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2 **Main Manuscript for**

3 Megaherbivores modify forest structure and increase carbon stocks through multiple pathways.

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30 **Abstract**
31

32 Megaherbivores have pervasive ecological effects. In African rainforests, elephants can increase
33 aboveground carbon, though the mechanisms are unclear. Here we combine a large
34 unpublished dataset of forest elephant feeding with published browsing preferences totaling >
35 120,000 records covering > 700 plant species, including nutritional data for 145 species.
36 Elephants increase carbon stocks by: 1) promoting high wood density plants via preferential
37 browsing on leaves from low wood density species, which are more palatable and digestible; 2)
38 dispersing seeds of trees that are relatively large and have the highest average wood density
39 among tree guilds based on dispersal mode. Loss of forest elephants could cause a 6-9%
40 decline in aboveground carbon stocks due to regeneration failure of elephant-dispersed trees
41 and an increase in abundance of fast-growing low wood density plants. These results show the
42 major importance of megaherbivores in maintaining diverse, high-carbon tropical forests.
43 Successful elephant conservation will contribute to climate mitigation at a scale of global
44 relevance.

45 **Significance Statement**

46 Very large herbivores (body mass >1000 kg), also known as megaherbivores, can significantly
47 influence the structure and functioning of ecosystems. Most of our knowledge on the ecological
48 role of megaherbivores is based on the African savanna; much less is known about forest-
49 dwelling megaherbivores. We show that forest elephants can promote higher aboveground

50 carbon through browsing preferences and seed dispersal. Forest elephants browsing promotes
51 high carbon density plants through the consumption of less carbon-dense plants. Elephant-
52 dispersed trees are larger and have higher densities of carbon compared to trees with other
53 dispersal modes. These results highlight the importance of forest elephants and other
54 megaherbivores for maintaining biodiversity and high carbon stocks in tropical forests.

55

56 **Main text**

57 **Introduction**

58 Megaherbivores (body mass > 1000 kg) can have profound effects on vegetation, carbon stocks,
59 and nutrient cycling¹⁻³. However, knowledge on the ecosystem role of megaherbivores comes
60 predominantly from African savanna ecosystems^{1,4}. In tropical forests, initial evidence suggests
61 that these large herbivores might also have profound effects^{2,5-7}. Until the late Pleistocene,
62 tropical forests hosted a variety of large- and megaherbivores playing critical roles in seed
63 dispersal networks and in the dynamics and functioning of tropical forest communities due to
64 their unique combination of ecological traits^{2,8,9}. Today, Asian elephants (*Elephas maximus*) and
65 African forest elephants (*Loxodonta cyclotis*) are the only forest-dwelling megaherbivores with
66 extensive ranges and unique ecological traits: large size, diverse behaviors, and highly varied
67 diets. Examples of “ecosystem engineering” have been observed in forest elephants
68 (“elephants”) through seed dispersal^{6,10} and disturbance, which includes consumption,
69 breakage, and trampling¹¹⁻¹³. Results from a process-based vegetation model suggested that
70 by reducing tree density, elephants could promote the growth of larger trees with consequent
71 drop in light and water availability in the understory. As a result, forests with elephants hold more

72 aboveground carbon (AGC) because of a greater abundance of large late-successional tree
73 species which have high wood density (WD)⁵. The same study reported that the average WD of
74 smaller trees (diameter < 30 cm) was lower compared to larger trees at a site where elephants
75 were extirpated 30 years ago. However, these results have not been validated experimentally.
76 Also, Berzaghi et al. (2019) evaluated the effect of elephants in terms of a generic elephant-
77 induced mortality not associated with any particular behavior. We aim to explore empirically the
78 other pathways through which megaherbivores' interactions with ecosystems might influence
79 forest structure and AGC, notably via more delicate processes such as herbivory and seed
80 dispersal^{6,14,15}. The high daily food consumption (100-200 kg^{16,17}) and broad diet (over 350
81 species¹⁸) of elephants suggest that feeding preferences could drive shifts in tree species
82 composition by promoting growth and survival of less-desirable browse species. Folivores prefer
83 leaves high in protein and minerals and low in fiber and chemical defenses (e.g., phenolics and
84 tannins)¹⁷. Among woody plants, slow-growing shade-tolerant species invest more in structural
85 and chemical defenses compared to fast-growing gap-colonizing species¹⁹. Because WD is
86 negatively correlated with growth rate²⁰, we hypothesize that high WD species are less palatable
87 and less digestible compared to low WD species. As a result, elephants would promote high
88 AGC by preferentially browsing leaves of low WD plants.

89 We also investigate the connection between trees dispersed exclusively by elephants
90 ("Obligate" trees *sensu*⁶) and AGC. Large-seeded animal-dispersed trees have relatively large
91 diameters, high WD, and contribute significantly to AGC²¹. Forest elephants are prodigious seed
92 dispersers, moving more seeds of more species than any other animal species⁶, but the
93 contribution of Obligate trees to forest structure and AGC has not been evaluated. We
94 hypothesize that the combined effects of elephant browsing, which *decreases* fitness of

95 preferred food species, and seed dispersal, which *increases* fitness of dispersed species, are
96 likely to have profound effects on forest structure and AGC. If supported, these two hypotheses
97 would confirm the ecosystemic role of elephants in promoting high carbon stock forests by
98 increasing the fitness of large, high WD trees⁵. To test these hypotheses, we combined forest
99 inventories and elephant feeding data collected in Ndoki (Republic of Congo) and forest
100 inventories in LuiKotale (Democratic Republic of Congo) with published diet preferences data
101 across the Afrotropics. We analyzed nutritional properties of leaves and fruit as a function of
102 elephant browsing preferences and WD to investigate the mechanisms driving elephant feeding
103 choices and the influence of these choices on AGC. We then synthesized, based on literature,
104 quantitative measures of the effects of elephants on forest properties and processes and
105 schematically organized these findings. This synthesis identifies research gaps and provides
106 input for modeling the ecological impacts of elephants using statistical and process-based
107 models. Our results greatly enhance our understanding of the contribution of elephants to forest
108 functioning and are key to evaluating the consequences of past megaherbivore extinctions and
109 to informing conservation and management policy.

110

111 **Results**

112 **Nutritional properties influence elephant food choices**

113 We investigated the mechanisms that influence food preferences by combining data from a
114 global database of plant nutritional values²² with forest elephant feeding data from seven
115 different sites across tropical Africa: West (N = 4), Central (2), and East (1 site from two separate
116 studies). The nutritional data covered 145 plant species and 1343 records of essential
117 biomolecules (crude proteins, minerals, fat, and carbohydrates), structural and defensive

118 compounds (fibers, tannins, and phenols), which reduce food palatability and digestibility (% of
119 assimilated food), and gross energy. Results from the ordinal logistic regression suggest that
120 elephants prefer to feed on leaves low in fibers, tannins, and phenols (Fig. 1A). We found
121 moderate to strong evidence that less-preferred species are more likely to contain higher
122 percentage of these defensive compounds compared to highly preferred plants ($p = 0.004$ -
123 0.063). As the percentage of defensive compounds of leaves increases from 0-10% to 30-70%,
124 the probability of a tree to be least preferred by elephants increases from ~37% to ~65-90%,
125 depending on the defensive property (Fig. 1A). Protein and minerals increase the probability of
126 a tree species being in the high-preference category compared to medium preference ($P \leq$
127 0.001), whereas nonstructural carbohydrates and fat decrease this probability ($P \leq 0.05$). We
128 found no consistent pattern across preference groups in the other nutritional properties of leaves
129 (Figs. 1A and S1). It appears that the main determinants of elephant preferences are defensive
130 structural properties rather than essential biomolecules. These choices do not appear to be
131 dictated by the relative abundance of plant species. Both at Ndoki NP and Kibale NP, very strong
132 evidence suggests that preference is inversely correlated to availability with very similar slopes
133 and fit in the linear model of both sites (Fig. S2, $R^2 = 0.51$ - 0.62 , $P < 0.001$). The analysis of fruit
134 properties consumed and not consumed by elephants also provides very strong evidence that
135 elephant-consumed fruits are larger ($P < 0.0001$) and contain more fiber and minerals ($P < 0.01$)
136 compared to non-consumed fruits (Fig. 1B). We found moderate evidence for differences
137 between the two groups in total tannins, sugars, and fat (Fig. 1B, $P < 0.05$), the latter being the
138 only property that was higher in non-consumed fruit. No statistical differences were found in
139 crude tannins and proteins across the two groups. At Ndoki, the analysis of the relationship
140 between total diameter at breast height (DBH), a possible proxy for total fruit availability, and

141 percentage of detection in dung did not reveal any significant relationship (Fig. S3A). However,
142 the correlation between presence of seeds in dung and fruit volume revealed a positive
143 correlation (Fig. S3B, $R^2 = 0.16$, $P = 0.003$). This may suggest that larger fruit are selected for
144 their size and not for their abundance. However, detectability of seeds during dung sampling
145 might be biased by other factors (see discussion). Overall, the analysis of nutritive properties
146 suggests that elephant seek more palatable and less fibrous leaves and large fruits, which are
147 also high in sugars and minerals but have the drawback of being more fibrous (i.e., less
148 digestible) and less fatty.

149 The comparison between nutritional properties of fruits and leaves consumed by elephants
150 revealed very strong evidence that, compared to fruit, leaves contain roughly twice as much
151 protein ($P < 0.0001$) and minerals ($P < 0.0001$), but six times less sugars and non-structural
152 carbohydrates ($P < 0.0001$) (Figs. 1C and S4). Moderate evidence suggests that fruits contain
153 less phenols and more protein compared to leaves. There were no significant differences in
154 gross energy content between the two organs (Fig. 1C). Thus, fruits provide short-term usable
155 energy, although gross energy content is similar to leaves, which instead contain biomolecules
156 useful for longer-term physiological processes. Forest elephants are making deliberate choices
157 that are mostly independent of plant availability and more dependent on nutritional or
158 morphological (i.e., fruit) properties.

