

# 1 **Recovery of forest structural complexity during secondary succession in the tropics**

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12

## 13 **Abstract**

14 Forest structural complexity is an essential determinant of forest ecosystem functions and  
15 biodiversity. The natural dynamics of structural complexity of tropical forests remain largely  
16 unexplored, especially for naturally regenerating forest during secondary succession. Better  
17 understanding the trajectories of forest structural complexity recovery is crucial to inform the  
18 development of forest landscape restoration strategies and to predict the reassembly of ecological  
19 networks during secondary succession. Here, we investigate the recovery of forest structural  
20 complexity during secondary succession following land use abandonment in a human-modified  
21 landscape in Ecuador. We employ a terrestrial laser scanning-based index of forest structural  
22 complexity to quantify three-dimensional vegetation structure of agricultural lands (cacao  
23 plantations and pastures), naturally regenerating sites, and primary old-growth forests along a  
24 chrono-sequence of secondary succession. We find that sites recovering after land use  
25 abandonment attain levels of forest structural complexity comparable to old-growth forest within  
26 40 years. Changes in forest structural complexity along the successional gradient follow a saturating  
27 pattern, with rapid increases in the first years and only minor potential for further increases after  
28 40 years. Increasing tree species diversity during secondary succession is identified as a major driver  
29 of the recovery of forest structural complexity. Using a structural equation modelling approach, we  
30 find that effects of tree species diversity on forest structural complexity are mediated by its effects  
31 on vertical stratification, as the effective number of canopy layers increases with increasing recovery  
32 age. Our results suggest that passive restoration is a suitable strategy to restore forest structural  
33 complexity in human-modified landscapes.

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35 **Keywords:** 3D forest structure; diversity-complexity relationships; forest landscape restoration

## 36 Introduction

37 Secondary tropical rainforests recovering from anthropogenic disturbance comprise approximately  
38 half of the world's total tropical rainforest area (FAO 2015). Secondary rainforests have been shown  
39 to play a crucial role in restoring ecosystem functioning and significantly contribute to the  
40 sequestration of atmospheric CO<sub>2</sub> and biodiversity conservation (Hérault & Piponiot, 2018; Poorter  
41 et al., 2016). Predicting recovery trajectories of recovering tropical rainforests is subject to a high  
42 degree of uncertainty, as the recovery of several forest attributes, such as species diversity, species  
43 composition, forest structure and ecosystem functioning, depends on past land use type and  
44 intensity (Hordijk et al., 2024; Zemp et al., 2023), as well as several environmental factors, such as  
45 proximity to remnant forests, soil properties, seed rain and species interactions (Chazdon, 2003;  
46 Norden et al., 2015; Poorter, Rozendaal, et al., 2021). A better understanding of recovery trajectories  
47 is crucial for actively or passively restoring biodiversity, forest structure, and ecosystem functioning  
48 in formerly forested, human-modified tropical landscapes (Aleeje et al., 2025).

49

50 Different forest attributes recover at different paces. While soil attributes or species diversity may  
51 recover within a few decades (Aleeje et al., 2025; Poorter, Craven, et al., 2021; van der Sande et al.,  
52 2024), other attributes, such as biomass and species composition may take more than a century to  
53 attain values comparable to old-growth forests (Escobar et al., 2025; Lennox et al., 2018; Poorter,  
54 Craven, et al., 2021). The recovery trajectory of different forest attributes, such as biomass, soil  
55 attributes, taxonomic or functional diversity have received some attention (Poorter, Rozendaal, et  
56 al., 2021; van der Sande et al., 2022; Veldkamp et al., 2020). However, we are still far from being  
57 able to generalize how tropical forest ecosystems recover structurally and functionally. In this light,  
58 understanding how forest structural complexity recovers during secondary succession is of  
59 particular interest, because it strongly influences ecosystem functions, such as productivity and  
60 carbon sequestration (Gough et al., 2019; Gough, Atkins, et al., 2021; Ray et al., 2023) as well as  
61 biodiversity, by defining habitat quality and niche space (Knuff et al., 2020; Wildermuth et al., 2024).

62

63 Forest structural complexity is defined by vegetation density and the heterogeneity of plant material  
64 distribution in three-dimensional space (*sensu* Atkins et al., 2018; de Conto et al., 2024; Ehbrecht  
65 et al., 2021; Gough et al., 2019). The structural complexity of forests is determined by several biotic  
66 and abiotic factors and their interactions. While the global variation of forest structural complexity  
67 is largely determined by climate, at local-scale, variations are predominantly controlled by site  
68 conditions as well as the natural and anthropogenic disturbance regime (Ehbrecht et al., 2021). The  
69 level of forest structural complexity that can be attained under specific site conditions and  
70 disturbance regimes is then determined by the successional stage and the associated tree species

71 composition (Juchheim et al., 2020). It depends on the tree species' complementarity in crown  
72 architectures, tree size diversity and functional diversity (Ehbrecht et al., 2021). In this context,  
73 several studies have pointed to a positive relationship between tree species diversity and forest  
74 structural complexity (e.g. Coverdale & Davies, 2023; Perles-Garcia et al., 2021; Zemp et al., 2023).  
75 Forest stands composed of tree species with varying functional and morphological traits may  
76 promote the formation of multiple canopy-layers as well as a more efficient occupation of available  
77 space and thereby show higher levels of forest structural complexity. While the formation of  
78 multiple canopy layers results in a more heterogeneous distribution of biomass in three-  
79 dimensional space, an efficient occupation of available canopy space determines overall vegetation  
80 density. The interplay of biomass distribution and vegetation density then determines the level of  
81 forest structural complexity. Describing forest structural complexity as a function of plant material  
82 distribution and vegetation density in a three-dimensional space is underpinned by several studies  
83 that used three-dimensional laser scanning data to quantify complexity using different indices, such  
84 as the stand structural complexity index (SSCI, Ehbrecht et al., 2017), box-counting-dimension (Db,  
85 (Seidel, 2018), canopy rugosity (Atkins et al., 2018) and space-borne metrics like the Waveform  
86 Structural Complexity Index (WSCI, de Conto et al., 2024).

87

88 In anthropogenically altered forest ecosystems, both biomass distribution and vegetation density  
89 and thus structural complexity, are strongly determined by land use type and management practices  
90 (Ehbrecht et al., 2017; Soto et al., 2024; Willim et al., 2022). Land use or management effects on  
91 structural complexity have been widely studied, mostly by comparing different land use systems or  
92 different types of management, with and without undisturbed forests as reference systems. The  
93 impacts of forest management or other land-uses on formerly forested lands on forest structural  
94 complexity have been investigated for boreal coniferous forests (Aalto et al., 2023; Pöldveer et al.,  
95 2021), temperate broadleaved forests (e.g. Asbeck & Frey, 2021; Soto et al., 2024; Willim et al.,  
96 2022), temperate coniferous forests (e.g. Juchheim et al., 2020; Seidel et al., 2016), tropical moist  
97 broadleaf forests (Ehbrecht et al., 2021; Milodowski et al., 2021; Rosen et al., 2024), or tropical  
98 agroforestry systems (Kikuchi et al., 2024; Steinfeld et al., 2024). Both positive and negative impacts  
99 of management on structural complexity have been reported. Whether management has positive  
100 or negative effects on forest structural complexity depends on the management practices and their  
101 intensity (i.e. enrichment planting vs. logging, selective logging or partial harvests vs. (partial) clear-  
102 cutting), as well as the selected reference a certain land use system or forest type is compared to.  
103 For instance, enriching monocultural cash crop plantations, such as oil palm plantations, increases  
104 complexity by planting trees (Kikuchi et al. 2024, Zemp et al., 2019). Management practices focusing  
105 on e.g. timber harvesting often result in lower levels of forest structural complexity compared to

106 unmanaged or late-successional stages of primary forests, but may also increase forest structural  
107 complexity in the mid- to long-term by accelerating the establishment of natural regeneration (e.g.  
108 Soto et al. 2024, Willim et al. 2022).

