

1 Title: **The association between rainforest disturbance and recovery, tree community**  
2 **composition and community traits in the Yangambi area in the Democratic Republic of the**  
3 **Congo**

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15 Running title: Tropical forest association with disturbance and recovery

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22 **ABSTRACT**

23 Despite their key role in biodiversity conservation, forests in the Congo Basin are increasingly  
24 threatened by human activities, but it remains challenging to assess the impact of forest  
25 degradation under a more or less intact canopy. Likewise, forest recovery following agricultural  
26 abandonment remains poorly understood in the Congo Basin. Here, we surveyed 125 vegetation  
27 quadrats across 25 forest inventory plots in the Yangambi area. We aimed to find associations  
28 between both selective logging and forest recovery, and a range of forest community and tree  
29 community trait characteristics, as compared to reference undisturbed old-growth forest. We  
30 found that plots in undisturbed old-growth forest harboured both more tree individuals and tree  
31 species with a higher wood density as compared to plots in disturbed old-growth forest. In  
32 addition, their tree community composition was significantly different, whereas species diversity  
33 recovered since relatively recent agricultural abandonment (< 60 years), community composition  
34 and forest structure remained significantly different from the plots in undisturbed old-growth  
35 forest. Our study provides some insights in the rate of forest recovery in the Congo basin after  
36 agricultural abandonment and highlights the need of proper conservation of the remaining  
37 relatively undisturbed old-growth forests. Finally, we stress the need for more extensive  
38 vegetation surveys in the Congo Basin to further unravel the effects of anthropogenic  
39 disturbance.

40

41 Keywords: Africa, historical land-use, lowland Rainforest, specific leaf area, wood density

42

## 43 1. INTRODUCTION

44 The African continent is home to the world's second largest area of tropical forests which are,  
45 together with earth's subtropical forests, by far the most floral rich biomes on earth (Kier *et al.*  
46 2005, Malhi *et al.* 2013). The African tropical forests appear to be less biodiverse than their  
47 Asian and American counterparts (Kier *et al.* 2005, Parmentier *et al.* 2007, Slik *et al.* 2015).  
48 Nevertheless, Sosef *et al.* (2017) recorded 22,577 plant species in tropical African forests, of  
49 which 30% are endemic, and concluded that much botanical exploration still remains to be done.  
50 The forests of the Congo Basin, which account for 89% of the African tropical forests, are also  
51 extremely important as carbon sinks, removing up to six times more carbon than the forests in  
52 the Amazon region (Harris *et al.* 2021, Malhi *et al.* 2013). Furthermore, the African carbon sink  
53 has been stable for the past three decades until 2015, while the Amazonian carbon sink has been  
54 declining since the 1990s (Hubau *et al.* 2020).

55 Despite their key role in biodiversity conservation and carbon sequestration, forests in  
56 tropical Africa are increasingly threatened by human activities (Tyukavina *et al.* 2018). Between  
57 1990 and 2000, the annual net loss of forests in Africa corresponded to 3.3 million ha yr<sup>-1</sup>,  
58 increasing to 3.4 million ha yr<sup>-1</sup> between 2000 and 2010, and further to 3.9 million ha yr<sup>-1</sup>  
59 between 2010 and 2020 (FAO & UNEP 2020). Africa is currently the only region where annual  
60 forest change rates are still increasing (FAO & UNEP 2020). The deforestation on the African  
61 continent is mainly caused by small-scale subsistence agriculture (Curtis *et al.* 2018, Jayathilake  
62 *et al.* 2021, Tegegne *et al.* 2016, Tyukavina *et al.* 2018), driven by very low agricultural yields  
63 (Carrasco *et al.* 2017, West *et al.* 2010), and a rapidly growing human population (Gerland *et al.*  
64 2014).