159

160

161 **Wood density is related to nutritional quality of leaves and fruits**

162 We then investigated if WD is a good predictor of leaf and fruit nutritional properties. Results
163 from linear regression models revealed strong to very strong evidence that for leaves, WD is
164 positively correlated with fibers ($R^2 = 0.11$, $P < 0.001$) and phenolics ($R^2 = 0.12$, $P = 0.007$) (Fig.
165 2A). There was weak to moderate evidence of a negative correlation between WD and fat ($R^2 =$
166 0.14 , $P = 0.025$) and a positive correlation with gross energy ($R^2 = 0.07$, $P = 0.057$) and tannins
167 ($R^2 = 0.04$, $P = 0.084$). We found no evidence of a correlation between WD and protein or
168 minerals, and little to weak evidence of a negative correlation with total nonstructural
169 carbohydrates (Fig. 2A and S5). Whereas leaves from higher WD plants had higher content of
170 structural and defensive properties, we found strong evidence that fruits from higher WD plants
171 were lower in essential biomolecules compared to fruits produced by lower WD species (Fig. 2).
172 In particular, we observed a negative correlation between WD and minerals, proteins, and fat
173 ($R^2 = 0.07$ - 0.15 , $P < 0.001$) and a positive correlation with sugars ($R^2 = 0.16$, $P = 0.014$) (Fig.
174 2B). Note however that total nonstructural carbohydrates (sugars plus starches) did not show
175 any significant correlation with WD (Fig. S5B). The data revealed weak to moderate evidence of
176 WD being positively correlated with fruit phenols and fiber content ($R^2 = 0.03$ - 0.10 , $P = 0.043$ -
177 0.085) (Fig. 2B). No other statistically-significant relationship was found for the other nutritional
178 properties (Fig. S5B).

179 **Forest elephants browse most frequently on low wood density species**

180 The elephant feeding data and browsing preferences included 197,557 feeding records from
181 730 plant species for which WD could be determined (eight total studies, Table S1). The actual
182 number of feeding records is higher because three studies did not report their total sample size
183 (Table S1). At all sites, except Bia National Park and Santchou Wildlife Reserve, feeding

184 preference metrics were reported in terms of relative preference for single species in relation to
185 all consumed plants. These metrics could be assimilated into three groups indicating high,
186 intermediate, and low preference (Methods). Data from Ndoki included number of feeding events
187 and quantity consumed, and, along the two sites mentioned above, were not included in the
188 global analysis. Results from the ordinal logistic regression provided strong evidence that
189 globally the probability of a species being more preferred decreases as WD increases across all
190 and between low and medium groups (Fig. 3, $P = 0.003-0.027$); in the aggregated model, the
191 average WD was systematically lower in higher preference groups but statistically different only
192 between the low and high groups (Fig. 3, T-test - $P < 0.01$). At Ndoki, the only site where both
193 feeding frequency and quantity consumed were available, we also found very strong evidence
194 that higher probability of preference was associated with decreasing WD (Fig. 3, $P < 0.0001$).
195 We found very significant differences between medium and high (Ordinal regression, $P < 0.0001$)
196 and low and high (T-test, $P < 0.0001$) preference groups at Ndoki. At Tai and Kibale (1996 study)
197 similar patterns were found with strong to weak evidence, respectively (Fig. S6). Data from Bia
198 provided moderate evidence that the probability for high preference increases with WD ($P =$
199 0.026). However, this study did not report site-relative preferences and many species were only
200 recorded as browsed once, resulting in a large number of species in the low preference group
201 (see Methods and discussion for further considerations). At Satchou Wildlife Reserve, there was
202 no correlation between preference rank and WD based on 16 species (Fig. S6). In five out of
203 seven studies, there was moderate to very strong evidence that the high preference group had
204 the lowest WD compared to the intermediate and/or low preference groups (Fig. 3 and S6). Only
205 at Bossematie there was no significant trend. The aggregated sites, Ndoki, and Tai analyses
206 provided moderate to very strong evidence that the WD of trees dispersed by elephants was

207 higher compared to the WD of species browsed by them (Fig. 3). No other statistical differences
208 between WD of fruit and browse were found at Bia where comparisons were possible. Overall,
209 these results are compatible with the hypothesis that elephants increase tree community WD
210 through dispersal of higher WD species and browsing of lower WD species.

211 Our data from Ndoki (understory and overstory) and LuiKotale (overstory) showed moderate to
212 strong evidence that high WD species are slightly more abundant than low WD ones with
213 abundance measured as a percentage of total stems or total DBH ($R^2 = 0.05-06$, $P = 0.003-$
214 0.019); the linear model explains however only a small fraction of the variability (Fig. S7-S8). As
215 shown previously for nutritive properties of leaves, elephants seem to make specific choices
216 regardless of the abundance of species. For example, in Ndoki understory vegetation plots,
217 *Rinora welwitschii* and *Diospyros bipindensis* were recorded 559 and 468 times respectively
218 (from a dataset of 6548 tree stems from at least 151 species, see Methods). Yet, of 5458 feeding
219 events, only two involved *D. bipindensis* and *R. welwitschii* was never browsed.

220 **Elephant-dispersed trees are larger and have higher wood density compared to trees** 221 **with other dispersal modes**

222 We identified five dispersal modes in Ndoki and LuiKotale: Gravity/dehiscence (GD), wind,
223 elephants and other animals (Non-Obligate), elephants (Obligate), and Other-Animals^{6,23} (total
224 of 307 species, complete list in Dataset S1). The analysis of the variation of WD as a function of
225 dispersal mode revealed that Obligate species had the highest average WD in both sites (Fig.
226 4A). However, at LuiKotale, only GD, Obligate and Non-obligate species had statistically
227 different WD ($P \leq 0.05$). At Ndoki, data showed moderate to very strong evidence that Obligate
228 species had statistically higher WD compared to Non-obligate, Other animal and wind dispersed
229 but not GD species. Wind-dispersed (Ndoki) and Non-obligate (LuiKotale) species had the

230 lowest WD. The total number of species identified at LuiKotale (n = 103) was almost half that of
231 Ndoki (n = 204), we thus urge more caution when interpreting these results from LuiKotale.

232 The distribution of stem size classes by dispersal mode was similar at the two sites (Fig.4B).
233 Obligate and wind-dispersed tree communities were characterized by few smaller trees, a higher
234 number of larger trees, and are overrepresented in the 125-250 cm range compared to trees
235 with other dispersal modes (Fig. 4B). Obligate trees represent the largest proportion of stems
236 with diameter > 150 cm in LuiKotale (35-75%) and second largest of stems with diameter 125-
237 225 cm in Ndoki, 18-38%, depending on size class. The proportion of wind-dispersed trees also
238 increased with size class at both sites. Gravity-dispersed, Non-obligate and Other-animals trees
239 are most abundant in the lower size classes between 40-125 cm (Fig. 4B). This distribution of
240 stem size across dispersal modes might reveal the long-term history of these forests emerging
241 after the recolonization of savannas as late as 800-250 year ago²⁴.

242 **Contribution of elephant-dispersed trees to aboveground carbon**

243 The distribution of aboveground carbon in trees (diameter \geq 40 cm) grouped by dispersal mode
244 reveals diverse patterns in Ndoki and LuiKotale (Fig. 5). In Ndoki, AGC is more evenly distributed
245 among dispersal modes. Abiotically-dispersed trees account for ~50% and Obligate for ~15% of
246 AGC. In LuiKotale, trees dispersed by Other-animals store 52% of AGC and ~19% is stored in
247 Obligate trees (second largest biomass pool). At Ndoki, our sampling of vegetation was slightly
248 biased toward the monodominant *Gilbertiodendron dewevrei* forest, which occupies a proportion
249 of Ndoki¹⁸ following watercourses, as do forest elephants. If *G. dewevrei* forest was removed
250 from the analysis, and only mixed species *terra firma* forest considered, the contribution of
251 Obligate species would increase by ~2-3% in DBH classes > 40 cm and > 70 cm, respectively.
252 When considering only larger trees (diameter \geq 70 cm), the percentage of AGC stored increases

253 in Obligate (23% LuiKotale and 19% Ndoki) and abiotically-dispersed trees (57% Ndoki) (Fig.
254 5). Notably, at both sites the few Obligate species have the highest relative contribution to AGC,
255 despite their low stem count (i.e. highest AGC to stem ratio represented by bar widths in Fig.
256 5). This is explained by their high WD, highest AGC per cm of diameter, and high relative
257 abundance in the large size classes (diameter > 125 cm) (Fig. 4B). The loss of forest elephants
258 likely greatly diminishes or prevents the recruitment of Obligate trees in addition to negatively
259 affecting Non-Obligate species^{6,23}. We quantified the loss of AGC by simulating a replacement
260 of Obligate trees with random trees with other dispersal modes proportionally to their relative
261 total DBH (Methods). The loss of AGC was estimated to be 9.2% (s.d. \pm 0.07) at LuiKotale and
262 5.8% (s.d. \pm 0.02) at Ndoki. Thus, the “other” trees cannot completely compensate the
263 contribution of Obligate trees to AGC. The important role of large trees in AGC^{25,26} and the
264 widespread decline of forest elephants make the plight of Obligate species critical for the future
265 of AGC in African tropical forests.

266

267 **Ecological processes influenced by elephants**

268 Both savanna and forest elephants are the largest megavertebrates in their respective
269 ecosystems, and there are similarities in their ecological roles in the physical and trophic
270 structuring of ecosystems. The effects of savanna elephants on their environment have been
271 heavily studied¹, yet few studies have quantified the impacts of forest elephants. We synthesized
272 the literature and selected studies that provided quantifiable measures of the mechanisms of
273 ecosystem engineering by elephants expressed in terms of rates, equations, or data. Of all the
274 possible ecological processes influenced by elephants²⁷, only a few have been quantified and
275 most of them only once or twice. Many other studies exist on seed dispersal or browsing

276 preferences but we could not quantify, or generalize with equations, their consequences on
277 ecosystem properties. Savanna and forest elephants alike topple small trees to access foliage,
278 scar and debark trunks but the impacts of these foraging effects on mortality in forests are poorly
279 quantified (Table 1). Data on debarking and scarring and forest properties (forest openness,
280 stem density, AGC, and WD) come from single studies (Table 1). Only one study quantified
281 forest properties as a function of elephant trails²⁸. However, a few general conclusions can be
282 drawn from our synthesis. The annual mortality rate inflicted by elephants to large trees (DBH
283 >10 cm) is between 1-2% which is similar to the background mortality of African tropical forests²⁰.
284 The mortality of seedling and saplings is several times higher compared to large trees. Distance
285 from trails is a key parameter when assessing the effect of elephants on forest properties. There
286 is also a clear relationship between canopy openness, reduced regeneration, and elephant
287 preference, however this has not been estimated in more quantitative terms such as visitation
288 frequency or biomass consumption. Less robust conclusions can be drawn on forest elephant
289 impacts on the density of small trees and the mortality rate of large trees due to debarking.

