compared  
603 to other neotropical studies (oldest plots: median 40 yrs, range: 15–100 yrs; Rozendaal et al.,  
604 2019).

605 Aboveground biomass (AGB) in old-growth forests in our study ( $358 \pm 31$  t/ha,  $n =$   
606 16 plots,  $DBH \geq 7.95$  cm) is considerably higher than in many other Neotropical old-growth  
607 forest sites (mean  $255 \pm 117$  t/ha,  $n = 19$  studies mostly with  $DBH \geq 5$  cm; Poorter et al.,  
608 2016). The same was observed for our AGB estimates for larger trees ( $DBH \geq 10$  cm; mean  
609 AGB  $354 \pm 30$  t/ha compared to other Neotropical forests calculated in the same way (mean  
610 of means  $301 \pm 31$  t/ha,  $n = 14$  sites, 146 1-ha plots, 333 0.1-ha plots; Poorter et al., 2015).  
611 The high precipitation in our chronosequence compared to others (Figure 6b) could explain  
612 the high AGB found because this variable is mostly driven by rainfall (Poorter et al., 2015).  
613 Beside AGB, aboveground carbon (AGC) for larger trees ( $DBH \geq 10$  cm) was  $167 \pm 14$  t/ha.  
614 These were comparable to the values reported for a 1-ha permanent plot within the study area  
615 (AGB: 307.2 t/ha, AGC: 153.6 t/ha) for a total of 441 individual trees ( $DBH \geq 10$  cm) from  
616 101 species (Lozano et al., 2023). AGB resilience in our study was similar to those in other  
617 Neotropical forests because we predicted an AGB of  $123 \pm 0.6$  t/ha for 20 yrs secondary  
618 forests while the estimated average of 20 yrs plots across other neotropical chronosequences  
619 was 122 t/ha (range: 20–225 t/ha; Poorter et al., 2016).

620 The recovery of AGB in Canandé occurred over a much longer time compared to  
621 other chronosequences in other Neotropical forests. For instance, in a review of Neotropical  
622 secondary forests, it was determined that a median time of 66 yrs was enough to recover 90

623 % of AGB levels of old-growth forests (Poorter et al., 2016). In our case, it would take 164  
624 years to recover 90 % of AGB, which is 2.5x the time estimated for the region. A meta-  
625 analysis across pairs of tropical secondary vs. old-growth forests revealed that complete  
626 recovery of AGB was reached after an average of 80 yrs (Martin et al., 2013), while we  
627 estimate this process would again take 2.5x longer in our system. These comparisons of AGB  
628 with other chronosequences may suggest that old-growth forests in Canandé have been  
629 mostly undisturbed during the last decades, accumulating biomass through a relatively long  
630 period of time.

631         The original datasets of tree diversities per plot and recovery times per study from the  
632 “2ndFOR” compilations were not available, hence we only compare the averages reported.  
633 Moreover, estimates for total recovery time differ strongly between statistical methods. The  
634 mean species richness per 25 tree stems (rarefaction method) was  $17.9 \pm 1.8$  across the 16  
635 old-growth plots in our study, higher than the  $14.7 \pm 4.5$  found across 43 studies summarized  
636 in Rozendaal et al. (2019) using this rarefaction method. This difference in tree species  
637 richness could be explained by the larger forest area studied here (Figure 6a), allowing the  
638 finding of rare species that increase this diversity measure. Recovery time estimates in the  
639 meta-analysis of Poorter et al. (2021b) were obtained by fitting a saturating function, and  
640 total recovery was then defined for the predicted function reaching 90% of the old-growth  
641 level. Applying this method to our data results in a predicted recovery time of 55 years, well  
642 comparable with the mean of 54 years (11–228 yrs) reported by Poorter et al. (2021b).  
643 Overall, this shows that our highly resolved chronosequence was typical for other neotropical  
644 sites, being relatively rich in tree species and with a high biomass density, and with  
645 comparable recovery rates.

646         These comparisons against other chronosequences show that the natural regeneration  
647 of the forests in Canandé would take a longer time than for others in the region, taking even

648 centuries. Although ecological restoration success is higher for natural regeneration than for  
649 active restoration (Crouzeilles et al., 2017), some assistance to natural regeneration could  
650 accelerate the recovery of variables such as AGB. For instance, trees from species with high  
651 wood density typical from old-growth forests (Poorter et al., 2021a, b) could be planted in  
652 regenerating forests that already have a structure that allows the survival of these planted  
653 trees. In that way, the recovery of AGB could happen over a shorter time span. On the other  
654 hand, variables that recover faster such as those related with forest structure or species  
655 diversity could be dependent only on the propagules arriving from the relatively high forested  
656 landscape around the plots (Figure 6b).