65           Although deforestation is a very conspicuous and severe threat to biodiversity,  
66 degradation of tropical forests may even be more detrimental, based on the large spatial scales at  
67 which it occurs (Barlow *et al.* 2016). Forest degradation refers to within-forest disturbance and  
68 mainly includes selective logging, removal of understorey vegetation, and overexploitation of  
69 specific species (Sasaki & Putz 2009, Chaturvedi *et al.* 2012, Tyukavina *et al.* 2018). In tropical  
70 Africa, forest degradation is primarily driven by selective logging, mainly for the small-scale  
71 production of charcoal which is still the most commonly used energy source in many Central  
72 African countries (Hosonuma *et al.* 2012, Jayathilake *et al.* 2021, Tyukavina *et al.* 2018). The  
73 degradation rate of forests in the Congo Basin has been estimated to be 317,000 ha yr<sup>-1</sup> between  
74 2000 and 2005 (Ernst *et al.* 2013). Lambin *et al.* (2003) estimated that the annual rate of human  
75 mediated forest degradation in Africa corresponds to approximately 50% of its annual rate of  
76 deforestation. This would imply that 1.95 million ha of African forest has been degraded  
77 between 2010 and 2020. However, Shapiro *et al.* (2021) recently reported that this assumption  
78 may drastically underestimate the actual rate. Using a forest condition index, they estimated that  
79 between 2000 and 2016, over 23 million ha of forest has been degraded in the Congo Basin  
80 alone.

81           Although remote-sensing techniques are constantly improving, it remains challenging to  
82 monitor the extent and impact of forest degradation due to the interference with the often  
83 untouched canopy in degraded forests (Matricardi *et al.* 2020, Peres *et al.* 2006). Evaluation of  
84 the extent and impact of forest degradation must therefore largely rely on vegetation surveys on  
85 the ground (Heym *et al.* 2021). Impacts of forest degradation on biodiversity have been  
86 frequently studied in the Neotropics (e.g. Carreño-Rocabado *et al.* 2012, Mirabel *et al.* 2020),  
87 Asia (e.g. Baboo *et al.* 2017, Millet *et al.* 2010), and other parts of Africa (e.g. Bobo *et al.* 2006,

88 Oyugi *et al.* 2008), but similar studies are rare in the Congo Basin (but see Hall *et al.* 2003, Ifo *et*  
89 *al.* 2016), and almost lacking in the Democratic Republic of Congo (DR Congo) (but see  
90 Makana & Thomas 2006). Yet, as DR Congo harbours almost 60% of the African tropical forests  
91 (Malhi *et al.* 2013), it is essential to comprehend how plant community composition is affected  
92 by forest disturbance.

93         While conservation of existing tropical forests should still be the priority, the restoration  
94 of deforested land and degraded forests can aid in the recovery of tropical biodiversity and the  
95 recovery of ecosystem services (Brudvig 2011, FAO & UNEP 2020). Between 2000 and 2019,  
96 0.7 million ha of forest has been actively or passively restored in Sub-Saharan Africa (FAO &  
97 UNEP 2020). By 2030, the African Forest Landscape Restoration Initiative (AFR100) aims to  
98 increase the land surface under restoration to 100 million ha. DR Congo has committed to restore  
99 8 million ha (AFR100 2021). Because of the high variability in restoration outcomes, more  
100 research is crucial to clear the path for restoration (Chazdon 2008). Again, studies on the  
101 biodiversity consequences of tropical forest recovery are more commonly found in the  
102 Neotropics (e.g. Letcher & Chazdon 2009, Oberleitner *et al.* 2021) and tropical Asia (e.g.  
103 Hayward *et al.* 2021, Mahayani *et al.* 2020), but are rare in the Congo Basin (but see Bauters *et*  
104 *al.* 2019, Makelele *et al.* 2021).

105         In this study, we established a set of forest inventory plots in the Yangambi region in DR  
106 Congo, and aimed to find associations between both anthropogenic disturbance through selective  
107 logging and forest recovery on the one side, and a range of forest and tree community  
108 characteristics on the other side, as compared to reference undisturbed old-growth forest. Based  
109 on 125 sampling quadrats across 25 forest inventory plots, we specifically quantified differences  
110 among (i) disturbed old-growth forest, (ii) regrowth forest, and (iii) undisturbed old-growth

111 forest in terms of tree species diversity, community traits, community composition, forest  
112 structure, and canopy cover. We hypothesise, that (i) selective logging results in the decreasing  
113 abundance of some specific tree species and in shifts in community composition, without having  
114 a profound effect on the overall tree species diversity, (ii) that selective logging results in a lower  
115 community-weighted mean wood density and in lower crown closure, and (iii) that forest  
116 recovery following agricultural abandonment will be fast for secondary tree species, with low  
117 wood density, but slower for primary forest tree species, with higher wood density, implying that  
118 the overall tree species diversity will have recovered, but that the community composition will be  
119 different from undisturbed old-growth forest.