290

291 **Discussion**

292 We have shown that elephant browsing preferences are likely driven by leaf nutritive values
293 rather than plant abundance. Low WD and frequently-browsed plants produced more digestible
294 leaves containing less structural and defensive properties. Fruit preferences were tightly
295 associated with fruit-size and possibly mineral and sugar content; fruit from high WD plants
296 appeared to contain more sugars and less fat, which are important nutritional properties in fruits.
297 Seed surveys in dung do not provide a quantitative measure to completely assess if fruit
298 preferences are also affected by fruit abundance. Across their range, African forest elephants

299 browse most frequently on tree species with low WD and consume fruit from high WD species
300 (Fig. 3). Our results at Ndoki accounted also for quantity consumed, a critical parameter for
301 assessing browsing preferences, and confirmed this general trend. The exception of Bia NP
302 could be due to the history of the forests in this park as it was extensively logged around the
303 middle of the 20th century. Bia NP has abundant presence of woody lianas and climbers both in
304 the forest and in elephant diets (more than 60% of all species consumed²⁹) compared to the
305 other side where trees dominate the diet. Probably the intense opening of the canopy changed
306 drastically the composition of plant species. These results strongly support our hypothesis that
307 elephant browsing increases the AGC of central African forests by reducing the fitness of
308 preferred fast-growing species and by promoting the survival and growth of slow growing high
309 WD species. Previous studies also suggested that if elephants are extirpated from forests, the
310 community average WD declines and composition might shift towards an alternative state
311 dominated by lianas, fast-growing and abiotically-dispersed species^{5,6,30}. The slightly higher
312 abundance of high WD species and low abundance of elephant-preferred plants might suggest
313 that elephants contribute in keeping a balance between low and high WD species within the
314 forest. Another study reported that the abundance of elephant-preferred species did not decline
315 overtime³¹, which might explain why elephant-preferred species remain at low abundance but
316 do not disappear.

317 Elephants also influence AGC by dispersing seeds of high WD tree species which are also over-
318 represented in large sizes (Fig. 4). The reason for a higher relative abundance of Obligate trees
319 in larger size classes is unclear, but may be due to the combination of life history traits of large
320 seeded species, phylogenetic signal, and forest succession history. Wood density is correlated
321 with structural strength, low mortality, and resistance to decay which favor large size and

322 longevity (though slow growth means that attaining large size takes longer for these species)^{32,33}.
323 However, some of the largest trees in the forest are also fast growing, wind dispersed species
324 of low WD (e.g. *Triplochiton scleroxylon* and *Ceiba pentandra*). Whatever the underlying reasons
325 for their large size and high WD, Obligate trees contribute significantly to AGC. Declines in
326 abundance or the complete extirpation of forest elephants will therefore reduce recruitment⁶ and
327 result in an important reduction in AGC, estimated at 6-9% at our two study sites. Many other
328 Non-Obligate tree species might also experience reduced recruitment rates because elephants
329 contribute to a large proportion of their seed dispersal⁶.

330 The current knowledge base on the processes and properties of forest that are influenced by
331 elephants is better developed in the early stages of plant development (Table 1). Elephant
332 density (individuals/km²) is not always reported; this parameter is useful when evaluating the
333 magnitude of elephant effects on forest properties and processes, particularly when
334 extrapolating results to other areas. We suggest that studies should report the equations of fitted
335 regressions, which would be useful for modelling approaches. In light of the important relations
336 between nutritive properties, feeding preferences, and WD, we also suggest that feeding studies
337 of forest-dwelling herbivores also consider these plant properties. These data will help better
338 understand the contribution of species to tropical forests carbon cycling. Many processes and
339 properties have received less attention or have been evaluated in more qualitative terms. For
340 example, trampling, unrooting, and other mechanical non-feeding processes may have profound
341 effects on forest structure, light regimes, soil compaction, etc.³⁴ . There is a lack of repeated
342 experiments in different sites to verify if locally-observed effects are consistent across sites and
343 to evaluate the relation between elephant density and the magnitude of the observed change.

344 Yet the current knowledge provides a good starting point to better characterize elephant effects
345 in modeling studies.

346 Our results add further evidence that megaherbivores contribute to enhance AGC in tropical
347 forests through a variety of mechanisms. Until the late Pleistocene, many large herbivores
348 inhabited Amazonian and southeast Asian tropical forests and could have had a significant effect
349 on the functioning of those ecosystems^{2,4,7}. The low protein and mineral content of fruit might
350 limit the maximum body mass attainable by obligate frugivores, who might not be able assimilate
351 enough of these nutrients to sustain all bodily functions over the long-term¹⁷. This information
352 might help to estimate diet composition of (extinct) forest-dwelling megaherbivores based on
353 maximum daily dry matter intake and nutritional requirement¹⁷.

354 The consequences of the loss of elephants we describe on AGC will take place over multiple
355 generations of trees. However, the combination of climate change, extreme climatic events (e.g.,
356 droughts) and land use change are occurring over years and decades and could accelerate
357 changes in AGC. Logging is systematically removing the largest elephant- and wind dispersed
358 (*Entandrophragma* spp.) trees across the entire Congo Basin outside of protected areas. When
359 they are not being hunted, forest elephants preferentially use light gaps in secondary forest
360 because they can find abundant secondary, fast growing, vegetation. If a functional elephant
361 population fills disturbed areas and light gaps with elephant-dispersed high WD species, then
362 the carbon sequestration begins immediately; if the gaps are filled with fast-growing species the
363 opportunity is lost. Maintenance of forest elephants in logged and naturally-disturbed forest will
364 be critical in an immediate wildlife-driven mitigation response to climate change^{35,36} by
365 encouraging the regeneration of high WD and removal of low WD species. Process-based
366 vegetation models based on our findings and the processes shown in Table 1 will help better

367 estimate the time scale and long-term consequences of elephants decline or repopulation³⁷. The
368 significant contribution of forest elephants to carbon stocks and biodiversity should be accounted
369 to prioritize conservation of the species and their habitat, implemented in climate change
370 mitigation policy, and leveraged to promote and finance nature-based solutions in tropical
371 Africa^{35,38}.

372

373 **Material and Methods**

374 *Study sites*

375 The Ndoki Forest (“Ndoki” 1.5–3° N, 16–17° E) lies in the northern Republic of Congo. The
376 climate is transitional between the Congolo-Equatorial and sub-equatorial zones with a mean
377 annual rainfall of ca. 1400 mm (Ndoki Forest records)^{6,18}. Topography varies from *terra firma*
378 uplands and flat plateaus to the northwest to the extensive floodplain of the Likouala aux Herbes
379 River to the southeast. Soils are of three main types: arenosols to the north and west, ferrasols
380 to the southeast in the Likouala aux Herbes basin on *terra firma*, and gleysols in the swamps
381 further southeast. Ndoki is embedded in wet Guineo-Congolian lowland rainforest, transitioning
382 to the north into dry Guineo-Congolian lowland rainforest, and into swamp forests to the south.
383 *Terra firma* is dominated by Sterculiaceae-Ulmaceae semi-deciduous forest, with large patches
384 of mono-dominant *Gilbertiodendron dewevrei* forest along watercourses and upland plateaus,
385 and swamp forests¹⁸. The Ndoki fauna includes several large charismatic species such as forest
386 elephants, western lowland gorillas (*Gorilla gorilla gorilla*), common chimpanzees (*Pan*
387 *troglodytes troglodytes*), forest buffalo (*Syncerus caffer nanus*), bongo (*Tragelaphus eurycerus*),

388 and leopards (*Panthera pardus*). The human population density is low (<1 inhabitant/ km²) and
389 the immediate study area contains no permanent human settlement.

390 The LuiKotale research site is located within the equatorial rainforest (2°470S, 20°210E), at the
391 south-western fringe of Salonga National Park in the Democratic Republic of the Congo²³. The
392 study site covers >60 km² of primary evergreen lowland tropical forest. The climate is equatorial
393 with abundant rainfall (>2000 mm/yr) and two dry seasons, a short one around February and a
394 longer one between May and August. Mean temperature at LuiKotale ranges between 21 °C
395 and 28 °C with a minimum of 17 °C and a maximum of 38 °C (2007–2010). Two major habitat
396 types can be distinguished. The dry (*terra firma*) forest and the wet temporarily and permanently
397 inundated forest. The dry habitat dominates with heterogeneous species composition covering
398 73% and patches of mono-dominant *Gilbertiodendron* spp. covering 6% of the site. The wet
399 habitat consists of heterogeneous forest temporarily (17%) and permanently (4%) inundated²³.
400 The LuiKotale fauna includes several large species such as elephants (almost extinct), bonobos
401 (*Pan paniscus*), forest buffalo, bongo (*Tragelaphus eurycerus*), and leopards (*Panthera pardus*).
402 Similarly to Ndoki, the human population density is low (<1 inhabitant/ km²) and the immediate
403 study area contains no permanent human settlement.

404 *Elephant food selection at Ndoki*

405 Fresh elephant trails were followed opportunistically over the course of two years in Ndoki across
406 a range of habitat types including permanent swamps, seasonally inundated forests, and *terra*
407 *firma* open and closed canopy forest. In the case of woody species, a single feeding event was
408 defined as all fresh feeding signs on an individual plant, regardless of plant parts consumed,
409 though all parts consumed were also recorded. At each feeding site data were collected on
410 location (using a handheld Garmin GPS) estimated age (fresh [<24 hrs] or recent [24-48hrs]),

411 plant species, plant part consumed (leaf, stem, bark, wood, roots, etc.), estimated amount
412 consumed on a 1-4 scale (rare, few, moderate, and abundant). Five thousand six hundred and
413 forty-eight feeding events were recorded. Quantifying diet selection based on secondary
414 evidence is open to several sources of bias – for example, one cannot detect a feeding event of
415 a sampling that was completely consumed. In an attempt to reduce and standardize observation
416 bias, we quantified feeding events based on identifiable remains (e.g. a terminal branch stripped
417 of its leaves) in close association with fresh elephant prints, and feel confident we captured gross
418 trends on diet selection.