657

## 658 **5. Outlook and future directions**

659 Although deforestation by far outweighs the areas of recovering tropical forests, increasing  
660 efforts are being taken to protect and restore rainforest ecosystems from agricultural land or  
661 heavily exploited forests. The recovery of secondary forests is expected to play an increasing  
662 role in sustainable timber production, water protection, and biodiversity management during  
663 the next decades (Poorter et al., 2016; Chazdon & Guariguata, 2016). However, these efforts  
664 need to be implemented based on scientific grounds if the aim is to increase their chances of  
665 success (Brancalion et al., 2019). The implementation of chronosequences to study the  
666 recovery of ecosystem processes maintained by different taxonomic groups can promote an  
667 urgently needed understanding of the recovery potential of tropical forests, important  
668 components involved and thus contribute to successful restoration and conservation. Some  
669 first studies analyzing the reassembly of different animal communities and ecosystem pools  
670 such as deadwood in Canandé have been recently published (Hoenle et al., 2022; 2023;  
671 Müller et al., 2023; Falconí-López et al., in press).

672 Moving forward, we will test predictions of diversity and network reassembly and  
673 empirically assess the dynamics of species interaction networks along our chronosequence.  
674 Targeted networks and ecological processes include pollination, seed dispersal, seedling  
675 establishment, herbivory, predation, and dead wood decomposition, and thus multiple  
676 important mutualistic and antagonistic relationships between animals and plants. Different  
677 types of traits may enhance population and network resistance and resilience, respectively,  
678 and determine the rate of ecosystem processes to which these networks contribute  
679 (Sakschewski et al., 2016). In general, we expect that the diversity of plant and animal  
680 communities, species community composition, and species interaction networks resemble  
681 those of mature old-growth forests with time. However, different reassembly trajectories of  
682 species communities and interaction networks should be determined by evolutionary history  
683 and background taxonomic and functional diversity (Farneda et al., 2021; Molina-Venegas et  
684 al., 2018; Mahayani et al., 2020).

685 Whereas chronosequences have revealed many insights in changes in tree biomass,  
686 species' abundances, diversity, and composition with time, data on species interactions or  
687 chronosequences looking at multiple taxa analyzed simultaneously remain scarce. Only a few  
688 studies have investigated the reassembly of interaction networks along chronosequences (see  
689 Staab et al., 2016; Redmond et al., 2019; Montoya-Pfeiffer et al., 2020), and thus a  
690 mechanistic understanding of the network dynamics is lacking. The degree of specialization  
691 of interaction networks varies across interaction types. Hence, mutualistic networks  
692 comprising pollinators, and primary and secondary seed dispersers are expected to be  
693 essential for forest recovery, particularly for maintaining the diversity of trees (Jordano,  
694 2000; Ollerton et al., 2011). In turn, antagonistic interactions such as seed predation, seedling  
695 herbivory, or their reduction by predators, parasitoids, and pathogens, can also play an  
696 important role in the reassembly of tree communities (Bagchi et al., 2014). Our functional

697 understanding of how communities and ecosystems reassemble, and why some systems  
698 return to a natural state while others do not, strongly depends on understanding these  
699 interaction networks.

700

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716

## 717 **Conflict of Interest Statement**

718 The authors declare no conflicts of interest.

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1091 **Boxes**

1092 **Glossary - Concepts for temporal trends, disturbance, and stability**

1093 **Chronosequence:** A series of plots in a region, with comparable site attributes, that primarily  
1094 differ in time span after a specific event (Bakker et al. 1996, Walker et al. 2010, Dunn 2004).  
1095 The **chronosequence** in our study represents forest recovery (natural succession) from  
1096 **disturbance** (agricultural use).

1097 **Disassembly:** An altered community composition and network characteristics immediately  
1098 following a perturbation, typically affecting a fraction of the original set of species,  
1099 relationships, or interactions.

1100 **Disturbance:** A relatively discrete event that disrupts components of an ecosystem, e.g. its  
1101 community or population structure, resource or substrate availability and environmental  
1102 conditions, and is more broadly defined as perturbation (Pickett & White 1985). May involve  
1103 human influences as well as natural causes of fluctuations, e.g. forest gaps created by tree  
1104 falls and small-scale mosaics of succession.

1105 **Intermediate disturbance hypothesis (IDH):** A hypothesis that states that repeated  
1106 disturbances at an intermediate level promote species coexistence via maintaining a non-  
1107 equilibrium state, thus providing a release from competitive dominance of particular species  
1108 (Connell 1978). Explicitly focusing on tropical rainforest trees, Connell suggested that “the  
1109 highest diversity (...) should occur either at an intermediate stage in succession after a large  
1110 disturbance or with smaller disturbances that are neither very frequent nor infrequent”. The  
1111 IDH thus makes an explicit prediction for changes in tree richness (upper graph) and  
1112 composition (lower graph) with forest recovery.