## 120 **2. METHODS**

### 121 **2.1. Study site and set up**

122 Vegetation sampling was conducted in the Yangambi region, in the Tshopo province in North-  
123 Eastern DR Congo, approximately 100 km west of Kisangani between November 2020 and  
124 January 2021. In Yangambi, the climate is characterised by two drier seasons (December-March  
125 & June-July) and two rainy seasons (April-May & August-November) (van Vliet *et al.* 2018).  
126 The general landscape type in this area is typical for the Congo Basin and is characterised by a  
127 mosaic of land tenures: the Yangambi Man and Biosphere Reserve created in 1979; the Ngazi  
128 Forest Reserve; a logging concession; and customary land. To date, there is no official  
129 management plan (van Vliet *et al.* 2018). Prior to the Congolese independence in 1960,  
130 Yangambi housed a research station of INEAC (Institut National pour l'Etude Agronomique du  
131 Congo Belge), with multiple pastures for research and breeding activities. However, in 1962 the  
132 INEAC changed into INERA (Institut National des Etudes et Recherches Agronomiques) in  
133 1962. Since then, the research activities at Yangambi started to diminish (Vanden Abeele *et al.*

134 2021), and as a result, many agricultural fields were abandoned, allowing forest to recover on  
135 these lands.

136 We established 25 forest inventory plots of 125 m x 125 m (1.56 ha), covering an area of  
137 ca. 50 by 20 km, just North of the Congo River (Figure 1). Plots were assigned to three different  
138 forest categories by georeferencing (WGS84) them onto historical land-use maps, which were  
139 issued by Evrard (1954). Plots located on historical agricultural land from INERA were  
140 classified as regrowth forest (R). Whereas we could not exactly date the start of the forest  
141 recovery, our best estimate based on local sources is that the agricultural land was abandoned  
142 somewhere between 40 and 60 years ago. Plots located in historically old-growth forests areas  
143 but where we observed indications of considerable small-scale selective logging through the  
144 presence of tree stumps and other remnants of logging such as parts of trees or active logging,  
145 were considered as disturbed old-growth forest (DO). These areas were typically located near  
146 regrowth forest and close to places with relatively higher population density. The remaining plots  
147 that were located in areas with historical old-growth forests, and where no signs of selective  
148 logging were present, were categorised as undisturbed old-growth forest (UO). The current UO  
149 and DO forests have previously been classified by Evrard (1954) as heterogenous forests with  
150 primitive character. These old vegetation maps have been shown to be very reliable (Vancutsem  
151 *et al.* 2009), and we may therefore assume there were no pre-existing differences between both  
152 forest categories.

153

154

## 155 **2.2. Vegetation survey and plant trait quantification**

156 Five quadrats (25 m x 25 m) were randomly established in each forest inventory plot, resulting in  
157 a total of 125 sampling quadrats. In each quadrat, the abundance and diameter at breast height  
158 (DBH) of all woody species with a DBH  $\geq$  5 cm were measured. The quadrats were geolocated  
159 (Garmin GPSMAP 64 series) and crown closure (%) was quantified from four readings in the  
160 cardinal directions with a model A convex spherical densiometer.

161 The specific leaf area (SLA) of all taxa identified at species level was assessed based on  
162 dry herbarium specimens from the Meise Botanic Garden herbarium following Pérez-  
163 Harguindeguy *et al.* (2013). Perez *et al.* (2020) provided evidence that the SLA can reliably be  
164 estimated from herbarium-based measurements. Leaf area was assessed using leaf disks (4.9  
165 cm<sup>2</sup>), to minimise damage to herbarium specimens, or otherwise by a whole leaf scan using  
166 ImageJ (Rueden *et al.* 2017). The obtained disks and whole leaves were weighted afterwards.  
167 Five replicates were used per species. In addition to the SLA, the wood density (WD) of 141  
168 species was compiled from the literature (Table S1) (RMCA 2021, Carsan *et al.* 2012, Harja *et*  
169 *al.* 2021).

### 170 **2.3. Data analyses**

171 Statistical analyses were performed using R software (R Core Team 2020). The data of the five  
172 quadrats were summed into one value per plot. For each plot, the number of trees, species  
173 richness (Hill's  $N_0$ ), species diversity (Hill's  $N_1$  or the exponential of the Shannon entropy) and  
174 evenness (Hill's  $N_1/\log(\text{Hill's } N_0)$ ) were then calculated. These numbers were then standardised  
175 for sample completeness, as sample sizes differed between plots. Sample completeness  
176 represents the proportion of the total number of individuals in a unit that belongs to the species in  
177 the sample, contrary to sample size standardisation which allows less biased comparisons (Chao  
178 & Jost 2012). Extrapolation of the estimates was done to twice the reference sample size per plot,



179 following Chao & Jost (2012). Calculations were based on the collected species abundance data  
180 with 9999 bootstraps using the *iNEXT* function of the iNEXT package (Hsieh *et al.* 2016).  
181 Subsequently, the different variables and the crown closure were tested for differences between  
182 the three forest categories using Kruskal-Wallis rank sum tests and pairwise Wilcoxon rank sum  
183 tests with correction for multiple testing, as the assumptions for parametric testing were not met.