109

110 Compared to the well-studied effects of land use, management or disturbances on forest structural  
111 complexity, much less is known about the natural dynamics of structural complexity in general, and  
112 in particular about the development of structural complexity during succession. For instance, Gough  
113 et al. (2021) showed that the development of forest structural complexity followed a sigmoidal  
114 trend over a period of 200 years in temperate forest ecosystems, with a short period of slow  
115 recovery of forest structural complexity in the first years, followed by longer periods of constant  
116 increase in complexity over time and, reaching a saturation after around 150 years. For human-  
117 modified tropical landscapes, an in-depth understanding of forest structural complexity dynamics  
118 and its drivers is lacking, as most knowledge on the structural recovery of tropical rainforests is  
119 based on single structural attributes, such as above-ground biomass, canopy height or basal area  
120 (e.g Aryal et al., 2024; Nytch et al., 2023; Oberleitner et al., 2021).

121

122 Here, we investigate the recovery of forest structural complexity during secondary succession and  
123 test whether changes in forest structural complexity along the successional trajectory are driven by  
124 changes in tree diversity. To quantify forest structural complexity, we employ a widely used,  
125 terrestrial laser scanning-based index of forest structural complexity (SSCI) and conduct tree  
126 inventories along a chrono-sequence of tropical forest succession after land use abandonment in a  
127 Neotropical lowland rainforest landscape in Ecuador. Additionally, we sample forest structural  
128 complexity and tree diversity in old-growth forests as reference sites. We first hypothesize that  
129 forest structural complexity increases over time, but follows a saturating pattern. Thus, we expect  
130 an asymptotic relationship between SSCI and recovery age. Second, we hypothesize that the  
131 recovery of forest structural complexity is driven by increases in tree diversity along the successional  
132 gradient. To test how recovery age and tree diversity drive forest structural complexity, we employ  
133 a structural equation modelling approach.

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## 141 **Methods**

### 142 *Study site*

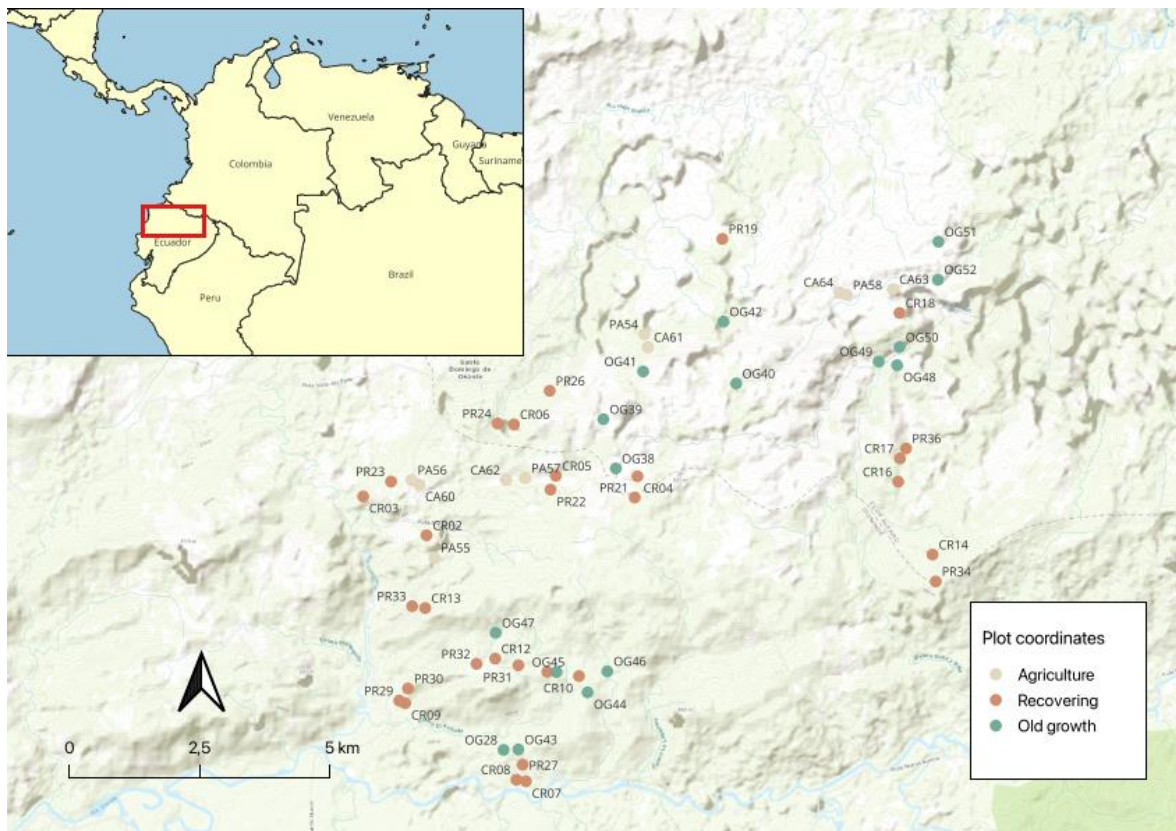
143 This study was conducted in the neotropical lowland rainforest of the Canandé watershed of the  
144 Chocó-Region, located in the province of Esmeraldas in North-Western Ecuador (see Fig. 1). Mean  
145 annual precipitation was reported to range between 4000-5000 mm per year, while mean annual  
146 temperature ranges between 21° to 25°C (Escobar et al., 2025). The topography of this region is  
147 shaped by the foothills of the Andes. The elevation of the plots ranges from 159 m to 615 m above  
148 sea level, with vegetation types randomly distributed across this range. The research area is mainly  
149 marked by old-growth forest, recovering secondary forest, cacao plantations and pastures. Active  
150 land use in this area is mainly characterized by smallholders managing cacao plantations and cattle  
151 pastures. Actively managed cacao plantations are characterized by small stands of cacao trees  
152 systematically arranged, with rarely any other vegetation present. Actively grazed cattle or horse  
153 pastures are mainly characterized by tall grasses up to 1.5 meters, some thickets and occasionally  
154 large remnant trees from former forest cover. This study was conducted within the frame of the  
155 *Reassembly* project, which is an interdisciplinary research unit to study the resistance, resilience  
156 and functional recovery of rainforest ecosystems. For further details, see (Escobar et al., 2025).