419 Over a 3-yr period throughout the Ndoki Forest, the seed content of 855 piles of fresh intact
420 elephant dung was quantified. Dung piles were broken apart with sticks, and fibers were
421 thoroughly teased apart. In each dung pile, all seeds were identified to species and the
422 percentage of presence of each species was calculated based on all sampled dung piles..

423 *Elephant feeding preference data*

424 Our data of forest elephant feeding preferences at Ndoki were combined with data from the
425 MegaFeed database, which contains feeding preferences of all elephant species including the
426 forest elephant. We only retained data from studies that quantified feeding preferences per plant
427 species through ordinal ranking, count of browsing events, selection ratio, or browsing
428 frequency. We excluded studies providing only a list of consumed species. Our Ndoki data and
429 the data from MegaFeed included a total of eight studies. Five out of eight studies classified
430 feeding preference in three categories: rare, medium, high. The Ndoki data contained four
431 categories that were recategorized in three by combining the rare and medium categories into
432 “low”. The remaining two studies had different data compared to the other five. The data from
433 Bia National Park²⁹ reported the number of browsing events per tree species. We assigned

434 species to three categories (low, medium, high browsing preference) based on the frequency
435 distribution of browsing events. Species with less than three browsing events were assigned to
436 the “low” category, species with more than six were assigned to the “high” category, and the
437 species in between to the “medium” category. Feeding preferences at Santchou Wildlife Reserve
438 ³⁹ were reported with an ordinal scale and thus are presented without using categories. The
439 feeding preferences at Kibale also accounted for relative abundance of elephant-preferred
440 species in relation to all plant availability. Dispersal mode of trees was determined following^{6,10}
441 and complemented with data collected at LuiKotale ²³. The Obligate elephant species were:
442 *Ochna gilletiana*, *Omphalocarpum lecomteanum*, *Omphalocarpum procerum*, *Autranella*
443 *congolensis*, *Balanites wilsoniana*, *Detarium macrocarpum*, *Drypetes gossweileri*, *Irvingia*
444 *excelsa*, *Irvingia gabonensis*, *Irvingia grandifolia*, *Irvingia robur*, *Klainedoxa gabonensis*,
445 *Mammea Africana*, *Maranthes sp.*, *Omphalocarpum elatum*, *Panda oleosa*, *Tridesmostemon*
446 *omphalocarpoides*, *Picralima nitida*, and *Strychnos aculeata*. The complete species list is
447 provided in Dataset S1. Note that not all Obligate elephant species indicated by²³ were classified
448 as such, as we found evidence in literature that some of those species can be dispersed also by
449 other animals.

450 *Tree inventory data and taxonomy harmonization*

451 Tree inventory data were collected in Ndoki (along and perpendicularly from nine large elephant
452 trails, 5674 trees DBH > 40 cm) and LuiKotale (16 1-ha plots, 6579 trees DBH >10 cm). In Ndoki,
453 1664 understory circular plots were enumerated, in which 6479 trees and shrubs were measured
454 and identified. Tree species data (browsing preference plus forest inventories) from other sites
455 spanned over several decades and species names were homogenized and updated following
456 the taxonomy provided by World Flora Online through their associated R package.

457 *Wood density data and AGC analysis*

458 We used the R package “BIOMASS” to assign WD to each feeding record starting at the species
459 level, to the genus, and finally to the site family average. If none of these were available, we
460 assigned the plot-average WD for the inventory data. Feeding data records without WD were
461 removed because the plot-average WD was not available. Aboveground carbon (AGC) was
462 calculated using the “BIOMASS” package with the following equation:

463
$$AGC = \exp(-2.024 - 0.896 * E + 0.920 * \log(WD) + 2.795 * \log(DBH) - 0.0461 * (\log(DBH)^2))$$

464 Where E is a measure of environmental stress estimated from site coordinates⁴⁰. We simulated
465 the loss of AGC due to the lack of recruitment of Elephant-obligate trees by adapting a
466 methodology used to study the consequences of changes in tree species composition on AGC⁴¹.
467 We replaced the total DBH of Obligate trees with new trees which were randomly sampled from
468 the remaining trees proportional to their total DBH. The relative total DBH of trees of each
469 dispersal mode was maintained. This process was repeated 10,000 times for each of the two
470 sites and the difference between pre- and post-replacement calculated for each iteration. The
471 mean and standard deviation of the 10,000 iterations were used to estimate the loss of Obligate
472 trees on AGC.

473

474 *Nutritional values of plants*

475 We gathered nutritional values of plant species consumed by elephants from *PNuts*, a global
476 database of plant nutritional values (Berzaghi et al. in prep). *PNuts* contained nutritional values
477 of leaves, bark, roots, fruits, and stems. However, we retained only data for leaves and fruits
478 because they were the most comprehensive and included data of several nutritional properties.
479 We selected the nutritional properties for which more data were available, these properties were:
480 crude protein (in the main text referred as “protein” or CP), acid detergent fiber (“fibers” or ADF),
481 crude tannins (“Cr. tannins” or CT), total tannins (“Tot. Tannins” or TT), total phenols (“tot.
482 Phenols” or TP), ash (“minerals”), water structural carbohydrates (“sugars” or WSC), total
483 nonstructural carbohydrates (starch + sugars, TNC), and gross energy (GE). The retained data
484 covered 1343 records (fruits and leaves) and 145 plant species included in the forest elephant
485 diet and 45 species and 346 records of fruit not consumed by elephants. Fruit volume was
486 calculated by multiplying fruit length and width found in the African Plant Database⁴². The same
487 database was used to retrieve seed length.

488 *Analysis of effects of forest elephants on ecosystems*

489 We researched the literature using Google Scholar and Web of Science to find studies
490 investigating the physical effect of forest elephants on the ecosystem. The following keywords
491 were used: “forest elephant”, “*Loxodonta cyclotis*”, “ecosystem engineering”, “ecosystem
492 engineer”, “regeneration”, “mortality”, “tree density”, “stem density”, “debarking”, and “nutrients”.
493 We also examined any relevant publication within the references cited by the articles found
494 during the systematic literature search.

495 *Statistical analyses.*

496 Linear regressions were performed with the R “stats” package function “lm”. For each regression,
497 we checked for normality of the data with Q-Q plots. The homogeneity of variances was checked
498 by examining the graphs of residuals vs. fitted values for each model. In the cases where a trend
499 in the residuals was detected, the data were log-transformed and the homogeneity of the
500 variances re-examined. The log-transformed variables are indicated in their respective figures’
501 descriptions. We used ordinal logistic regressions to analyze the association between wood
502 density or nutritional properties and browsing preferences⁴³. In our case, the ordinal logistic
503 regression allows to calculate the probability of a species being in a certain feeding group. The
504 results estimate the slope of the regression as well as the odds-ratio of being in successive
505 categories (low, medium and high preference). For each estimate, a t-value along with a p-value
506 were estimated. P-values were calculated by comparing t-value against the standard normal
507 distribution. From these models, we calculated the specific probability of an observation being
508 in each level of the ordinal category in our fitted model by simply calculating the difference
509 between the fitted values from each pair of adjacent stratified binomial models⁴³. The model
510 slope is based on the low preference group as the “focal group” being compared with the other
511 two preference groups. These analyses were performed with functions “polr” and “ggpredict”
512 from R package “MASS” and “ggeffect”, respectively. The aggregated analysis of WD across
513 preference groups on the four sites (five studies) was carried out by including a random effect
514 to account for site effect. For this ordinal logistic regression with random effect, we used the
515 function “clmm” from R package “ordinal”. In this analysis, the Ndoki, Santchou, and Bia sites
516 were removed because the methodology used to quantify elephant food preferences was
517 different from the other sites. The five studies aggregate and the Ndoki analysis were presented
518 as the main results. The aggregate result allows to discern a general pattern beyond site-specific

519 trends and the Ndoki data is ideal because it reports not only frequency of feeding but also
520 quantity, which is critical when assessing the potential effects of biomass consumption. The
521 single-site analysis of the relationships between WD and browsing preferences was conducted
522 using an ordinal logistic regression model and are presented in the supplementary. Analysis of
523 leaf nutritional properties across preference groups was performed following the same
524 procedure. Given that ordinal logistic regression models do not provide any information on
525 statistical differences between the low and high categories, we compared the means of these
526 two groups using additional t-tests. The normality of the distributions was verified and if this was
527 not verified (only in data from Ndoki) a non-parametric Wilcoxon test was used. The same
528 procedure was followed to compare the means of other categories (fruit eaten or not eaten by
529 elephants, fruit vs leaf constituents). Complete test results are included in the supplementary
530 material. We report our scientific findings by following recently suggested methodology that
531 avoids interpreting P-values with arbitrary cutoff point but instead through evidence language
532 associated to ranges of P-values⁴⁴. Evidence language include: very strong, strong, moderate,
533 weak, and little or no evidence according to P-value ranges⁴⁴.

534

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549 **References**

- 550 1. Malhi, Y. *et al.* Megafauna and ecosystem function from the Pleistocene to the Anthropocene.
551 *PNAS* **113**, 838–846 (2016).
- 552 2. Doughty, C. E. *et al.* Megafauna extinction, tree species range reduction, and carbon storage in
553 Amazonian forests. *Ecography* **39**, 194–203 (2016).
- 554 3. Sitters, J., Kimuyu, D. M., Young, T. P., Claeys, P. & Olde Venterink, H. Negative effects of cattle on
555 soil carbon and nutrient pools reversed by megaherbivores. *Nature Sustainability* **3**, 360–366
556 (2020).
- 557 4. Owen-Smith, N. Pleistocene extinctions: the pivotal role of megaherbivores. *Paleobiology* **13**, 351–
558 362 (1987).
- 559 5. Berzaghi, F. *et al.* Carbon stocks in central African forests enhanced by elephant disturbance. *Nat.*
560 *Geosci.* **12**, 725–729 (2019).
- 561 6. Blake, S., Deem, S. L., Mossimbo, E., Maisels, F. & Walsh, P. Forest elephants: tree planters of the
562 Congo. *Biotropica* **41**, 459–468 (2009).