1113 **Reassembly:** The process of re-organization of community composition and network  
1114 characteristics in an ecosystem that recovers from a perturbation.

1115 **Recovery:** The process of regaining original ecosystem properties and any measure after  
1116 disturbance until levels of the reference state are reached (**recovery time**). Recovery includes  
1117 reassembly that is more narrowly defined for more complex community composition or  
1118 network properties.

1119 **Resilience:** One aspect of temporal **stability**: the "rebounding" of a system following a  
1120 disturbance. While the concept of resilience may involve different capacities of  
1121 reorganization to retain an equilibrium level or non-equilibrium states of a function, structure,  
1122 identity or relationship (McCann 2000), its most applicable metric is simple: **recovery rate**  
1123 (Pimm 1984). It could be defined as a rate of the initial loss (i.e. as the slope of the trajectory  
1124 from the disturbed level to the reference level, or percent of losses regained in a given time)  
1125 avoids that resilience and resistance are confounded (Pimm 1984, Pimm et al. 2019, White et  
1126 al. 2020). Our project thus aims at quantifying resilience, and at studying *mechanisms*  
1127 contributing to its variation.

1128 **Resistance:** Another aspect of **stability** of an ecosystem: the act of opposing and  
1129 withstanding a disturbance or perturbation. Resistance is measured as the proportion of an  
1130 initial property that remains following the disturbance (or while a disturbed phase continues),  
1131 applicable to ecosystem characteristics such as species composition, network properties or  
1132 ecosystem processes (Pimm 1984, Pimm et al. 2019, White et al. 2020).

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1138 **Tables**

1139 **Table 1.** Mean values and standard errors of plot characteristics and response variables.

1140 Maximum tree height was calculated as the average of the five tallest stems per plot while

1141 maximum tree DBH was calculated as the average of the five widest stems per plot.

1142 Maximum tree height, maximum tree DBH, basal area, and number of stems were estimated

1143 for wild trees with DBH  $\geq 7.95$  cm. Reversed TSF was calculated as 1-TSF.

1144 Aboveground biomass (AGB) was estimated using species-specific wood density along with

1145 DBH and height data, and rarefied tree species richness was estimated with a coverage of

1146 0.75 for wild trees with DBH  $\geq 7.95$  cm.

<b>Variable</b>	<b>Cacao</b>	<b>Regenerating cacao</b>	<b>Pasture</b>	<b>Regeneration pasture</b>	<b>Old-growth</b>
Number of plots	6	17	6	16	17
Elevation (masl)	299.5 $\pm$ 40.46	368.52 $\pm$ 30.63	365.16 $\pm$ 47.93	369.31 $\pm$ 34.82	360.05 $\pm$ 28.84
Regeneration time (yrs)	0	15.82 $\pm$ 2.61	0	15.12 $\pm$ 2.97	NA
Maximum tree height (m)	10.44 $\pm$ 2.39	24.03 $\pm$ 1.3	15.48 $\pm$ 2.24	21.82 $\pm$ 2.72	35.54 $\pm$ 0.92
Maximum tree DBH (cm)	19.2 $\pm$ 5.53	46.13 $\pm$ 3.34	22.1 $\pm$ 4.05	38.7 $\pm$ 4.36	93.99 $\pm$ 4.9
Basal area (m <sup>2</sup> /ha)	1.12 $\pm$ 0.46	14.29 $\pm$ 1.87	1.24 $\pm$ 0.39	12.27 $\pm$ 2.21	39.54 $\pm$ 2.58
Number of stems	6.67 $\pm$ 1.67	119.47 $\pm$ 10.41	9.33 $\pm$ 2.46	105.5 $\pm$ 16.9	178.35 $\pm$ 7.32
Vertical vegetation heterogeneity	1.07 $\pm$ 0.14	1.64 $\pm$ 0.04	0.47 $\pm$ 0.2	1.49 $\pm$ 0.08	1.86 $\pm$ 0.01
Reversed TSF (%)	61.79 $\pm$ 8.71	78.16 $\pm$ 1.43	53.18 $\pm$ 9.6	74.68 $\pm$ 2.82	82.29 $\pm$ 1.38
AGB	6.35 $\pm$ 3.55	82.11 $\pm$ 13.62	7.7 $\pm$ 3.35	67.42 $\pm$ 14.63	358.54 $\pm$ 30.72
Tree species richness	4.5 $\pm$ 1.99	24.25 $\pm$ 4.4	5.27 $\pm$ 1.56	20.63 $\pm$ 4.54	71.8 $\pm$ 6.45