184         Dissimilarity of the tree community composition across the forest categories was  
185 visualised using a non-metric multidimensional scaling (NMDS), based on the Bray-Curtis  
186 distance matrix, the most suitable distance for abundance data (Ricotta *et al.* 2017). A two-axis  
187 solution provided a stress level (0.13) well below the 0.20 threshold, indicating a good  
188 representation in reduced dimensions. The ordination was performed with the *metaMDS* function  
189 in the *vegan* package (Oksanen *et al.* 2020). Subsequently, a Mantel test with 9999 permutations  
190 was performed to test the effect of geographic distance on the dissimilarity of the tree  
191 community composition using the *mantel* function in the *Vegan* package (Oksanen *et al.* 2020).  
192 In addition to the NMDS, a permutational multivariate analysis of variance with 9999  
193 permutations was performed to test for differences in community composition between the three  
194 different forest categories using the *adonis* function in the *Vegan* package (Oksanen *et al.* 2020).  
195 Pairwise comparisons of the different forest groups with 9999 permutations and Bonferroni  
196 correction were performed with the *pairwise.adonis.function* in the *pairwiseAdonis* package  
197 (Arbizu 2017). Tree species indicative for each of the three forest categories were identified  
198 based on their indicator value (Dufrene & Legendre 1997). This value was calculated using the  
199 *multipatt* function in the *indicspecies* package (De Cáceres & Legendre 2009), with 9999  
200 permutations.

201 Community-weighted means (CWM) for each plot were calculated for the SLA and the  
202 WD, using the number of individuals as weight. The community-weighted means are known to  
203 be very responsive to environmental change and highly sensitive to disturbance (Carreño-  
204 Rocabado *et al.* 2012). Both variables met the assumptions for parametric testing and to evaluate  
205 differences in community-weighted mean SLA and WD among forest categories, a one-way  
206 analysis of variance (ANOVA) was conducted followed by pairwise Tukey post hoc tests.

207 Forest structure was evaluated using the measured DBHs. Trees were categorised in ten  
208 size classes according to their DBH. An interval of 10 cm was used in the first nine classes,  
209 starting from 5 cm DBH. The tenth class had a wider interval, aggregating individuals with a  
210 DBH over 95 cm. Subsequently, the diameter frequency distribution was compared among the  
211 three forest categories with Pearson's chi-squared tests with simulated p-values based on 9999  
212 replicates using the *chisq.test* function in the stats package (R Core Team 2020). The chi-square  
213 test was preferred over the frequently used Kolmogorov-Smirnov test, because the former is  
214 more flexible than the latter (Mitchell 1971). The diameter frequency distribution comparison  
215 was complemented by the calculation of the structural diversity, which is a single measure  
216 summarising the diameter distribution (McElhinny *et al.* 2005). Structural diversity was  
217 quantified by the first order Hill number which was standardised for sample completeness  
218 following the same methodology as outlined above. Calculation and extrapolation of structural  
219 diversity was done with the *iNEXT* function with 9999 bootstraps in the *iNEXT* package, as  
220 above. All figures were made using the *ggplot2* package (Wickham 2016).

221

### 222 3. RESULTS

### 223 3.1. Species richness and diversity

224 A total of 7375 woody plant individuals (DBH  $\geq$  5 cm) from 211 taxa was recorded across the 25  
225 forest inventory plots. Among the 211 species, 79.15% were identified to species level, 14.22%  
226 to genus level, and 6.63% were unidentified species. Crown closure ( $\chi^2 = 2.21$ ,  $P > 0.05$ ) and  
227 hill's  $N_0$  ( $\chi^2 = 1.98$ ,  $P > 0.05$ ) were not significantly different among the three forest categories.  
228 Whereas the tree abundance ( $\chi^2 = 10.90$ ,  $P < 0.01$ ), Hill's  $N_1$  ( $\chi^2 = 9.78$ ,  $P < 0.01$ ) and evenness  
229 ( $\chi^2 = 10.06$ ,  $P < 0.01$ ) did differ between the forest categories (Figure 2). Tree abundance was  
230 17.2% lower ( $P < 0.05$ ) in DO forests, and 24.3% ( $P < 0.05$ ) lower in R forests, when compared  
231 to UO forests. Species diversity and evenness were 33.6% ( $P < 0.001$ ) and 31.9% ( $P < 0.001$ )  
232 higher, respectively, in DO forests as compared to UO forests (Table S2).