157 A total of 56 experimental plots have been established in a forest reserve owned by the *Fundación*  
158 *Jocotoco* and the *Tesoro Escondido* Reserve. These plots represent a chrono-sequence of secondary  
159 succession following land use abandonment, utilizing a space-for-time approach. In total, 10 plots  
160 were established in active agriculture, 30 plots in recovering sites after land use abandonment, and  
161 16 plots in primary old-growth forest. The plots are randomly distributed within an area of 140 km<sup>2</sup>  
162 and differ in their current land use status (see Fig. 1). The recovering plots comprise former  
163 agricultural land left to regenerate naturally without the influence of active restoration  
164 interventions. They can be further characterized by their historical land use (cacao plantation or  
165 pasture) and their stage of succession (time since land use abandonment, termed “recovery age” in  
166 the following). Abandoned cacao plantations are characterized by small- to mid-sized cacao trees  
167 with mostly dense vegetation in the lower levels during the first years of recovery. Later recovery  
168 stages are marked by mixed stands and larger cacao trees.

169 Abandoned pastures characterized by mixed stands of small pioneer species and dense ground  
170 vegetation in the young stages of recovery. Later recovery stages are characterized by larger trees  
171 and less dense ground vegetation. The recovery age of the plots ranges from 0 to 39 years.

172 The age of reference old-growth forest plots is unknown. Each plot is arranged in a square shape  
173 with a side length of 50 meters (2500 m<sup>2</sup>), except for five plots, where plot size had to be adjusted  
174 to 20 x 20 meters (400 m<sup>2</sup>) due to practical reasons resulting from active land use. The distribution

175 of the vegetation type along the measured 56 plots is shown in Fehler! Verweisquelle konnte nicht  
176 gefunden werden..  
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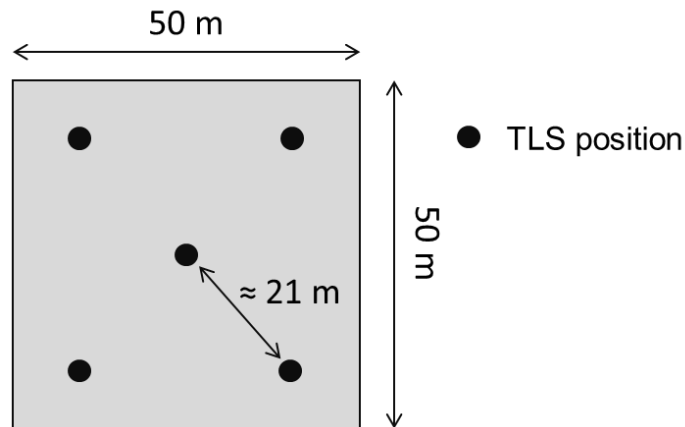


178  
179 Figure 1. Map of the study area with plot locations and elevation above sea-level (m).

180  
181 *Quantification of forest structural complexity*

182 At each plot, forest structural complexity was quantified based on single terrestrial laser scans,  
183 which were conducted using a Faro Focus M70 (Faro Technologies Inc., Lake Mary, USA) terrestrial  
184 laser scanner in November 2023. Scans were systematically distributed across a plot, with one scan  
185 made in the center and the other four scans towards the corners with a diagonal distance of  $\approx 21$   
186 m from the center (see Fig. 2). The terrestrial laser scanner was mounted on a tripod at a height of  
187 1.3 m. The area was scanned with a field of view of  $360^\circ$  horizontally and  $300^\circ$  vertically with an  
188 angular step width of  $\approx 0.035^\circ$ . The maximum scan distance of this device reaches up to 70 m.  
189 Where big trees or inaccessibility due to dense vegetation, ponds or steep slopes did not allow for  
190 setting up the scanner, the nearest possible area within the plot was used. The resulting three-

191 dimensional point cloud was then used to calculate the stand structural complexity index (SSCI)  
192 following Ehbrecht et al. (2017, 2021).



193  
194 Figure 2. Distribution of single terrestrial laser scans within each plot.

195  
196 Scan data was processed using the hardware-specific software FARO SCENE (Faro Technologies Inc.,  
197 Lake Mary, USA, v.7.1.1.81). Point clouds were then exported as text file in .xyz-format. The .xyz-  
198 files were then been imported to R to calculate the stand structural complexity index (SSCI) for each  
199 scan using the publically available R-code by Ehbrecht et al. (2021, GitHub-Link). The SSCI is based  
200 on two components: (1) the mean fractal dimension of cross-sectional polygons derived from the  
201 point cloud (MeanFrac) and (2) the effective number of layers (ENL) (Ehbrecht et al., 2016). The  
202 cross-sectional polygons represent slices through the point cloud for 1280 azimuthal directions with  
203 the scanner position being the center for each polygon. Polygons are constructed by connecting  
204 points of the point cloud along the hemisphere for each azimuthal direction. For each polygon, the  
205 shape complexity is then calculated based on a modified perimeter-area-ratio following McGarigal  
206 & Marks (1995, for details see Ehbrecht et al. 2021). The resulting 1280 fractal dimension values are  
207 then aggregated to a mean fractal dimension (MeanFRAC) as a measure of structural complexity.  
208 The MeanFRAC index increases with increasing canopy space filling and can be interpreted as a  
209 proxy for vegetation density. As the fractal dimension is a scale-invariant measure of complexity,  
210 MeanFrac values are scaled using the effective number of layers (ENL) to take stand height and  
211 vertical stratification into account. ENL is based on the concept of Foliage Height Diversity,  
212 introduced by MacArthur & MacArthur (1961), and quantifies the number of layers that are  
213 effectively occupied by foliage and woody components. Here, the inverse Simpson-Index is applied  
214 to the vertical distribution of points binned in 1 m-thick layers. As such, ENL quantifies the  
215 distribution of canopy elements in three-dimensional space and be interpreted as measure of  
216 vertical foliage distribution. SSCI is then calculated following equation 1:

217

218 
$$\text{SSCI} = \text{MeanFrac}^{\ln(\text{ENL})} \quad (\text{eq. 1})$$

219

220 The SSCI value for each plot was then calculated as the mean SSCI based on the SSCI values derived  
221 from the five scans per plot. For detailed information on the SSCI calculation see Ehbrecht et al.  
222 (2017, 2021). SSCI has been used in several studies as a measure of vegetation complexity, incl.  
223 (Kikuchi et al., 2024; Perles-Garcia et al., 2021; Ray et al., 2023; Soto et al., 2024; Willim et al., 2022)  
224

#### 225 *Assessment of tree species diversity*

226 At each plot, a tree census was conducted between February 2022 and July 2023, whereby each  
227 tree with a circumference  $\geq 25$  cm at 1.3 m above the ground ( $\text{dbh} \geq 7.95$  cm) was recorded,  
228 including palms and lianas (Escobar et al., 2025). A total of 7542 cultivated and non-cultivated trees  
229 were surveyed, from which 7408 trees ( $> 98\%$ ) were identified to species or genus level. We  
230 identified 539 tree species and morphospecies within the census. Tree species diversity was then  
231 estimated based on coverage-based rarefied species using the iNEXT package. To investigate the  
232 role of tree diversity in shaping forest structural complexity, we focus on the diversity of dominant  
233 tree species using Hill number  $q = 2$  for all analyses.