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1155 **Table 2.** Summary of total recovery (in years), resistance (%), and resilience (%/y) of forest  
 1156 structure variables, aboveground biomass (AGB), and tree species richness. maximum tree  
 1157 height was calculated as the average of the five tallest stems per plot while maximum tree  
 1158 DBH was calculated as the average of the five widest stems per plot. Forest structure  
 1159 variables, AGB estimated using species-specific wood density along with DBH and height  
 1160 data, and rarefied tree species richness estimated with a coverage of 0.75 were calculated for  
 1161 wild trees with DBH  $\geq$  7.95 cm. Reversed total site factor was calculated as 1-TSF.

Variable	Total recovery (y)	Total recovery cacao (y)	Total recovery pasture (y)	Resistance (%)	Resistance cacao (%)	Resistance pasture (%)	Resilience (%/y)	Resilience cacao (%/y)	Resilience pasture (%/y)
Maximum tree height	57.76	58.12	57.31	35.54	30.69	41.47	1.12	1.19	1.02
Maximum tree DBH	119.5	114.41	127.21	22.39	20.23	22.6	0.65	0.7	0.61
Basal area	103.88	101.84	106.82	2.52	2.4	2.93	0.94	0.96	0.91
Number of stems	30.82	29.83	32	4.44	4.44	4.73	3.1	3.2	2.98
Vertical vegetation heterogeneity	29.89	23.35	38.57	35.15	65.95	22.74	2.17	1.46	2
Reversed TSF	32.56	26.38	40.72	77.89	77.89	62.78	0.68	0.84	0.91
AGB	202.27	199.36	208.1	0.92	0.86	1.52	0.49	0.5	0.48
Tree species richness	101.2	99.09	104.24	6.27	4.3	7.04	0.93	0.97	0.89

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1172 **Table 3.** Results of linear models. Aboveground biomass (AGB) was estimated using  
1173 species-specific wood density along with DBH and height data, and rarefied tree species  
1174 richness was estimated with a coverage of 0.75 for wild trees with  $DBH \geq 7.95$  cm. *P* values  
1175 in bold < 0.05.

<b>AGB - ANOVA type II error</b>				
<b>Variable</b>	<b>Sum Sq</b>	<b>Df</b>	<b>F value</b>	<b><i>p</i></b>
Elevation (masl)	0	1	0.0003	0.987
Forest 100 m around	471	1	0.4947	0.4861
Land-use legacy	506	1	0.5318	0.4703
sqrt(Regeneration time)	75727	1	79.5234	7.46E-11
Duration use	256	1	0.2684	0.6074
Land-use legacy (pasture) : sqrt(Regeneration time)	426	1	0.4469	0.5079
Residuals	36186	38		

<b>Tree species richness - ANOVA type II error</b>				
<b>Variable</b>	<b>Sum Sq</b>	<b>Df</b>	<b>F value</b>	<b><i>p</i></b>
Elevation (masl)	398.4	1	3.1849	0.08231
Forest 100 m around	175.2	1	1.4007	0.24396
Land-use legacy	29	1	0.2318	0.63294
sqrt(Regeneration time)	3507.9	1	28.0424	5.26E-06
Duration use	177.3	1	1.4175	0.24119
Land-use legacy (pasture) : sqrt(Regeneration time)	46.3	1	0.37	0.54664
Residuals	4753.5	38		

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1188 **Figure captions**

1189 **Figure 1.** Conceptual framework of ecosystem dis- and reassembly, showing the proposed  
1190 dynamics of diversity, interactions, and related ecosystem processes. The first step describes  
1191 the reduction or disassembly of old-growth forests' properties, following disturbance by  
1192 deforestation and subsequent land use as pasture or plantation. The amount of diversity,  
1193 interactions, and processes that withstands disturbance and remains present during  
1194 agricultural use defines the resistance of that property. The recovery of forests or reassembly  
1195 of communities and networks starts as soon as the utilization as pasture or plantation stops.  
1196 We generally predict a continuous increase of diversity, network complexity, and process  
1197 rates with forest recovery until some saturation level (if there is no support for a hump-  
1198 shaped trajectory for alpha diversity as predicted by the intermediate disturbance hypothesis).  
1199 The level reached after a given time of recovery may be lower (incomplete reassembly) or  
1200 similar to the average level of old-growth forests. The speed and extent of recovery represent  
1201 the resilience of the forest ecosystem.