### 233 3.2. Community composition

234 The NMDS of the 25 forest inventory plots showed a spatial separation of the three forest  
235 categories (Figure 3). This was supported by the permutational multivariate analysis of variance  
236 which indicated differences between categories ( $R^2 = 0.253$ ;  $P < 0.001$ ). Specifically, community  
237 composition significantly differed between DO forests and UO forests ( $R^2 = 0.207$ ;  $P < 0.001$ ),  
238 and between R forests and UO forests ( $R^2 = 0.267$ ;  $P < 0.001$ ). The Mantel test between the  
239 geographic distance matrix and the Bray-Curtis distance matrix showed a significant but weak  
240 correlation ( $r = 0.200$ ;  $P < 0.05$ ). Across the 25 forest inventory plots, 27 out of the 167 to  
241 species level identified trees were indicative based on their indicator value (both  $P < 0.05$  and  
242 indicator value  $> 0.20$ ). Seventeen indicator species were indicative for UO forest (e.g. *Cola*  
243 *griseiflora*), four for DO forest (e.g. *Leplaea cedrata*), and six for R forest (e.g. *Heisteria*  
244 *parvifolia*) (Table S3).

### 245 3.3. Specific leaf area and wood density

246 The community-weighted means for SLA did not differ between forest categories ( $F= 1.38$ ,  $P>$   
247  $0.05$ ), however, the community-weighted mean WD exhibited a significant difference ( $F= 9.03$ ,  
248  $P < 0.01$ ) (Figure 4). In DO forests, community-weighted mean WD was 5.0 % lower ( $P<0.05$ ),  
249 whereas in R forests it was 6.2 % lower ( $P < 0.01$ ) as compared to UO forests.

### 250 3.4. Forest structure

251 The diameter distribution showed that there were more small sized trees in UO forests than in  
252 DO forests ( $\chi^2= 34.51$ ;  $P < 0.001$ ), and than in R forests ( $\chi^2=61.54$ ;  $P < 0.001$ ) (Figure 5). These  
253 differences between categories were also present for the structural diversity, represented by  
254 Hill's  $N_1$ , ( $\chi^2= 13.76$ ;  $P < 0.01$ ), where UO forests differed from DO forests ( $P < 0.01$ ), and from  
255 R forests ( $P < 0.01$ ).

## 256 4. DISCUSSION

257 Understanding how anthropogenic disturbance affects forest structure, tree species richness and  
258 community composition is of major importance for biodiversity conservation in globally  
259 underexplored biodiversity hotspots such as the Congo Basin. The goal of our study was to find  
260 associations between both anthropogenic disturbance through selective logging and forest  
261 recovery on the one side, and a range of forest community and tree community trait  
262 characteristics on the other side, as compared to reference undisturbed old-growth forest.  
263 Whereas our three delineated forest categories certainly enveloped some within-variation in  
264 terms of time since abandonment (for the regrowth forest plots) and in terms of degree of  
265 disturbance (for the disturbed old-growth forest plots), we were able to identify differences in  
266 community composition, forest structure, and community traits among forest categories.