234

#### 235 *Statistical analyses*

236 All data analyses were conducted using the statistical software environment R, version 4.3.3 (R  
237 Development Core Team 2024). Differences of SSCI between the vegetation types and their  
238 significance have been tested with an analysis of variance (ANOVA) and a subsequent TukeyHSD  
239 post-hoc test. To test how forest structural complexity changes along the successional gradient from  
240 actively managed cacao plantations and pastures to late-successional stages, SSCI as well as both of  
241 its components were first regressed against recovery age. Data from old-growth forest plots was  
242 used as reference. For SSCI, we tested a linear model with a log-transformation of recovery age, a  
243 generalized additive model (*mgcv*-package (Wood, 2023)) and an asymptotic, non-linear least  
244 squares regression model (*nls2* package (Grothendieck & Team (nls), 2024)). For ENL, we tested a  
245 linear and log-linear model. For MeanFRAC, we tested a linear model only, as no clear trend was  
246 apparent in the data. Afterwards, the Akaike Information Criterion (AIC) values of the models have  
247 been compared and the model with the lowest AIC value was selected as the best fitting model. In  
248 figures, only regression lines for the best fitting model are shown.

249 From a mechanistic perspective, recovery age can be considered an indirect driver of forest  
250 structural complexity. We hypothesize that effects of recovery age on forest structural complexity  
251 are mediated by increasing tree species diversity and tree size (canopy height) along the studied-  
252 chrono-sequence, whereby canopy height determines the three-dimensional canopy space (niche



253 space) that can be occupied by canopy elements. Tree species diversity is then hypothesized to  
 254 determine niche occupation and therewith canopy space filling as well as vertical stratification. To  
 255 investigate whether effects of recovery age on forest structural complexity are mediated by tree  
 256 species diversity and canopy height, we employ a structural equation modelling approach, in which  
 257 we test the effects of those predictors on the above-mentioned components of SSCI, namely ENL  
 258 and MeanFRAC, using the *piecewiseSEM* package (Lefcheck et al., 2024). Additionally, we tested  
 259 whether tree species diversity ( $q = 2$ ) explained the residuals of models explaining the structural  
 260 metrics as a function of recovery age. We thereby aimed to test whether tree species diversity  
 261 affects three-dimensional forest structure regardless of recovery age. We set a significance level of  
 262  $p < 0.05$  for each analysis. Residuals of all regression models were tested using the *DHARMA* package  
 263 (Hartig et al., 2024). Residuals diagnostics are presented in the Supplementary Information (Fig. SI  
 264 3 and Fig. SI 4).

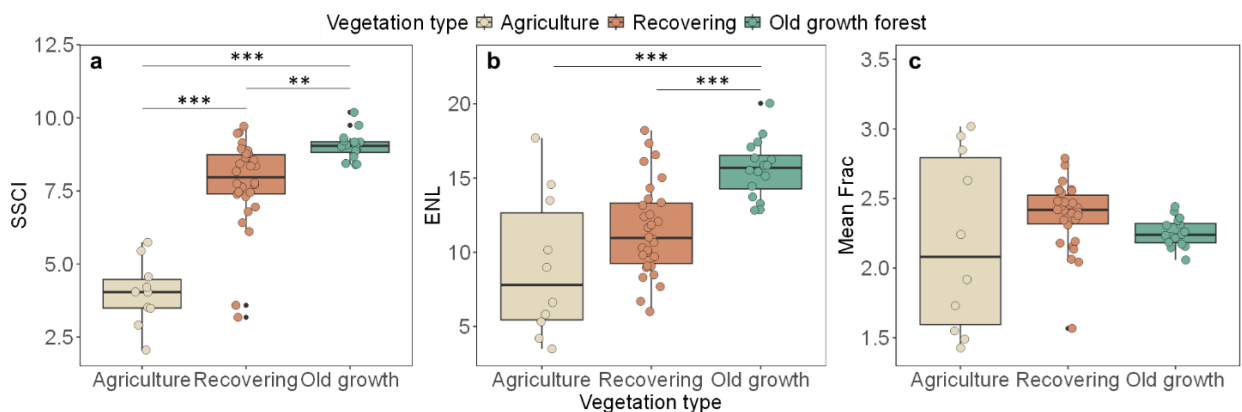
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## 267 Results

268 Forest structural complexity (SSCI) increases from active agriculture to recovering sites to old-  
 269 growth forest and significantly differs between the three vegetation types (Fig. 3a,  $p < 0.01$ ). A  
 270 similarly increasing pattern can be observed for ENL, whereas no significant differences between  
 271 agriculture and regenerating sites could be detected (Fig. 3b,  $p = 0.10$ ). For MeanFRAC, we did not  
 272 find significant differences between the three vegetation types (Fig. 3c,  $p > 0.21$ ).

273



274

275 Figure 3. (a) Forest structural complexity (SSCI), (b) vertical stratification (ENL) and (c) vegetation  
 276 density (MeanFRAC) of agricultural, recovering and old-growth forest plots, respectively.

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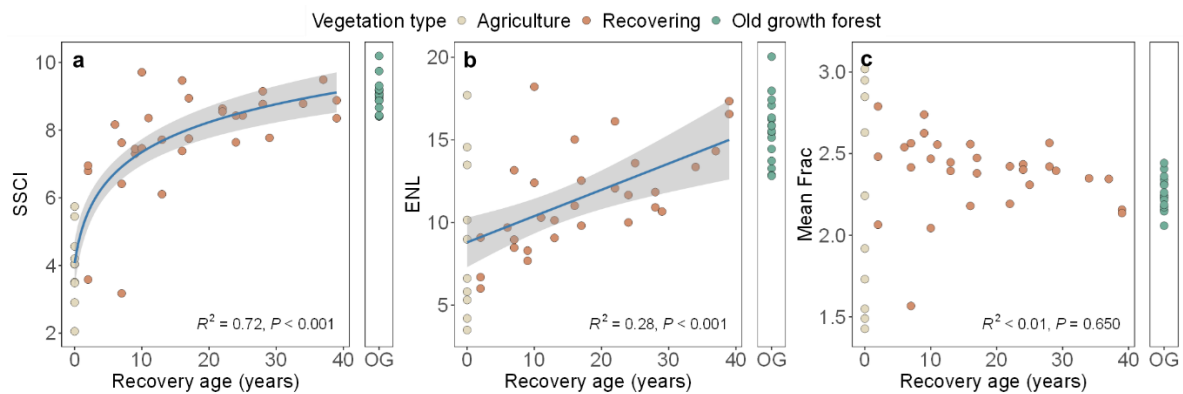
280 Table 1. Model type, response variable (structural metrics), (log-transformed) predictor, adjusted  
 281  $R^2$ , AIC and p-value of models used to predict 3D stand structure as a function of recovery age