1202

1203 **Figure 2.** Plot spatial characteristics. a) Location of the 62 study plots at the Canandé and  
1204 Tesoro Escondido reserves in the lowland rainforests of northwestern Ecuador. b) Distances  
1205 between plots within each legacy type. The Y axis is log-transformed to facilitate the  
1206 visualization of low values. c) Elevational distribution of each legacy type. d) Elevation is not  
1207 correlated with square-root transformed regeneration time in all active and regenerating cacao  
1208 and pasture plots. Old-growth forest (OG) plots were not included in the analyses because  
1209 their time without human disturbance is unknown. Dark circles in all boxplots represent mean  
1210 values.

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1212 **Figure 3.** Recovery of forest structure. a) Maximum tree height, b) maximum tree DBH, c)  
1213 basal area, d) number of stems, e) vertical vegetation heterogeneity, and f) reversed total site  
1214 factor (TSF) increase significantly with regeneration time in cacao and pasture plots.

1215 Maximum tree height was calculated as the average of the five tallest stems per plot while  
1216 maximum tree DBH was calculated as the average of the five widest stems per plot.

1217 Maximum tree height, maximum tree DBH, basal area, and number of stems were estimated  
1218 for wild trees with  $DBH \geq 7.95$  cm. Reversed TSF was calculated as  $1-TSF$ . Old-growth  
1219 forest (OG) plots were not included in the analyses because their time without human  
1220 intervention is unknown and were used only as a reference.

1221

1222 **Figure 4.** Recovery of aboveground biomass (AGB) and tree species richness. a) AGB  
1223 estimated using species-specific wood density along with DBH and height data, and b)  
1224 rarefied tree species richness estimated with a coverage of 0.75 increase significantly with  
1225 regeneration time in cacao and pasture plots. AGB and rarefied tree species richness were  
1226 calculated for wild trees with  $DBH \geq 7.95$  cm. Old-growth forest (OG) plots were not  
1227 included in the analyses because their time without human intervention is unknown and were  
1228 used only as a reference. Boxplots show the median value of the Y axis.

1229

1230 **Figure 5.** Recovery dynamics of forest structure, aboveground biomass (AGB), and tree  
1231 species richness. Cacao and pasture plots (active and regenerating) are plotted separately.  
1232 maximum tree height was calculated as the average of the five tallest stems per plot while  
1233 maximum tree DBH was calculated as the average of the five widest stems per plot. Forest  
1234 structure variables, AGB estimated using species-specific wood density along with DBH and  
1235 height data, and rarefied tree species richness estimated with a coverage of 0.75 were

1236 calculated for wild trees with  $DBH \geq 7.95$  cm. Reversed total site factor (TSF) was calculated  
1237 as 1-TSF.

1238

1239 **Figure 6.** Context of our chronosequence (green) in 56 other forest chronosequences (red)

1240 from various dry to moist forests in the Neotropics (Rozendaal et al., 2019). Total area

1241 covered from all old-growth versus regenerating forest plots shown, as well as rainfall and

1242 forest cover in the landscape. Dots scaled by total number of tree morphospecies.

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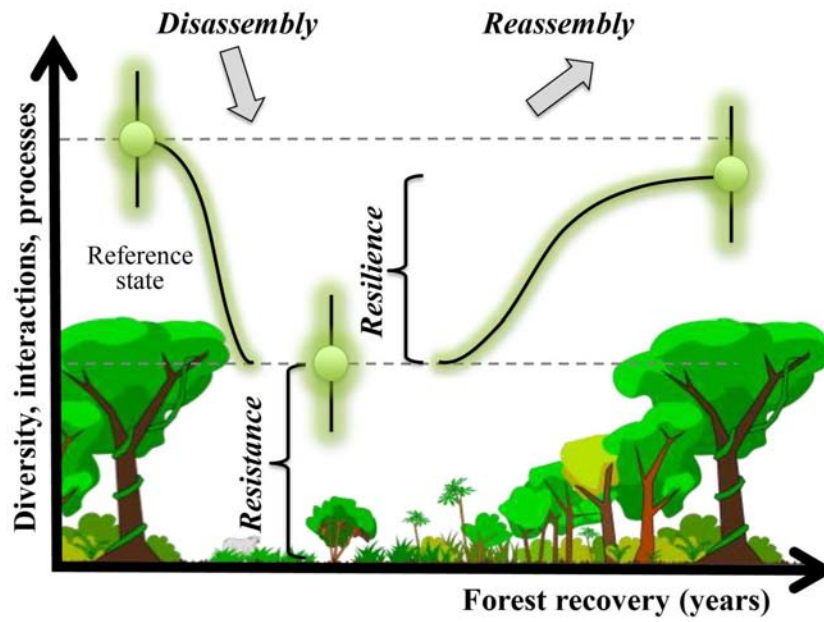
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1261 **Figures**