- 563 7. Corlett, R. T. Megafaunal extinctions and their consequences in the tropical Indo-Pacific. *Terra*
564 *Australis* **32**, 117–131 (2010).
- 565 8. Pires, M. M. *et al.* Reconstructing past ecological networks: the reconfiguration of seed-dispersal
566 interactions after megafaunal extinction. *Oecologia* **175**, 1247–1256 (2014).
- 567 9. Lundgren, E. J. *et al.* Introduced herbivores restore Late Pleistocene ecological functions. *PNAS* **117**,
568 7871–7878 (2020).
- 569 10. Campos-Arceiz, A. & Blake, S. Megagardeners of the forest – the role of elephants in seed dispersal.
570 *Acta Oecologica* **37**, 542–553 (2011).
- 571 11. Terborgh, J. *et al.* Megafaunal influences on tree recruitment in African equatorial forests.
572 *Ecography* **39**, 180–186 (2016).
- 573 12. Piironen, T., Valtonen, A. & Roininen, H. Vertebrate herbivores are the main cause of seedling
574 mortality in a logged African rainforest—implications for forest restoration. *Restor Ecol* **25**, 442–
575 452 (2017).
- 576 13. Lawes, M. J. & Chapman, C. A. Does the herb *Acanthus pubescens* and/or elephants suppress tree
577 regeneration in disturbed Afrotropical forest? *Forest Ecology and Management* **221**, 278–284
578 (2006).
- 579 14. Forrister, D. L., Endara, M.-J., Younkin, G. C., Coley, P. D. & Kursar, T. A. Herbivores as drivers of
580 negative density dependence in tropical forest saplings. *Science* **363**, 1213–1216 (2019).
- 581 15. Terborgh, J. Using Janzen–Connell to predict the consequences of defaunation and other
582 disturbances of tropical forests. *Biological Conservation* **163**, 7–12 (2013).
- 583 16. Ruggiero, R. G. Seasonal forage utilization by elephants in central Africa. *African Journal of Ecology*
584 **30**, 137–148 (1992).

- 585 17. Rode, K. D., Chiyo, P. I., Chapman, C. A. & McDowell, L. R. Nutritional Ecology of Elephants in Kibale
586 National Park, Uganda, and Its Relationship with Crop-Raiding Behaviour. *Journal of Tropical*
587 *Ecology* **22**, 441–449 (2006).
- 588 18. Blake, S. The ecology of forest elephant distribution and its implications for conservation.
589 (University of Edinburgh, 2003).
- 590 19. Coley, P. D. Herbivory and defensive characteristics of tree species in a lowland tropical forest.
591 *Ecological monographs* **53**, 209–234 (1983).
- 592 20. Rozendaal, D. M. A. *et al.* Competition influences tree growth, but not mortality, across
593 environmental gradients in Amazonia and tropical Africa. *Ecology* **101**, e03052 (2020).
- 594 21. Osuri, A. M. *et al.* Contrasting effects of defaunation on aboveground carbon storage across the
595 global tropics. *Nature Communications* **7**, 11351 (2016).
- 596 22. Berzaghi, F., Ben Abdallah, A., Ratshikombo, Z. & Bretagnolle, F. PNuts: Global dataset of plant
597 nutritional values. *bioRxiv* (2022).
- 598 23. Beaune, D., Fruth, B., Bollache, L., Hohmann, G. & Bretagnolle, F. Doom of the elephant-dependent
599 trees in a Congo tropical forest. *Forest Ecology and Management* **295**, 109–117 (2013).
- 600 24. Hubau Wannes, Van den Bulcke Jan, Van Acker Joris, & Beeckman Hans. Charcoal-inferred
601 Holocene fire and vegetation history linked to drought periods in the Democratic Republic of
602 Congo. *Global Change Biology* **21**, 2296–2308 (2015).
- 603 25. Lutz, J. A. *et al.* Global importance of large-diameter trees. *Global Ecology and Biogeography* **27**,
604 849–864 (2018).
- 605 26. Slik, J. W. F. *et al.* Large trees drive forest aboveground biomass variation in moist lowland forests
606 across the tropics. *Global Ecology and Biogeography* **22**, 1261–1271 (2013).

- 607 27. Poulsen, J. R. *et al.* Ecological consequences of forest elephant declines for Afrotropical forests.
608 *Conservation Biology* **32**, 559–567 (2018).
- 609 28. Inogwabini, B.-I., Ngama-nkosi, Wema-wema & Longwango. Elephant effect on forest physical
610 structure and plant species composition in Salonga and and Malebo (Lac Tumba landscape),
611 Democratic Republic of Congo. *Pachyderm* **0**, 28–37 (2013).
- 612 29. Short, J. Diet and feeding behaviour of the forest elephant. **45**, 177–186 (1981).
- 613 30. Poulsen, J. R., Clark, C. J. & Palmer, T. M. Ecological erosion of an Afrotropical forest and potential
614 consequences for tree recruitment and forest biomass. *Biological Conservation* **163**, 122–130
615 (2013).
- 616 31. Omeja, P. A. *et al.* Changes in elephant abundance affect forest composition or regeneration?
617 *Biotropica* **46**, 704–711 (2014).
- 618 32. Kraft, N. J. B., Metz, M. R., Condit, R. S. & Chave, J. The relationship between wood density and
619 mortality in a global tropical forest data set. *New Phytologist* **188**, 1124–1136 (2010).
- 620 33. Visser, M. D. *et al.* Functional traits as predictors of vital rates across the life cycle of tropical trees.
621 *Functional Ecology* **30**, 168–180 (2016).
- 622 34. Kortlandt, A. Vegetation research and the 'bulldozer' herbivores of tropical Africa. in *Tropical rain-*
623 *forest: the Leeds symposium/edited by AC Chadwick and SL Sutton* (Leeds: Leeds Philosophical and
624 Literary Society, 1984., 1984).
- 625 35. Berzaghi, F. *et al.* Value wild animals' carbon services to fill the biodiversity financing gap. *Nat.*
626 *Clim. Chang.* 1–4 (2022) doi:10.1038/s41558-022-01407-4.
- 627 36. Malhi, Y. *et al.* The role of large wild animals in climate change mitigation and adaptation. *Current*
628 *Biology* **32**, R181–R196 (2022).

- 629 37. Berzaghi, F. *et al.* Assessing the role of megafauna in tropical forest ecosystems and
630 biogeochemical cycles – the potential of vegetation models. *Ecography* **41**, 1–21 (2018).
- 631 38. Berzaghi, F., Chami, R., Cosimano, T. & Fullenkamp, C. Financing conservation by valuing carbon
632 services produced by wild animals. *Proceedings of the National Academy of Sciences* **119**,
633 e2120426119 (2022).
- 634 39. Tchamba, M. N. & Seme, P. M. Diet and feeding behaviour of the forest elephant in the Santchou
635 Reserve, Cameroon. *African Journal of Ecology* **31**, 165–171 (1993).
- 636 40. Réjou-Méchain, M., Tanguy, A., Piponiot, C., Chave, J. & Hérault, B. biomass: an r package for
637 estimating above-ground biomass and its uncertainty in tropical forests. *Methods in Ecology and*
638 *Evolution* **8**, 1163–1167 (2017).
- 639 41. Bunker, D. E. *et al.* Species Loss and Aboveground Carbon Storage in a Tropical Forest. *Science* **310**,
640 1029–1031 (2005).
- 641 42. African Plant Database (version 4.0.0). Conservatoire et Jardin botaniques de la Ville de Genève
642 and South African National Biodiversity Institute, Pretoria. <https://africanplantdatabase.ch/>.
- 643 43. Ordinal Logistic Regression | R Data Analysis Examples. [https://stats.oarc.ucla.edu/r/dae/ordinal-](https://stats.oarc.ucla.edu/r/dae/ordinal-logistic-regression/)
644 [logistic-regression/](https://stats.oarc.ucla.edu/r/dae/ordinal-logistic-regression/).
- 645 44. Muff, S., Nilsen, E. B., O’Hara, R. B. & Nater, C. R. Rewriting results sections in the language of
646 evidence. *Trends in Ecology & Evolution* **37**, 203–210 (2022).
- 647 45. Ssali, F., Sheil, D. & Nkurunungi, J. B. How selective are elephants as agents of forest tree damage
648 in Bwindi Impenetrable National Park, Uganda? *Afr. J. Ecol.* **51**, 55–65 (2013).
- 649 46. Höft, R. & Höft, M. The differential effects of elephants on rain forest communities in the Shimba
650 Hills, Kenya. *Biological Conservation* **73**, 67–79 (1995).

- 651 47. Struhsaker, T. T., Lwanga, J. S. & Kasenene, J. M. Elephants, selective logging and forest
652 regeneration in the Kibale Forest, Uganda. *Journal of Tropical Ecology* **12**, 45–64 (1996).
- 653 48. Rosin, C. *et al.* Assessing the effects of elephant foraging on the structure and diversity of an
654 Afrotropical forest. *Biotropica* **52**, 502–508 (2020).
- 655 49. Sheil, D. & Salim, A. Forest Tree Persistence, Elephants, and Stem Scars. *Biotropica* **36**, 505–521
656 (2004).