Model	Response	Predictor	Adjusted $R^2$	AIC	p-value
lm	SSCI	log(Recovery age)	0.72	126.49	< 0.001
nls	SSCI	Recovery age	0.72	128.64	< 0.001
gam	SSCI	Recovery age	0.69	131.85	< 0.001
lm	SSCI	Recovery age	0.54	145.79	< 0.001
lm	ENL	Recovery age	0.26	210.06	< 0.001
lm	ENL	log(Recovery age)	0.2	213.03	0.002
lm	MeanFRAC	Recovery age	< 0.01	40.49	0.65

282

283 Forest structural complexity positively correlates with recovery age ( $p < 0.001$ ), which explained  
 284 72% of variation in SSCI. The best model is a linear model with a logarithmic predictor (Fig. 4a, Tab.  
 285 1). SSCI increased sharply in the first years of recovery, then the shape of the curve starts to flatten  
 286 out from the fifth to tenth year after reaching mean values of 6 to 7. At a recovery age of about 20  
 287 years, SSCI doubled its value (mean SSCI  $\approx 8$ ) and levels similar to old growth forest are reached  
 288 after 40 years of recovery (mean SSCI  $\approx 9$ ). ENL increased linearly with recovery age with no  
 289 apparent saturation at higher recovery age (Fig. 4b). MeanFRAC showed no significant correlation  
 290 with recovery age (Fig. 4c).

291



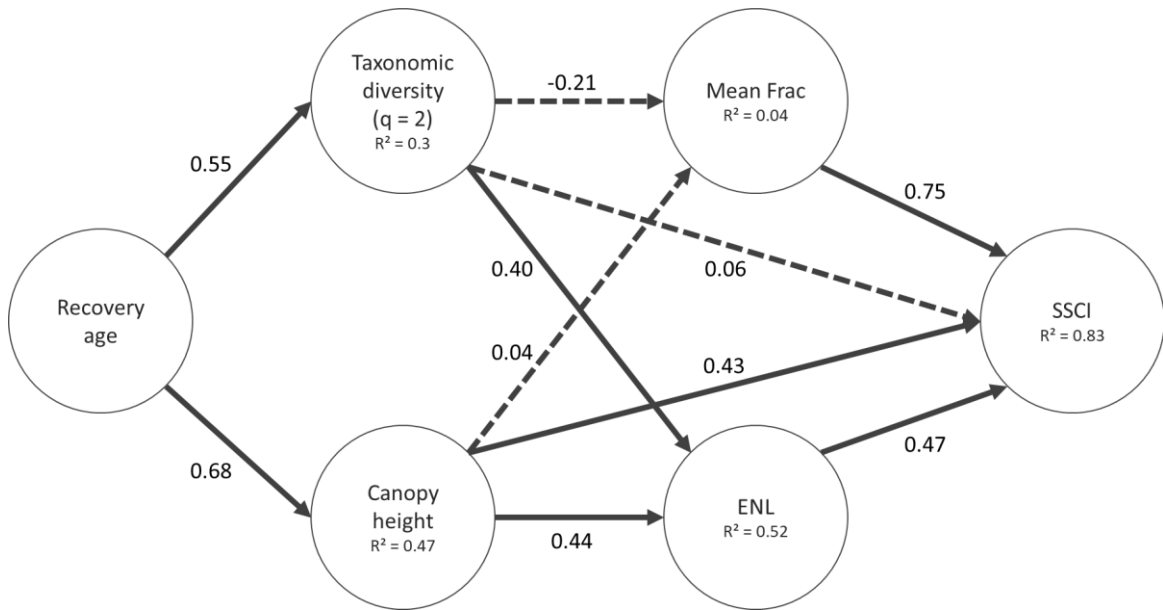
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293 Figure 4. Relationships between (a) forest structural complexity (SSCI), (b) vertical stratification  
 294 (ENL), (c) vegetation density (MeanFRAC) and recovery age. Values of old-growth forest plots are  
 295 plotted for comparison.

296

297 We hypothesized that the correlation between recovery age and forest structural complexity is  
 298 mediated by the effects of tree species diversity on the components of SSCI. With increasing  
 299 recovery age, tree species diversity and canopy height are increasing, which has been already shown  
 300 by Escobar et al. (2024) for the same plots. By employing a structural equation model, we found  
 301 that effects of tree species diversity on forest structural complexity are mediated by its effects on

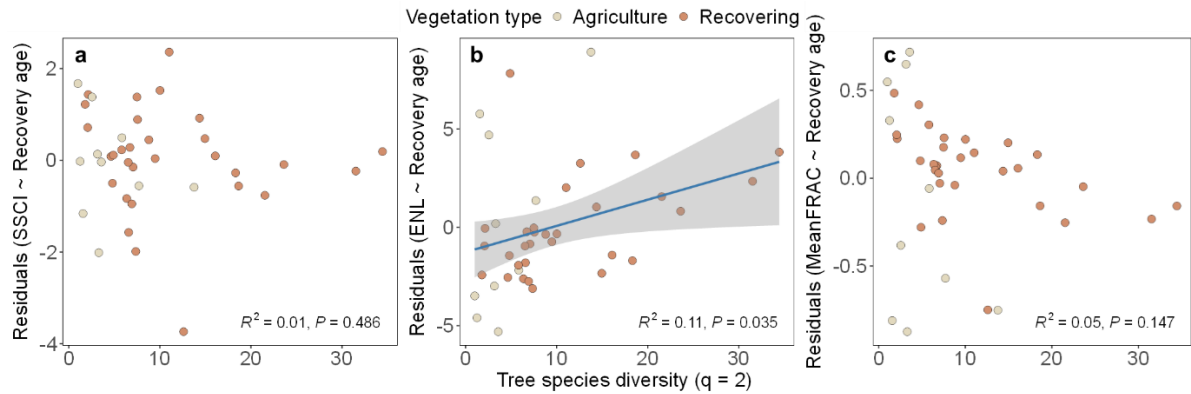
302 vertical stratification, as quantified by ENL, whereas no significant effects on vegetation density,  
 303 quantified by MeanFRAC, nor direct effects were found (Fig. 5). Canopy height had both direct and  
 304 indirect effects on SSCI with ENL as mediator.



305  
 306  
 307 Figure 5. Piecewise structural equation model relating variation in forest structural complexity (SSCI)  
 308 to tree species diversity and canopy height effects, mediated by vertical stratification (ENL) and  
 309 vegetation density (MeanFRAC) and driven by recovery age. Solid lines indicate significant effects,  
 310 dashed lines indicate non-significant relationships. Figures represent standardized effect sizes.

311  
 312 Additionally, we tested whether the residuals of models predicting SSCI, ENL and MeanFRAC as a  
 313 function of recovery age could be explained by tree species diversity (q = 2) (Fig. 6). We thereby  
 314 aimed to test whether there is an effect of tree species diversity on each of the structural metrics  
 315 regardless of recovery age. We found a significant correlation between tree species diversity and  
 316 the residuals of the model predicting ENL as a function of recovery age (Fig. 6b). Correlations  
 317 between tree species diversity and the residuals of the SSCI (Fig. 6a) and MeanFRAC model (Fig. 6c)  
 318 were not significant.

319



320

321 Figure 6. Relationships between tree species diversity and the residuals of models used to predict  
 322 (a) forest structural complexity (SSCI), (b) vertical stratification (ENL) and (c) vegetation density  
 323 (MeanFRAC) as function of recovery age.