1262 **Figure 1**



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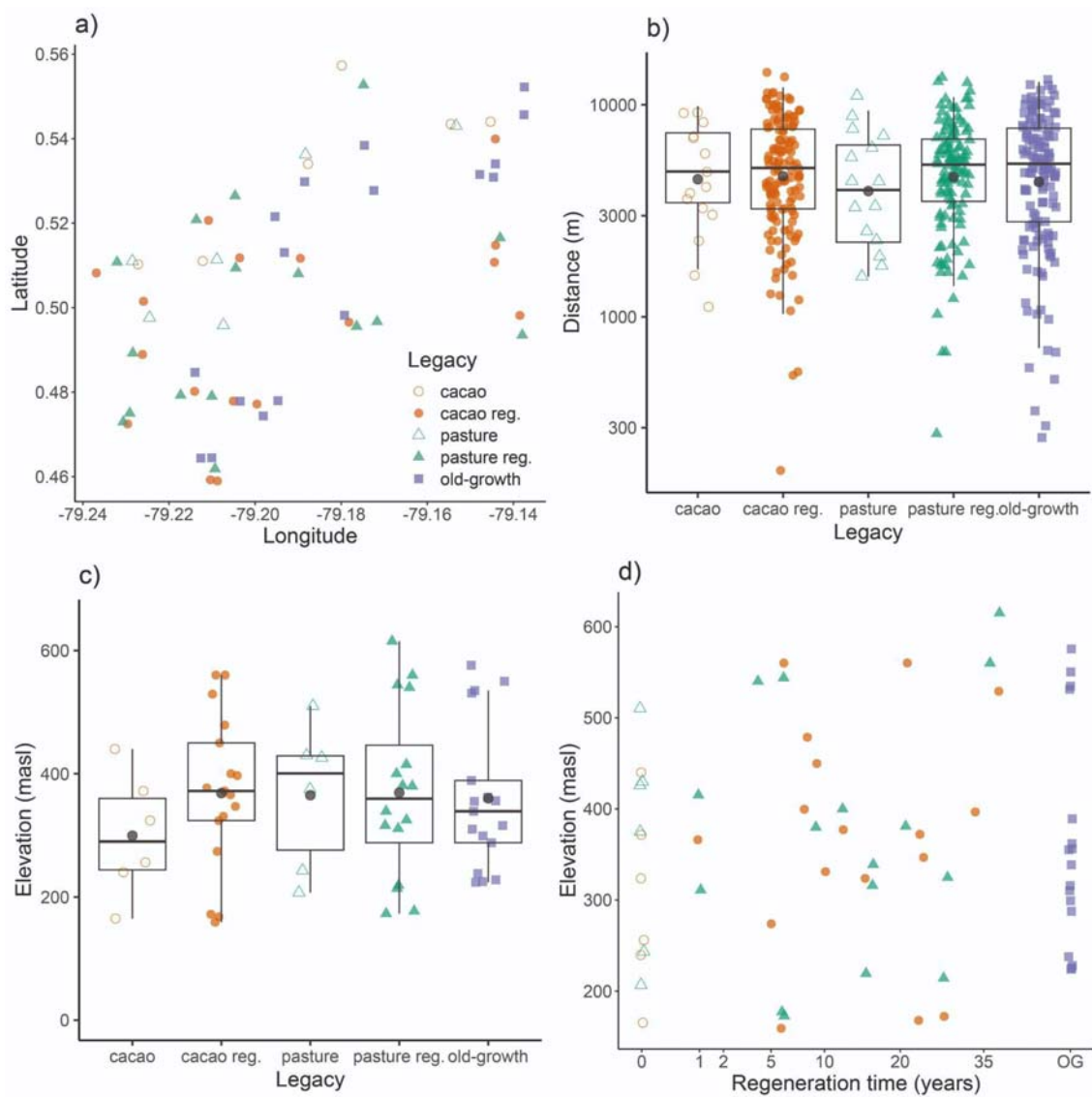
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1277 **Figure 2**