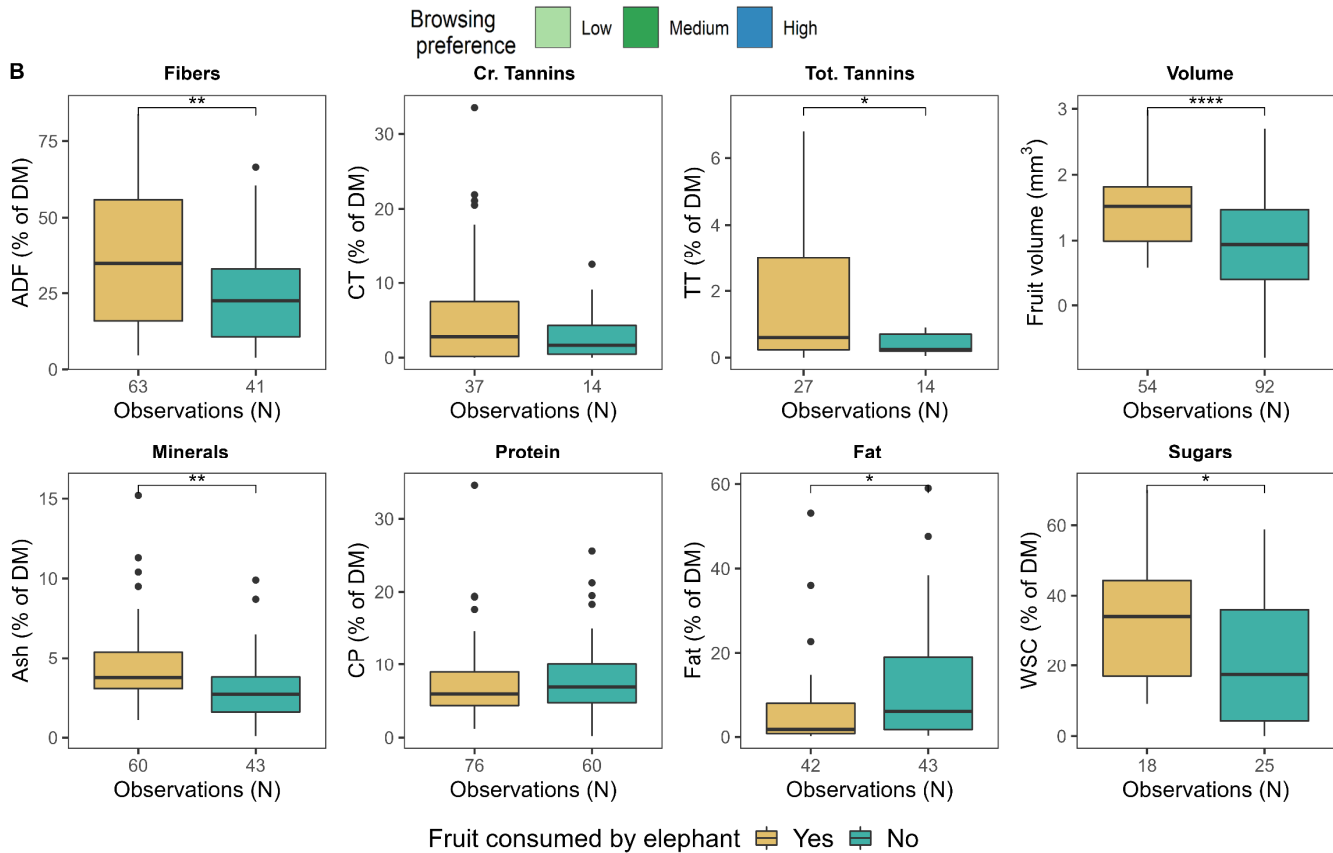
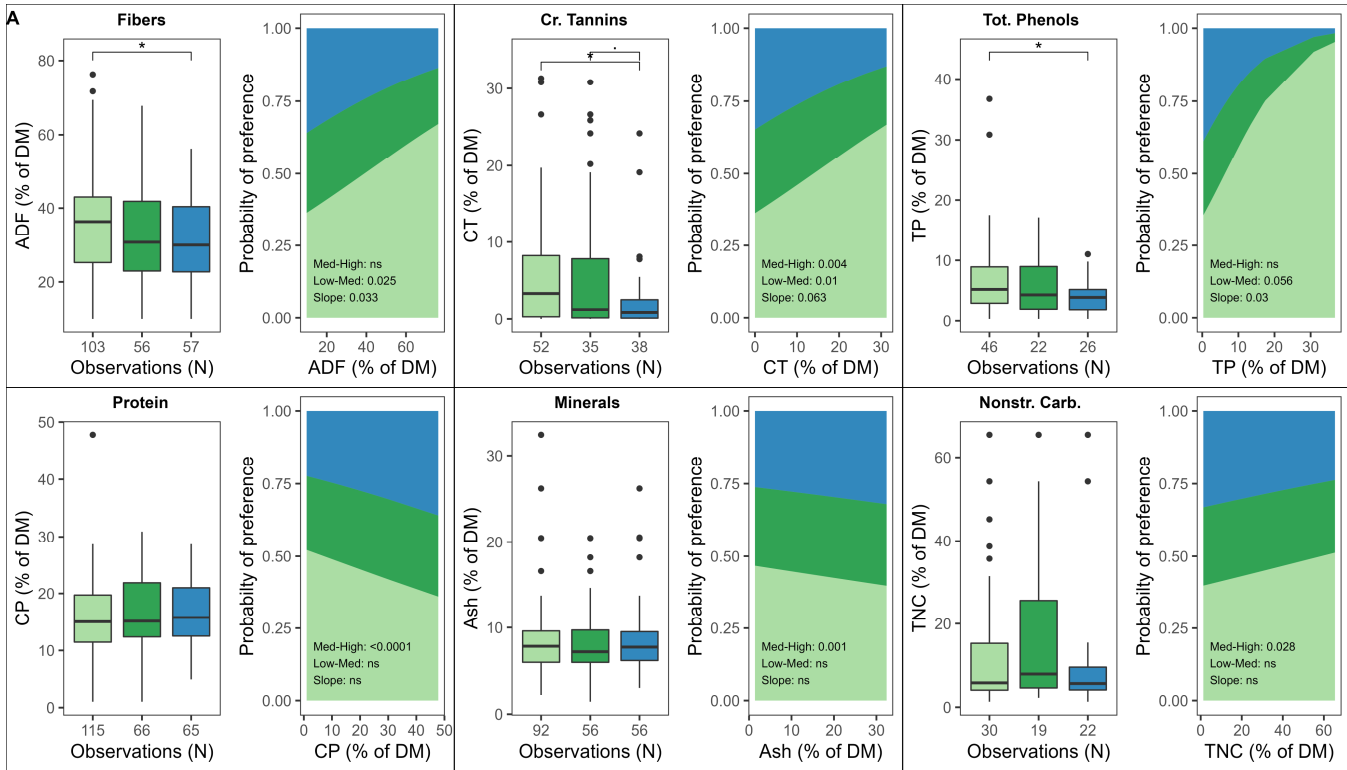
657

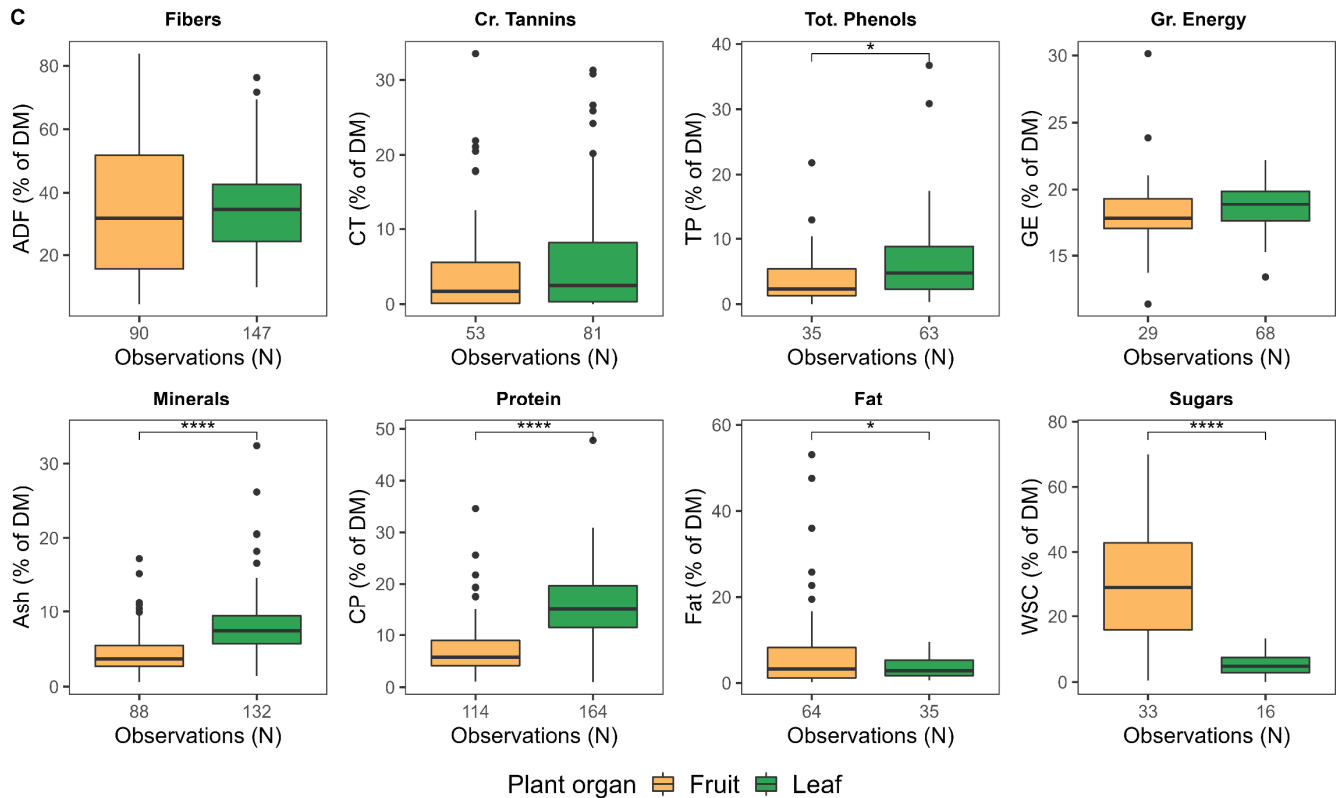
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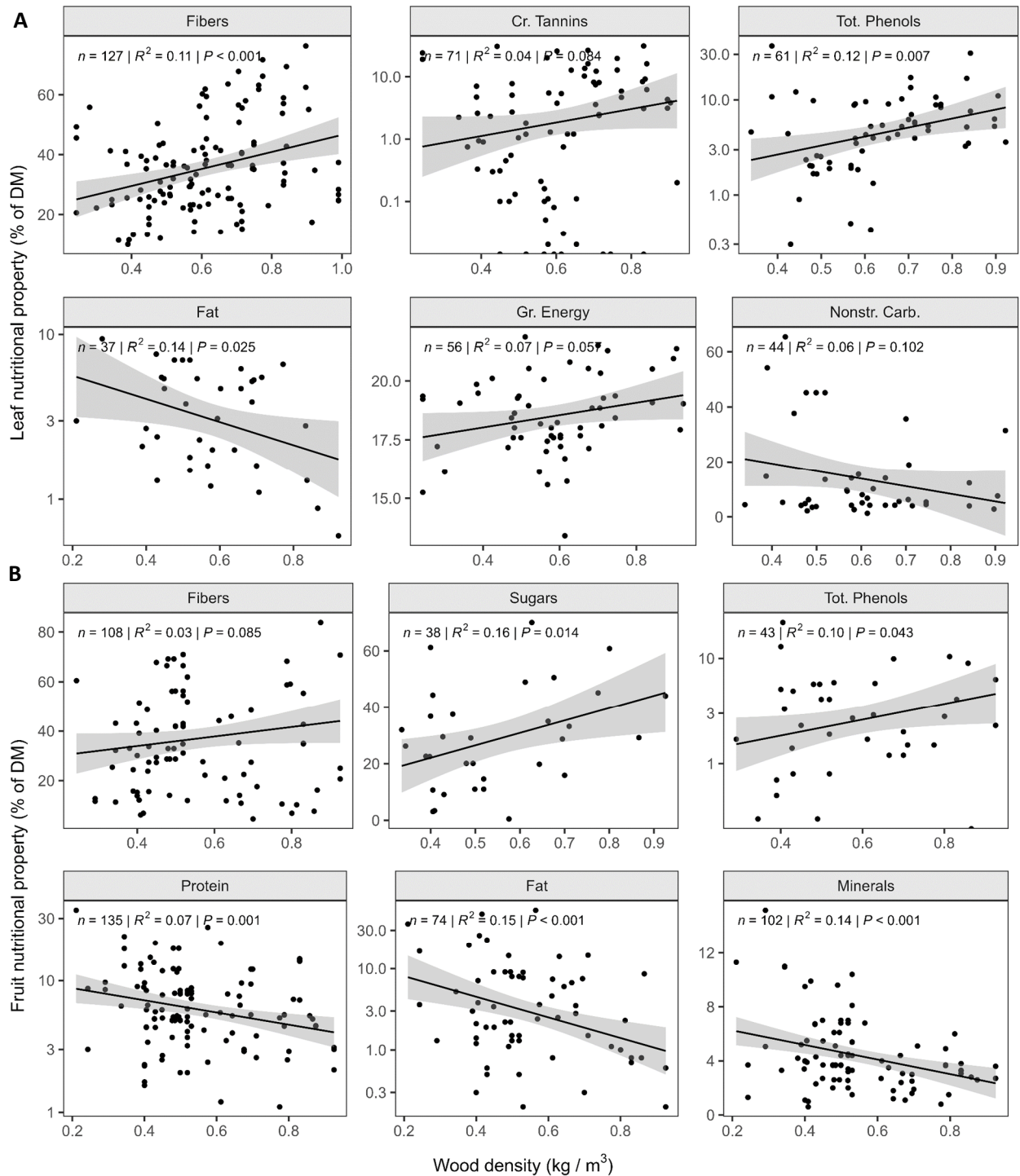
662 **Data availability:** LuiKotale and Ndoki vegetation plot data are available at ForestPlots.net; all
663 other data are available from their respective sources list in Table S2.