324

### 325 Discussion

326 We assessed forest structural complexity of recovering Neotropical lowland rainforest along a  
 327 successional gradient using a space-for-time substitution approach. We hypothesized that the  
 328 structural recovery along the successional gradient for the studied tropical rainforest follows a  
 329 saturating model, characterized by an asymptotic curve. However, the most suitable model  
 330 identified describes the relationship between recovery age and SSCI with a logarithmic function,  
 331 suggesting continued (minimal) growth potential in late successional stages. Following an initial  
 332 sharp increase in forest structural complexity during the early years of forest recovery, the  
 333 successional dynamics start to slow down in the mid-stage of the chrono-sequence. Later, the curve  
 334 begins to flatten, indicating that once a certain level is reached, continued, but marginal increases  
 335 in forest structural complexity can be expected in subsequent time periods. During forest recovery,  
 336 the variance of SSCI values appears to decrease with increasing age and stabilize at comparable  
 337 levels towards late successional stages. Approximately 40 years after the initiation of forest  
 338 recovery, forest structural complexity has attained levels comparable to those identified for the old-  
 339 growth forest plots (mean SSCI  $\approx$  9). This indicates a rapid recovery of forest structural complexity  
 340 in comparison with temperate forest ecosystems, which were reported to attain levels of forest  
 341 structural complexity similar to old-growth forests after around 150 years (Gough, Bohrer, et al.,  
 342 2021). Contrary to the recovery trajectory reported for temperate forest ecosystems, we did not  
 343 observe a sigmoidal pattern of structural recovery. It is important to note that similar forest  
 344 structural complexity does not equate to a fully developed old-growth forest, as other structural  
 345 attributes, such as biomass, tree species composition, and certain ecosystem functions are still  
 346 recovering and may require much longer time to reach pre-disturbance levels (Aryal et al., 2024).  
 347 For instance, Escobar et al. (2024) show that for the same plots, above-ground biomass and tree

348 species richness have not yet reached levels comparable to old-growth forests. In a meta-analysis  
349 on tropical forest recovery, compiled from chrono-sequence data from 2275 plots in the Neotropical  
350 lowland rainforest and West Africa, Poorter, Craven, et al., (2021) reported a recovery time of  
351 around 60 years for the forest's structural heterogeneity to resemble that of old growth forests. In  
352 this study, structural heterogeneity was quantified as tree size variability using the Gini-coefficient.  
353 Even though the Gini-coefficient of tree sizes does not directly quantify the complexity of three-  
354 dimensional forest structure, it has shown to closely correlate with SSCI (Ehbrecht et al. 2021),  
355 because a heterogeneous plant material distribution in three-dimensional space as well as higher  
356 vegetation density can only be realized through a high variability of tree sizes and associated greater  
357 canopy packing. Against this background, the recovery pace of forest structural complexity in our  
358 study is similar to what has been observed in other tropical landscapes. Furthermore, after 40 years  
359 of recovery, SSCI values are comparable with old-growth tropical rainforests of other realms, such  
360 as Indo-Malayan or Oceanian rainforests (Ehbrecht et al., 2021).

361

362 We hypothesized that the increase of forest structural complexity with recovery age is mediated by  
363 the effects of tree diversity and size on vegetation density and increasing heterogeneity of biomass  
364 distribution in three-dimensional space. The structural equation modelling approach employed  
365 here suggests that increases of forest structural complexity with recovery age are driven by an  
366 increase in the effective number of canopy layers (ENL), i.e. by increasing vertical stratification,  
367 resulting in a more heterogenous distribution of foliage and woody components in three-  
368 dimensional space. We did not find significant effects of tree diversity on MeanFRAC, a proxy for  
369 vegetation density. These finding suggests that the increase in tree diversity along the recovery  
370 trajectory is associated with a more efficient occupation of different forest strata by tree species  
371 with varying functional and morphological traits. This finding is further supported by the correlation  
372 between tree diversity and the residuals of the recovery age-ENL model, which suggests that tree  
373 diversity drives vertical stratification regardless of recovery age. We did not find significant  
374 correlations between residuals of the recovery age-SSCI model or the recovery age-MeanFRAC  
375 model, which suggests that the effects of tree diversity on three-dimensional forest structure are  
376 solely driven by tree diversity effects on vertical stratification, and not by tree diversity effects on  
377 vegetation density. Successional theories state that the establishment of early-successional pioneer  
378 trees after land use abandonment is followed by the establishment of mid- to late successional  
379 species along the recovery trajectory (Poorter et al., 2023). Early-, mid- and late-successional  
380 species differ in their functional traits (Manhães et al., 2022). While early-successional species are  
381 often light demanding species, late-successional species have to be more shade-tolerant in order to  
382 be able to thrive in lower strata beneath an otherwise closed-canopy forest. Thus, a multi-layered

383 canopy may develop along the recovery trajectory where light-demanding species occupy upper  
384 canopy layers and shade-tolerant species mid- to lower canopy layers, resulting in a higher effective  
385 number of canopy layers. In late-successional stages, canopy openness and tree height marginally  
386 changed (see Fig. SI 1), whereas ENL continued to increase at similar rates. This suggests the  
387 formation of additional canopy layers by climax tree species emerging from the understory, resulting  
388 in the characteristic vertical stratification of tropical rainforests. The assumption that increases in  
389 vertical stratification along the recovery trajectory are driven by greater functional diversity is  
390 supported by other studies that reported increases in functional diversity with recovery age.  
391 Surprisingly, we did not detect changes in vegetation density with recovery age. However, the  
392 slightly lower MeanFRAC-values of old-growth forest in comparison with recovering forest may  
393 suggest that vegetation density might decrease as succession continues. As dominant trees mature,  
394 competition for light and belowground resources can result in insufficient resource availability for  
395 trees of lower canopy strata, which might lead to a less dense understory vegetation (Matsuo et al.,  
396 2024). Similar results have been reported for other old-growth forests (Ehbrecht et al. 2021).

397

398 Positive effects of tree diversity on forest structural complexity were further reported for temperate  
399 broadleaved forests (Juchheim et al., 2020), savannas (Coverdale et al., 2024), tropical tree diversity  
400 experiments (Perles-Garcia et al., 2021; Ray et al., 2023; Zemp et al., 2019) and tropical agroforestry  
401 systems (Steinfeld et al., 2024). So far, evidence of positive tree diversity effects was lacking for  
402 naturally regenerating tropical forests. By reviewing diversity-complexity relationships, Coverdale &  
403 Davies, (2023) suggest that effects of tree diversity on forest structural complexity usually follow a  
404 saturating pattern. At higher levels of tree diversity, additional species may not increase forest  
405 structural complexity further due to functional redundancy as soon as the available niche space is  
406 fully occupied. Even though we find a saturating relationship between tree diversity and forest  
407 structural complexity in our study (see Fig. SI 2), we cannot draw robust conclusions from this  
408 relationship because the tree diversity-SSCI relationship is confounded by recovery age in a  
409 univariate regression.