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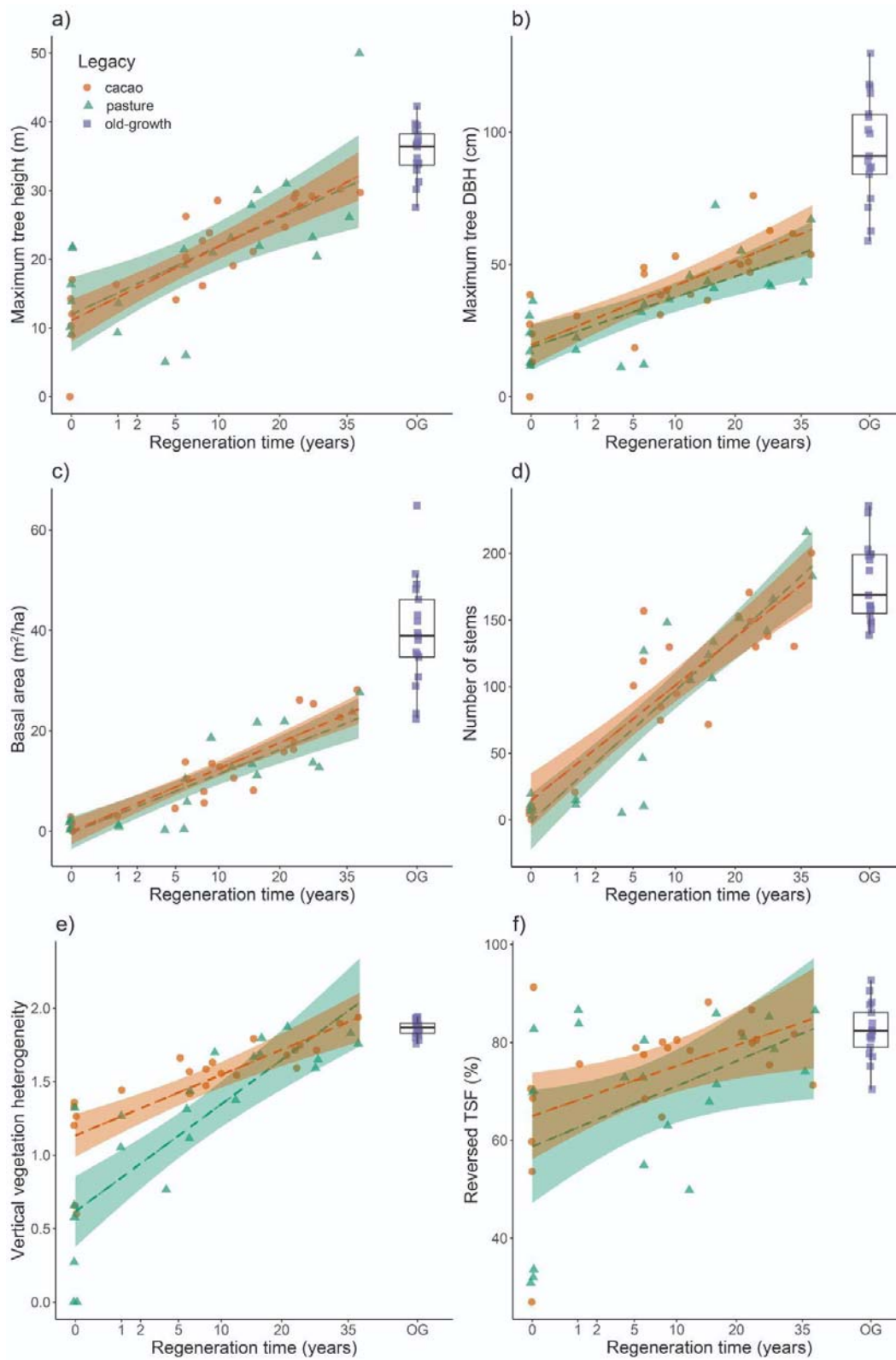
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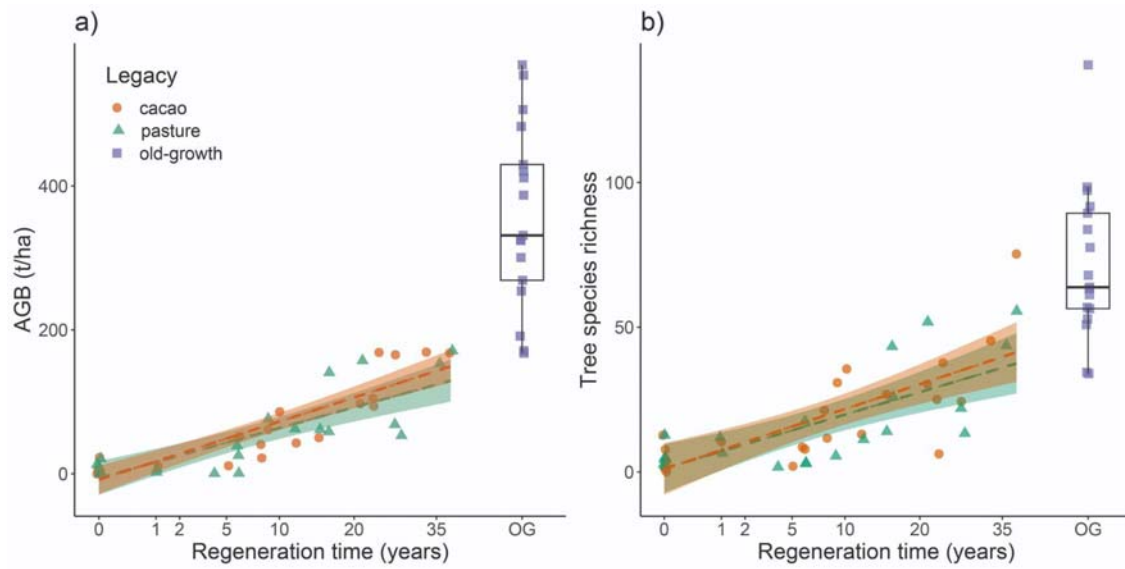
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1289 **Figure 4**