664

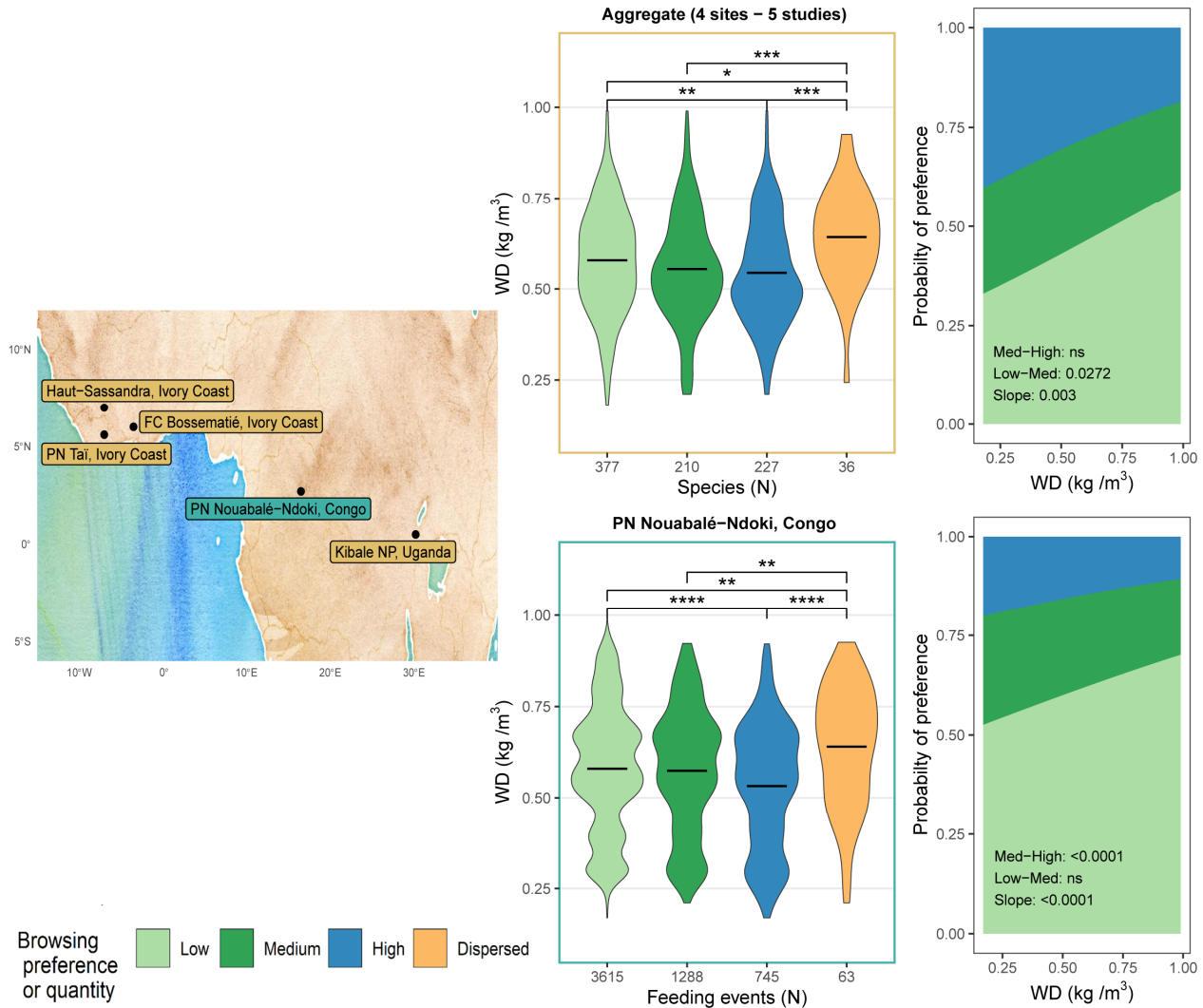




667 **Fig. 1** Nutritional and morphological characteristics of leaves and fruit according to elephant food
668 preferences (low, medium, high) and elephant fruit consumption or avoidance. (A) box plot (left panel)
669 and predicted probability of preference based on results from ordinal logistic regression (right panel)
670 indicating the probability of a plant falling within a preference category with increasing abundance of each
671 nutritional property. For fibers, for example, the figure shows that as fibers increase there is a higher
672 probability of a species being in the low preference category. The P-values shown in the right panels
673 indicate whether the slope differs significantly from zero and is calculated based on the ordinal logistic
674 regression results (Material and Methods). Acid Detergent Fiber (ADF); Crude Tannins (CT); Total
675 Phenols (TP); Total Non-structural Carbohydrates (sugars + starches, TNC); Crude Protein (CP); Ash
676 (minerals); Total Tannins (TT); Water Soluble Carbohydrates (sugars, WSC); Gross Energy (GE). (B)
677 Comparison of fruit consumed or not consumed by elephants at Ndoki; these include Obligate and Non-
678 Obligate trees (see text). (C) Comparison of fruit and leaves consumed by elephants. In box-plots, P-
679 values calculated using t-test indicate statistical significance between the mean of the two groups.
680 Significance levels: *P < 0.10; *P < 0.05; **P < 0.01; ***P < 0.001; ****P < 0.0001.