410

411 With increasing interest in forest recovery after agricultural use and the goal of restoring  
412 biodiversity and ecological functionality by promoting forest stand structural complexity, the need  
413 for adequate management practices is growing. Future studies, efforts or experiments on forest  
414 recovery under similar conditions may refer to results from this study, as a comparison with natural  
415 succession dynamics can be employed to evaluate the success of assisted management approaches  
416 and the impact of implemented techniques on the development of forest structural complexity.

417 With its hyper-diverse and evergreen tree communities, tropical rainforests are among the forest  
418 types with the highest SSCI values worldwide (de Conto et al., 2024; Ehbrecht et al., 2021). In  
419 addition to trees, other vegetation groups in the rainforest should be focused on further research  
420 when assessing stand structural complexity. The uniquely high occurrence of epiphytes and lianas  
421 in tropical rainforests and their effects have not been extensively considered in SSCI evaluations to  
422 date. Given that these plant groups establish themselves in highly specialized niches on tree  
423 surfaces, their impact on the structural complexity of the forest would be of particular interest.

424

425 Overall, our results suggest that passive restoration is an effective tool to restore forest structural  
426 complexity after land use abandonment, allowing a structural recovery within time frames of at  
427 least 40 years. However, the recovery trajectory strongly depends on the colonization success by  
428 pioneer tree species, which in turn depends on a vital seed bank and/or seed dispersal from  
429 adjacent intact forests and subsequent recruitment success. This questions whether our findings  
430 hold for other human-modified tropical landscapes, where landscape fragmentation, limited seed  
431 dispersal, or colonization of abandoned sites by grasses or shrubs might hinder the recovery of  
432 forest structural complexity in comparable time frames.

433

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448

#### 449 **Conflict of interest statement**

450 The authors declare no conflicts of interest

451

452 **Data availability statement**

453 The data and code used in this study will be made available upon acceptance of the study in a peer-  
454 reviewed journal.

455

456 **Author contributions**

457 ME and NB conceptualized the study. ME wrote the first draft and analyzed the data. TL collected  
458 the data and contributed to data analyses and manuscript writing. SE contributed data and  
459 contributed to manuscript writing. DD, MJE and JGA contributed equally to manuscript revisions.

460

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**Supplementary Information**

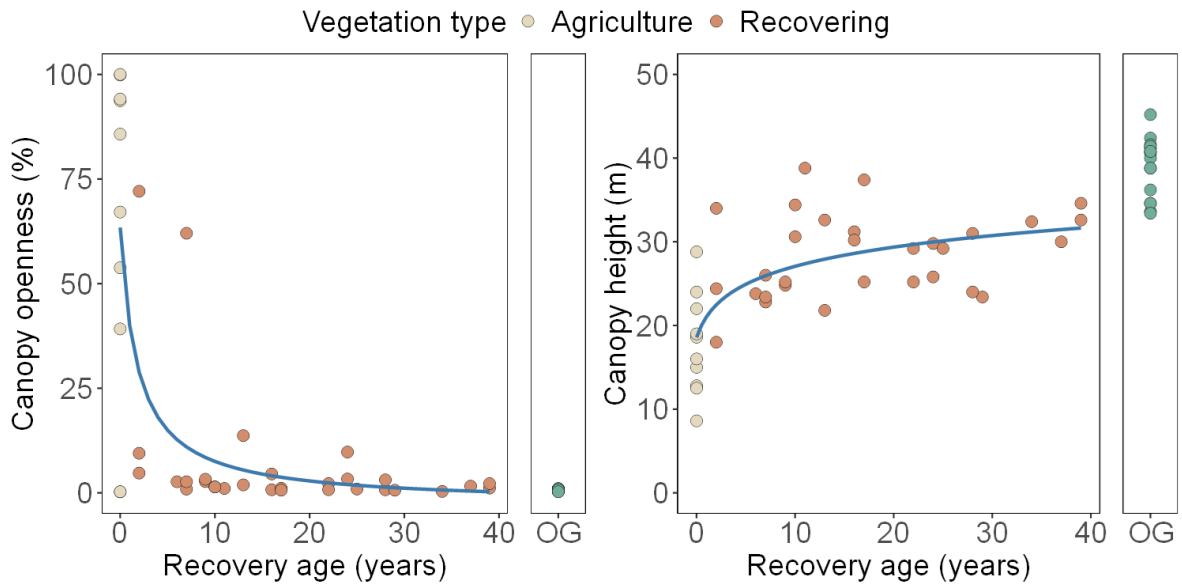


Figure SI.1. Canopy openness (*left*) and canopy height (*right*) as a function of recovery age. Canopy openness was modelled using a negative exponential asymptotic function. Canopy height was modelled with a log-linear model. Values for old-growth forest are shown for comparison, respectively.

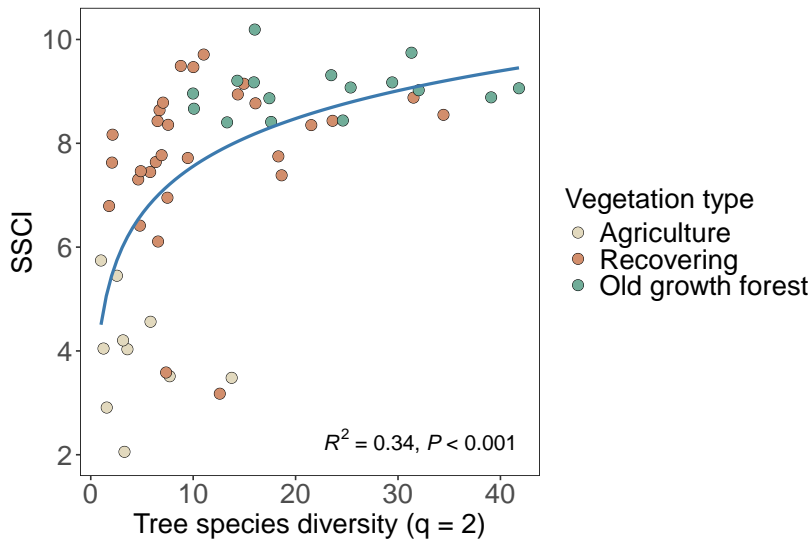


Figure SI.2. Relationship between forest structural complexity (SSCI) and tree species diversity (q = 2).