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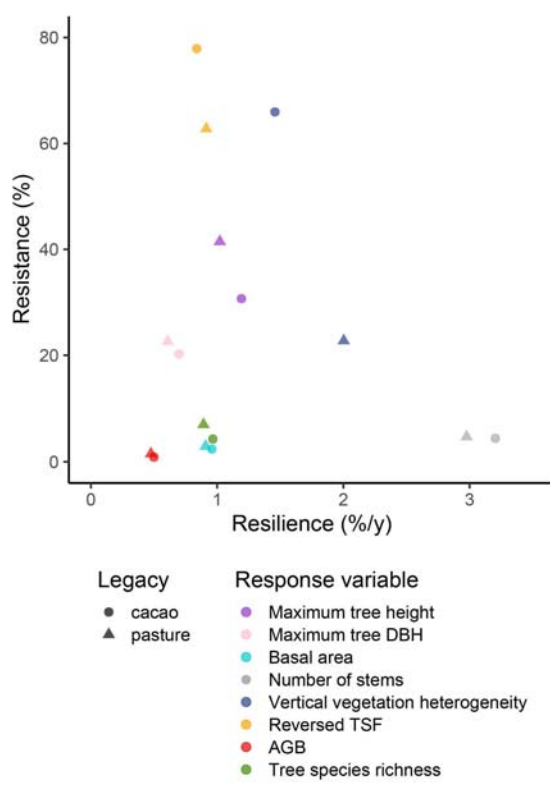
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1306 **Figure 5**



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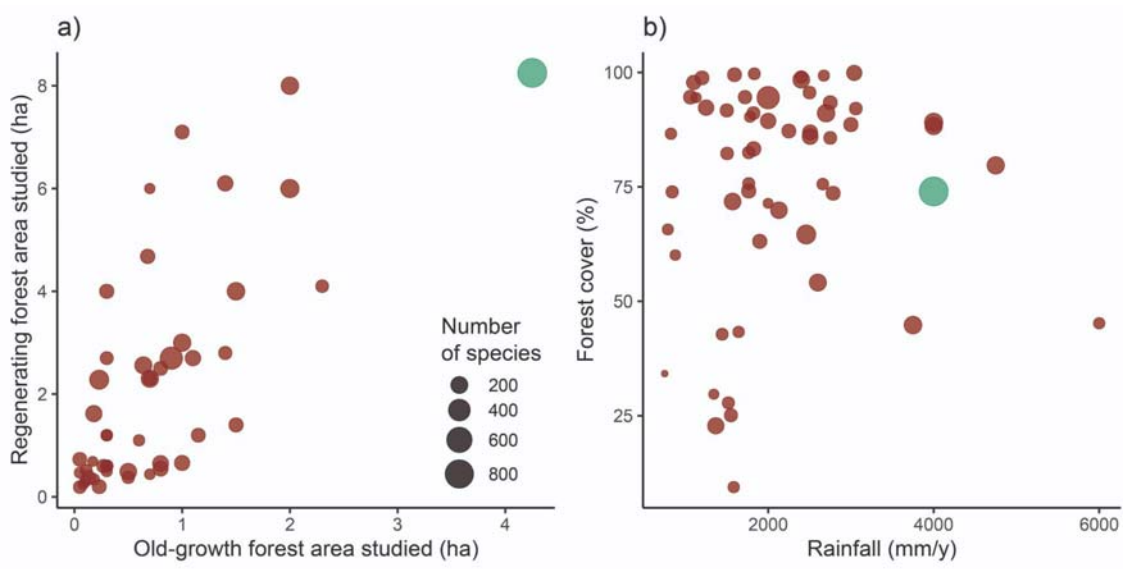
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1320 **Figure 6**



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