267

#### 268 **4.1. Selective logging**

269 As hypothesised, our survey showed that there were clear differences in community composition  
270 and forest structure between plots in undisturbed old-growth forest and plots in disturbed old-  
271 growth forest. There were no differences in the number of woody species between any forest  
272 category, yet contrary to our expectation, species diversity was higher in plots in disturbed old-  
273 growth forest. This higher species diversity likely indicates that the species abundance  
274 distribution is less skewed in disturbed forests. Indeed, the evenness index showed that the  
275 species abundance is more evenly distributed in plots in disturbed forest than in plots in  
276 undisturbed old-growth forest. Combined with the observation that plots in disturbed old-growth  
277 forests contained significantly less tree individuals, these findings are in line with our hypothesis  
278 and might suggest that selective logging, in which exploitable species are selectively removed, is  
279 closely associated to these patterns. In the Yangambi area, timber is removed by local villagers  
280 using e.g. machetes, mainly for construction material and cooking purposes (Depecker pers.  
281 obs.). The stems that are removed are usually from species from which the wood is exploitable.  
282 We expect that selective logging at first only results in a shift in densities, but that it will  
283 ultimately result in the loss of certain species as selective loggers will specifically target species  
284 with stems that are exploitable. For example, *Prioria oxyphylla* and *Scorodophloeus zenkeri*  
285 were much less abundant in disturbed forests and are used as construction material. Moreover, in  
286 DR Congo the latter species is also exploited for its medicinal properties and is used to make  
287 charcoal (PROTA 2021). Other species that were highly reduced in numbers were *Dialium*  
288 *pachyphyllum*, *Greenwayodendron suaveolens*, and *Strombosia pustulata*, which all bear seeds  
289 that are dispersed by animals (PROTA 2021, Meunier *et al.* 2015). Animal-dispersed seeded tree

290 species tend to have higher wood densities than abiotic-dispersed seeded tree species (Osuri *et*  
291 *al.* 2016). Possibly indicating that loggers specifically target species with high WD, suggesting  
292 that species with animal-dispersed seeds are more vulnerable to selective logging. Careful  
293 monitoring of tree species is needed to unravel the specific effects of selective logging at the  
294 species level.

295 Plots in disturbed old-growth forest on the other hand, still contained many shade-tolerant  
296 tree species of which the wood is non-exploitable, such as species of the genus *Cola* which have  
297 a low DBH and highly irregular trunks. *Cola acuminata* also produces highly valued nuts with  
298 stimulating and medicinal properties (Meunier *et al.* 2015). Likewise, the abundance of most  
299 *Diospyros* spp. is indicative for old growth forests, of which only *Diospyros crassiflora* is  
300 exploited for its wood (Meunier *et al.* 2015).

301 Although a clear difference was found in community composition between forest  
302 categories in our study, the response in terms of species diversity and species richness was more  
303 variable. An earlier study on selective logging in the Congo Basin, compared disturbed and  
304 undisturbed forests in terms of species diversity in two distinct locations (Makana & Thomas  
305 2006). Whereas a significant influence of disturbance was detected in the first location, similar  
306 values for species richness and species diversity were detected in the second location.  
307 Furthermore, a clear effect of disturbance, primarily selective logging, on species diversity was  
308 detected in tropical forests in the Republic of the Congo (Ifo *et al.* 2016) and India (Baboo *et al.*  
309 2017), but not in a tropical forest in Bolivia (Carreño-Rocabado *et al.* 2012). It is clear that solely  
310 assessing the possible impact of anthropogenic disturbance through species diversity can mask  
311 its impact and that community composition is a better indicator of the possible impacts of forest  
312 degradation (Edwards *et al.* 2014, Imai *et al.* 2014). Indeed, although, several studies across

313 different tropical regions have reported varying responses of species diversity to forest  
314 disturbance, they all reported significant shifts in the community composition following  
315 anthropogenic disturbance (Makana & Thomas 2006, Baboo *et al.* 2017, Carreño-Rocabado *et*  
316 *al.* 2012, Ifo *et al.* 2016)

317         Because ecosystem functioning is mainly mediated by species' functional traits  
318 (Chaturvedi *et al.* 2021, de Bello *et al.* 2021), community-weighted mean functional traits can  
319 provide better ecologically information than only using taxonomic data (Katovai *et al.* 2012). We  
320 hypothesised that we would find lower community-weighted mean wood density and a lower  
321 canopy closure in plots in disturbed old-growth forest, as compared to plots in undisturbed old-  
322 growth forest. In terms of canopy closure, we did not find differences between any forest  
323 category. Consequently, the light availability in the plots remained unchanged, which  
324 presumably clarifies why we observed no significant differences in community-weighted mean  
325 specific leaf area between forest categories. It has indeed been established by Carreño-Rocabado  
326 *et al.* (2012) that increased light availability can induce changes in community-weighted mean  
327 functional traits in tropical forests. Furthermore, although disturbed and undisturbed old-growth  
328 forests were characterised by different indicator species assemblages, the majority can still be  
329 classified as shade-tolerant, which is commonly associated with low SLA (Poorter 2009).  
330 Nonetheless, the absence of differences between forest categories should be interpreted  
331 cautiously as SLA is highly responsive to environmental variation, such as sun-exposure. In  
332 contrast to community-weighted mean SLA, community-weighted mean WD did significantly  
333 differ between plots in disturbed and plots in undisturbed old-growth forest. As anticipated, a  
334 decrease in community-weighted mean WD was observed in plots in disturbed old-growth forest,

335 which indicates that trees with high WD and of high qualities as construction material, have been  
336 selectively removed (Saranpää 2003).