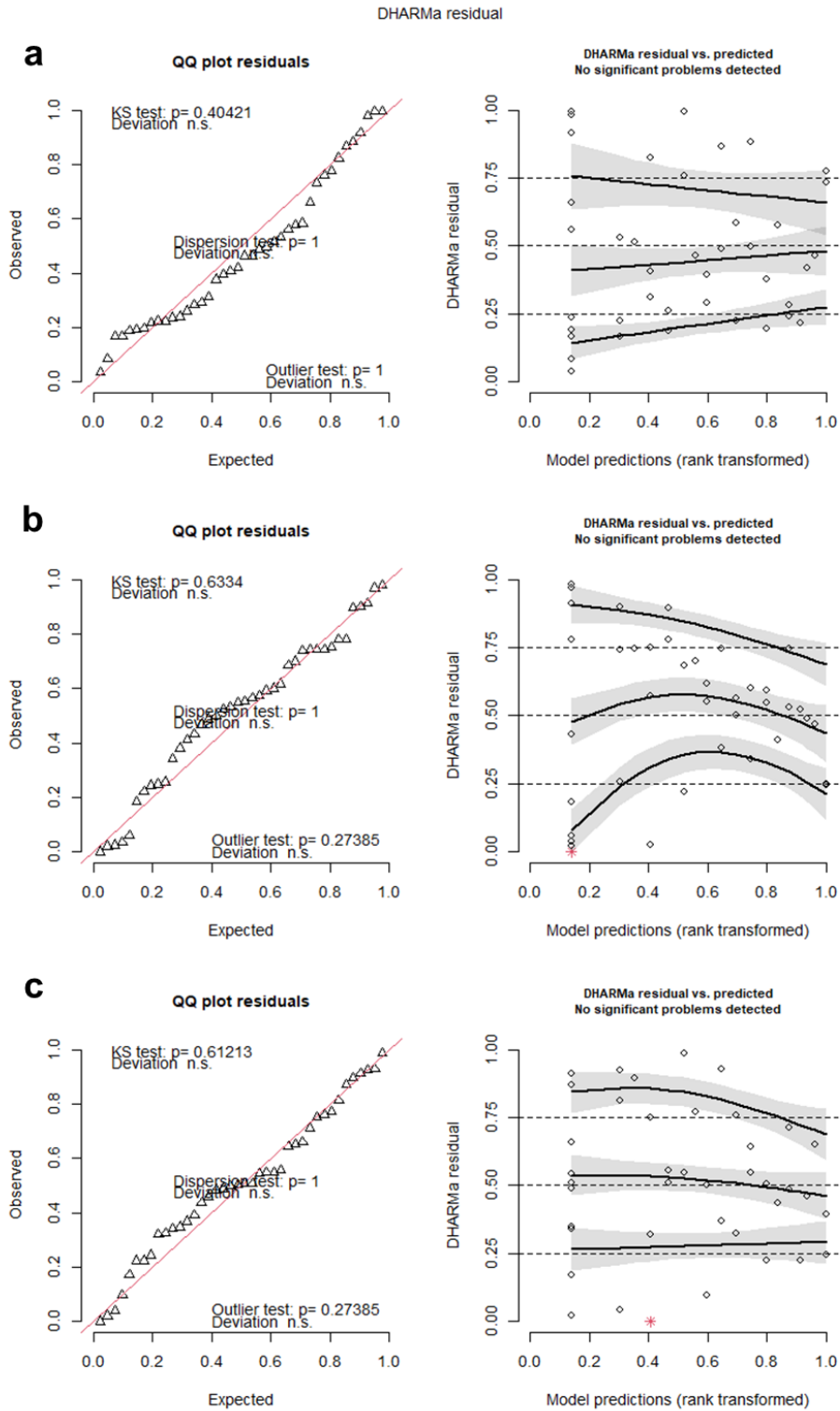


Figure SI.3. Residual diagnostics for models explaining (a) forest structural complexity (SSCI), (b) vertical stratification (ENL) and (c) vegetation density (MeanFRAC) as a function of recovery age.



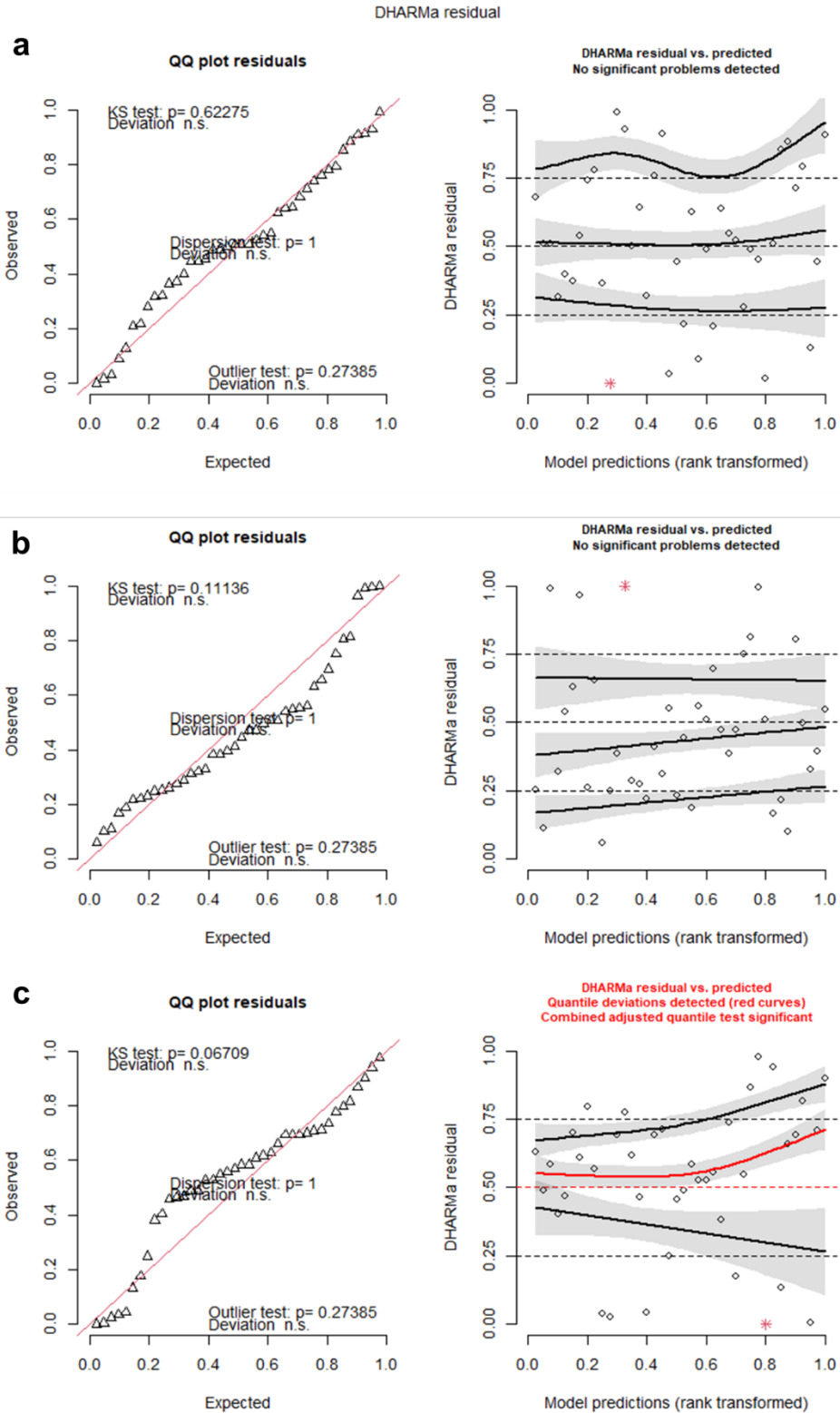


Figure SI.4. Residual diagnostics for models explaining the residuals of recovery age effects on (a) forest structural complexity (SSCI), (b) vertical stratification (ENL) and (c) vegetation density (MeanFRAC) as a function of tree species diversity.