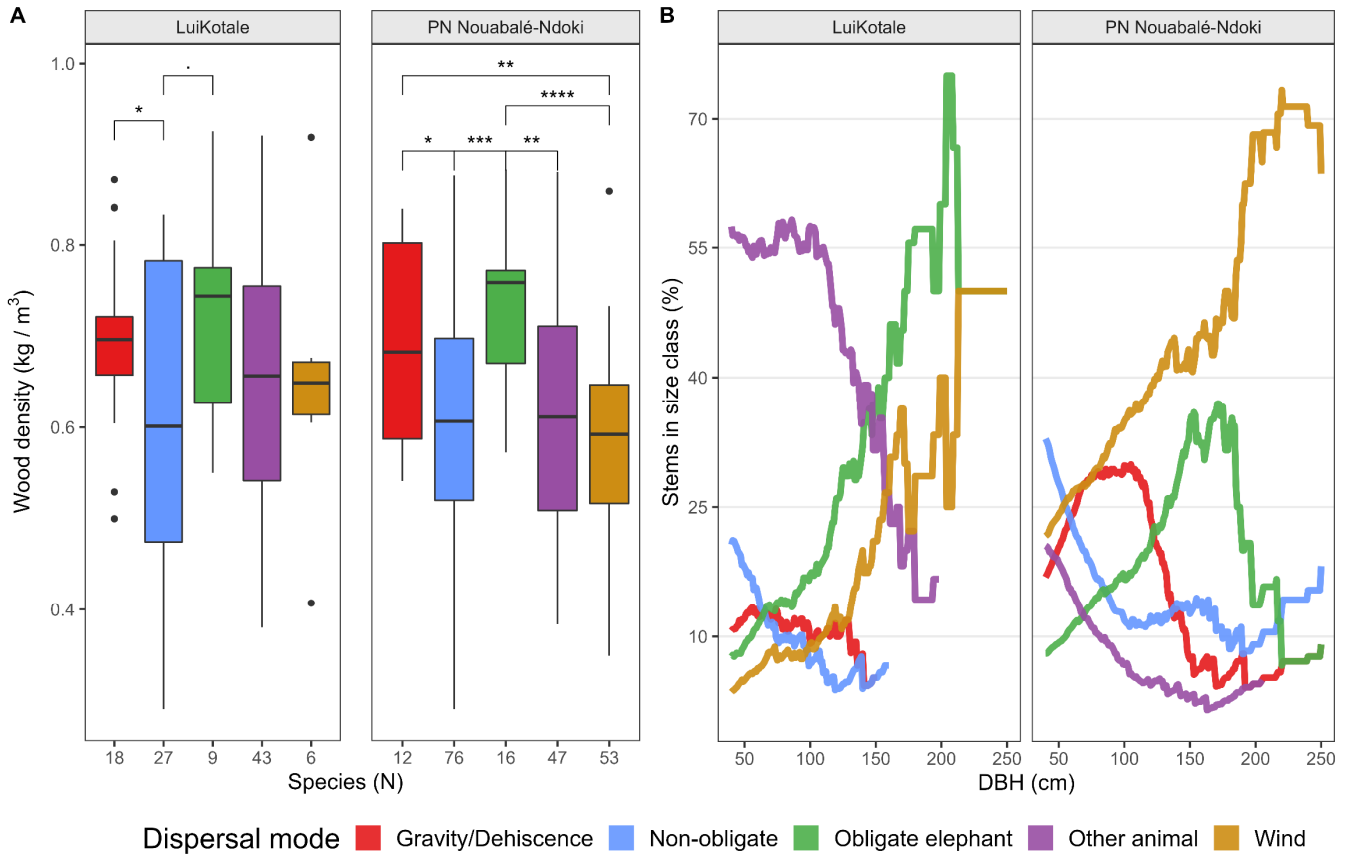


682 Fig. 2. Correlations between nutritional characteristics of leaves and fruits consumed by elephants and
 683 wood density. Each dot represents a tree species that is consumed by elephants. Gross Energy is
 684 expressed in MJ/kg. (A) Leaf and (B) fruit nutritional properties as a function of wood density. Crude
 685 tannins, total phenols, and Fat of leaves and fruit and were log transformed (see Materials and Methods).

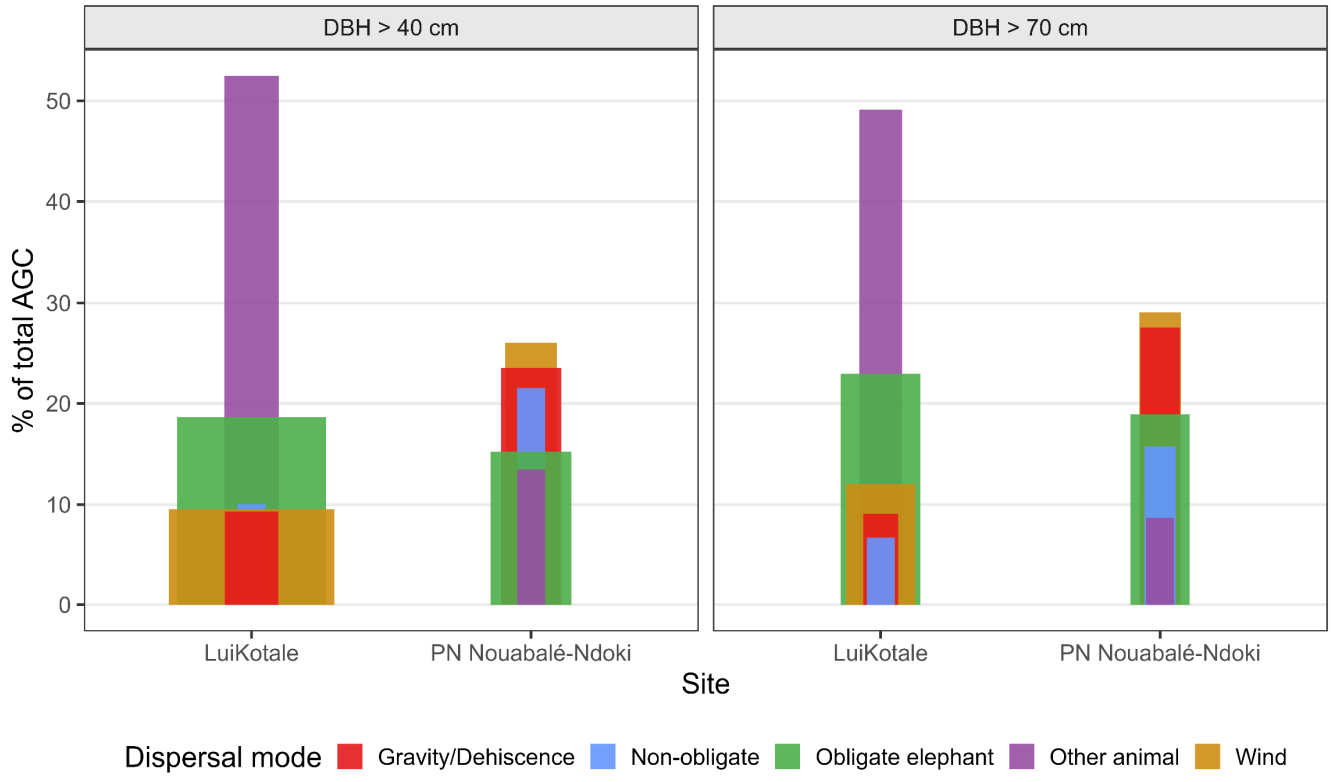


686

687 **Fig. 3.** Wood density in relation to elephant browsing preferences and elephant-dispersed plants in five
 688 sites across tropical Africa. Sites included in the aggregated analysis are highlighted in light yellow on
 689 the map. Variation of wood density across browsing preference groups and dispersed plants (left panels)
 690 and predicted probability of preference based on results from ordinal logistic regression (right panel)
 691 indicating the probability of a plant falling within a preference group with increasing wood density. In violin
 692 plots, P-values calculated using t-test indicate significance differences between the mean of the low and
 693 high groups and leaves vs. fruit groups. P-values in probability plots are based on the ordinal logistic
 694 regression. The elephant-dispersed group includes species dispersed by elephants and includes
 695 Obligate and Non-obligate species (see text). Significance level of pairwise statistical comparison:
 696 *P < 0.05; **P < 0.01; ***P < 0.001; ****P < 0.0001.



698 **Fig 4.** Wood density and forest structure at Ndoki and LuiKotale according to dispersal modes.
 699 (A) Variation in wood density in species with different dispersal mode. Significance level of
 700 pairwise statistical comparison: $\cdot P < 0.10$; $*P < 0.05$; $**P < 0.01$; $***P < 0.001$; $****P < 0.0001$. (B)
 701 Relative percentage of stems for each dispersal guild in relation to total stems at each increase
 702 of 1 cm in diameter at breast height (DBH). Complete list of species with their respective
 703 dispersal mode in Dataset S1.



705 **Fig. 5.** Relative contribution of dispersal guilds to aboveground carbon at different size
 706 thresholds. The bar width is an indication of the relative importance of each guild for AGC in
 707 relation to the total number of stems in the forest. It is calculated for each guild by dividing the
 708 percentage of total AGC by the percentage of stems at each site. Larger ratios (wider bars)
 709 indicate a large contribution to AGC relative to a small number of stems.

710

Description	Quantitative result	Qualitative result (if any)	Location, elephant density, and sampled area	Ref.
<i>Mortality - regeneration</i>				
Mortality rate after elephant damage (DBH > 10 cm)	1.4% (Annual rate)		Kibale NP, Uganda, 5.3 ha	31
Recovery rate after elephant damage (DBH > 10 cm)	1.2% (Annual rate)		Kibale NP, Uganda, 5.3 ha	31
Sapling mortality rate	4% (Annual rate)		Kibale NP, Uganda, logged	12
Seedling and saplings mortality (height > 10 cm)	15-18%		Kibale NP, Uganda, logged	13
Tree toppling & branch breaking	2 - 9.9 cm DBH: - topped 40.9% - broken branch 24% > 10 cm DBH - topped 6.9% - broken branch 7%	Tree toppling and broken branches decline sharply for trees > 10 cm DBH. Larger trees suffer more bark stripping	Bwindi NP, Uganda, 0.97 ha	45
	68% breaks by elephants	Most breaks between 1 m and 3 m height, 2 cm and 6 cm DBH	Several sites, Gabon	11
Reduced regeneration	- Browsed species contained 19% saplings of canopy and 48% subcanopy species - Trampling, movement, and grubbing prevents regeneration in 25% of the sampled area		Shimba Hills National Reserve, Kenya (both forest and savanna elephants common in the part)	46
	- Canopy opening < 20% and forest gaps < 300 m ² reduces elephant		Kibale NP, Uganda	47
<i>Forest properties</i>				
Mean DBH from trail (distance from trail)	52 cm (0-5m) 23 cm (21-25m)	Mean DBH decreases away from trails	Salonga, 0.05 ind/km ² , 100 km of transects	28
Understory openness & elephant encounter rate	$y = 0.2386x + 0.055$	Dung encounter rate increases linearly with understory openness	Salonga, 0.05 ind/km ² , 100 km of transects	28
Tree species composition & distance from trail		Distribution of fruit-preferred and browse-preferred trees varies as a function of distance from trails	Salonga, 0.05 ind/km ² , 100 km of transects	28
Seedling and sapling density and damage near elephant trees	Elephant presence increases chances of damage to seedlings (84%) and saplings (24%)		Ivindo NP, Gabon	48
Aboveground carbon	$y = -0.0841 + 0.3311x - 0.0630x^2$	Percentage change in aboveground carbon (y) as a function of elephant density (x)	Process-based vegetation model	5
Stem density	Reduced density of plants between < 1 cm and \geq 1 m in height			11
Stem scarring (DBH > 10 cm)	16% of stems scarred		Rabongo, Uganda, 7ha	49
Debarking height	Species-specific results	Percentage of debarked trees, average diameter and debarking height		29,39

711 **Table 1. Summary of literature review of the ecological effects of elephants in closed**
712 **canopy forests across the Afrotropics.** Only studies that provided a quantitative measure or
713 a mathematical function were included in the table. DBH = diameter at breast height

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