337 The mean diameter distribution in the forests in the Yangambi region, has previously  
338 been described by the reverse J-shaped model (Fayolle *et al.* 2014). In our study, the reverse J-  
339 shape was more pronounced in plots in undisturbed old-growth forests than in plots in disturbed  
340 old-growth forests which contained significantly less trees in the first size class, suggesting less  
341 regeneration (Gebeyehu *et al.* 2019). Similar results of reduced regeneration in forests were  
342 obtained in other tropical forests and were assigned to disturbance (Makana & Thomas 2006,  
343 Baboo *et al.* 2017, Hall *et al.* 2003).

344

#### 345 **4.2. Forest recovery following agricultural abandonment**

346 Based on the historical land-use maps, our results indicated that after 40 to 60 years there were  
347 no significant differences between plots in regrowth forests and plots in undisturbed old-growth  
348 forest in terms of species diversity, which was hypothesised. They only differed in terms of the  
349 number of tree individuals and community-weighted mean WD. These findings are in line with  
350 other studies in the Congo Basin (Bauters *et al.* 2019, Makelele *et al.* 2021) and in other tropical  
351 regions (Letcher & Chazdon 2009, Mahayani *et al.* 2020). In contrast to species diversity, we  
352 found that the community composition in plots in regrowth forest was significantly different  
353 from those in undisturbed old-growth forest. These findings are in line with what was expected  
354 and might suggest that community composition needs more time to converge in similarity to  
355 plots in undisturbed old-growth forest. Similar to our results, the community composition of  
356 young regrowth forest (25-30 years) differed from old-growth forest in the Maringa-Lopori-



357 Wamba forest landscape in DR Congo, whereas old regrowth forests (150-300 years) were very  
358 close to old-growth forests (Bauters *et al.* 2019). Likewise, a recent study in the Yoko forest  
359 reserve, at approximately 30 km Southeast of Kisangani, showed that the community  
360 composition of 60-year-old regrowth forests was still significantly different from local old-  
361 growth forests (Makelele *et al.* 2021). Our results are furthermore in line with studies in the  
362 Neotropics (Oberleitner *et al.* 2021, Poorter *et al.* 2021) and tropical Asia (Hayward *et al.* 2021).

363         Despite having a similar community composition as plots in disturbed old-growth forests,  
364 plots in regrowth forests were characterised by a different indicator species assemblage,  
365 including for example *Myrianthus arboreus*, a late pioneer species (PROTA 2021). Additionally,  
366 *Combretum lokele* and *Dacryodes edulis* were also characteristic for regrowth forests. Both  
367 species are commonly found in villages and are indicative for the important anthropogenic  
368 influence in these forests (Meunier *et al.* 2015).

369         The forest structure in the plots in regrowth forests was comparable to plots in disturbed  
370 old-growth forest, but significantly different from plots in undisturbed old-growth forest. These  
371 differences were not caused by the lower abundance of large-sized trees in regrowth forests, as  
372 would be expected because these are slower to recover (Bauters *et al.* 2019, Makelele *et al.*  
373 2021). On the contrary, the observed differences can be explained by the higher frequency of  
374 small-sized trees in plots in undisturbed old-growth forests. This possibly implies that the tree  
375 regeneration could be hampered. Similar to other studies (Bauters *et al.* 2019, Makelele *et al.*  
376 2021), we suggest that forests in the Yangambi landscape initially recover rapidly to a close  
377 canopy forest with shade-tolerant understorey species following agricultural abandonment, but  
378 that the recruitment of primary forest species is slow, and that further forest succession may be  
379 hampered by ongoing anthropogenic disturbance in these regrowth forests.

### 380 **4.3 Limitations**

381 We acknowledge that our study is not flawless and that there are some limitations that have to be  
382 considered. Our plots are to a certain extent spatially clustered, which was not possible to avoid  
383 due to the location of the disturbance zone in the Yangambi area. Furthermore, due to reasons of  
384 feasibility, it was not possible to incorporate additional control plots to account for potential spatial  
385 variability (Davies & Gray 2015). The Mantel test was used to test the effect of geographic distance  
386 on the dissimilarity of the tree community composition. It should be noted that the Mantel test  
387 showed a significant correlation between geographic distance and dissimilarity of community  
388 composition. This distance decay of similarity is commonly reported in tropical forests (Ganivet  
389 *et al.* 2020, Liebsch *et al.* 2008, Wittmann *et al.* 2006). Pre-existing spatial variation seems to  
390 confound with the main treatment (logging), but the correlation is rather weak. Although we cannot  
391 conclude that the observed differences between sites are a direct effect of selective logging per se,  
392 we can interpret the results in light of our ecological understanding of the area and its activities  
393 (Oksanen 2004, Davies & Gray 2015).

### 394 **4.4. Conclusion**

395 We found that the tree community composition and forest structure were significantly different  
396 between forest categories and that the response of species diversity was more variable. Based on  
397 our knowledge of the activities in the Yangambi region, our results suggest that these differences  
398 in community composition found between UO forest and DO forest can be attributed to selective  
399 logging. Our data furthermore suggest that 40 to 60 years after agricultural abandonment, species  
400 diversity is able to recover, but community composition and forest structure are not. Given the  
401 limitations of our sampling design, we suggest vegetation surveys across larger geographic areas  
402 to deal with the spatial clustering of our survey plots within one disturbance category. Based on

403 the known correlation between forest productivity and SLA, and between carbon storage and  
404 WD, our results also indicate that disturbed old-growth forest and regrowth forest are as  
405 productive as undisturbed old-growth forest but are not able to store as much carbon (Chave *et*  
406 *al.* 2009, Madani *et al.* 2018). Conservation of undisturbed old-growth forests is vital, although  
407 also regrowth forests can play a crucial role in maintaining biodiversity and providing ecosystem  
408 services (Crouzeilles *et al.* 2016, Van de Perre *et al.* 2018).

#### 409 **FIGURE LEGENDS**

410 Figure 1 Location of the 25 forest inventory plots across the three forest categories in the  
411 Yangambi region (DR Congo).

412 Figure 2 Comparisons between tree abundance and tree species diversity metrics across old-  
413 growth undisturbed forest (OU), old-growth disturbed forest (OD), and regrowth forest (R) in the  
414 Yangambi region (DR Congo). Hinges represent the 25th, 50th and 75th percentiles,  
415 respectively. Whiskers extend to maximum 1.5 times the interquartile range. Letters code for  
416 significant differences between groups. (a) tree abundance, (b) species richness, (c) species  
417 diversity, (d) evenness.

418 Figure 3 Non-metric multidimensional scaling (NMDS) ordination of 25 forest inventory plots in  
419 the Yangambi region (DR Congo) across the three forest categories: undisturbed old-growth  
420 forest (UO), disturbed old-growth forest (DO), and regrowth forest (R). Ordination is based on  
421 the Bray-Curtis distance matrix calculated from the tree species abundances and represents the  
422 woody species (DBH  $\geq$  5 cm) community composition.

423 Figure 4 Comparisons between the community weighted means of specific leaf area (SLA) and  
424 wood density (WD) across the three forest categories. Hinges represent the 25<sup>th</sup>, 50<sup>th</sup> and 75<sup>th</sup>

425 percentiles, respectively. Whiskers extend to maximum 1.5 times the interquartile range. Letters  
426 code for significant differences between groups.

427 Figure 5 Observed frequencies among the ten DBH size categories compared among the three  
428 forest categories. Size classes one to nine have a 10 cm interval, while the tenth has a wider  
429 interval, grouping individuals with a DBH of over 95 cm.

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## 440 **CONFLICT OF INTEREST**

441 All authors confirm that there is no conflict of interest regarding the publication of this article.

## 442 **AUTHOR CONTRIBUTIONS**

443 OH, FV and JD designed this study. JD, JA, RM, YH and JK participated in fieldwork. JD  
444 analysed the data. FV, OH and JD wrote the manuscript. All authors contributed to finalising the  
445 manuscript.

446 **DATA AVAILABILITY STATEMENT**

447 The data that support the findings of this study will be made openly available in Zenodo

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757 **SUPPORTING INFORMATION**

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759 Additional supporting information can be found online in the Supporting Information section at  
760 the end of the article.

761 APPENDIX S1 Plant trait values per species

762 APPENDIX S2 Plot level data

763 APPENDIX S3 Indicator species per forest category