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10 **Standardised genetic diversity-life history correlates for improved**  
11 **genetic resource management of Neotropical trees**

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13 Running title: Standardised tree life history-population genetic correlates

14  
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57 **(A) ABSTRACT**

58 **(B) Aim.** Life history traits and range size are key correlates of genetic diversity in trees. We  
59 used a standardized sampling protocol to explore how life history traits and range size relate to  
60 the magnitude, variance and structuring (both between and within population) of genetic diversity  
61 in Neotropical tree species.

62 **(B) Location.** The Neotropics

63 **(B) Methods.** We present a meta-analysis of new population genetic data generated for 23  
64 Neotropical tree species (= 2966 trees, 86 populations) across a shared and broad geographic  
65 area. We compared established population genetic metrics across these species (e.g. genetic  
66 diversity, population structure, fine-scale genetic structure), plus we estimated the rarely used  
67 variance in genetic diversity among populations. We used a multivariate, maximum likelihood,

68 multi-model inference approach to explore the relative influence of life history traits and range  
69 size on patterns of neutral genetic diversity.

70 **(B) Results.** We found that pioneer and narrow range species had lower levels but greater

71 variance in genetic diversity – signs of founder effects and stronger genetic drift. Animal

72 dispersed species had lower population differentiation, indicating extensive gene flow.

73 Abiotically dispersed and pioneer species had stronger fine-scale genetic structure, suggesting

74 restricted seed dispersal and family cohort establishment.

75 **(B) Main conclusions.** Our multi-variable and multi-species approach allows ecologically

76 relevant conclusions, since knowing whether one parameter has an effect, or one species shows a

77 response in isolation, is dependent on the combination of traits expressed by a species. Our study

78 demonstrates the influence of ecological processes on the distribution of genetic variation in

79 tropical trees, and will help guide genetic resource management, and contribute to predicting the

80 impacts of land-use change.

81

82 **Keywords:** effective population size, founder effects, gene flow, genetic resource management,

83 seed dispersal

84

## 85 **(A) INTRODUCTION**

86 The life history traits and range size of tree species play critical roles in defining the magnitude

87 and spatial arrangement of their genetic diversity (Duminil *et al.*, 2007; Meirmans *et al.*, 2011;

88 Breed *et al.*, 2015; Broadhurst *et al.*, 2017). Consequently, traits and geographic ranges have

89 become key considerations for planning genetic resource management (Montoya *et al.*, 2008;

90 Breed *et al.*, 2013), the next generation of species distribution models (Swab *et al.*, 2012;

91 Fordham *et al.*, 2014), and for underpinning studies of ecosystem function, conservation and

92 restoration strategies (FAO, 2014; IPBES, 2014; Suding *et al.*, 2015).

93 For over 30 years, researchers have debated the relative influence of a range of life history

94 traits and geographic patterns on population genetic variation in tree species (Loveless &

95 Hamrick, 1984; Hamrick *et al.*, 1992; Hamrick *et al.*, 1993; Hamrick & Godt, 1996; Nybom &

96 Bartish, 2000; Degen *et al.*, 2001; Hardy *et al.*, 2006; Duminil *et al.*, 2007; Montoya *et al.*, 2008;

97 Meirmans *et al.*, 2011; Harata *et al.*, 2012; Broadhurst *et al.*, 2017). Previous meta-analyses have

98 shown that range size, growth form and mating system can be important predictors of the

99 magnitude of genetic diversity, and that growth form, seed dispersal vector and mating system

100 are associated with species-wide genetic structure. While these previous meta-analyses have

101 advanced our understanding of patterns of population genetic variation, most have explored

102 single life history traits or geographic patterns in isolation (but see Hamrick & Godt, 1990;

103 Hamrick & Godt, 1996; Broadhurst *et al.*, 2017). Multivariate approaches are superior to single  
104 variable approaches when attempting to rank the importance of several competing predictor  
105 variables. Additional work is warranted to explore predictors of population genetic structure  
106 within populations, and whether patterns of population genetic variation within populations scale  
107 up to species-level patterns.

108 In this study, we present a meta-analysis of new data generated by a collaboration of  
109 researchers from ten institutions. Our study used standardized sampling of 23 tree species across  
110 a shared and broad geographic area – the Neotropics – to explore how key life history traits (seed  
111 dispersal vector and successional stage) and range size associated with the magnitude and  
112 structure of genetic diversity. We also estimated the standard deviation ( $\sigma$ ) and coefficient of  
113 variation ( $CV = \sigma/\bar{x}$ ) of genetic diversity among populations, which have rarely been used to  
114 compare differences among species since they were first proposed by Brown and Weir (1983)  
115 and further developed by Schoen and Brown (1991). We expect that variation in genetic diversity  
116 among populations will be higher in species that have traits that increase the risk of episodic but  
117 dramatic losses in genetic diversity, such as pioneer species that undergo strong founder effects  
118 (Davies *et al.*, 2010).

119 We used a multi-variable statistical approach that explores the relative influence of life  
120 history traits and range size on patterns of neutral genetic diversity, while accounting for potential  
121 correlations among characters. Our multi-variable and multi-species approach allows more  
122 ecologically relevant conclusions, since knowing whether one parameter has an effect, or one  
123 species shows a response in isolation, is dependent on the combination of traits expressed by a  
124 species. We investigated the following questions: (1) how do life history traits and range size  
125 relate to the magnitude, variance and structuring (both between and within population) of genetic  
126 diversity in 23 Neotropical tree species? (2) are these patterns consistent with findings from  
127 previous meta-analyses? Finally, we interpret our results in terms of relevance to the management  
128 of Neotropical tree genetic resources.

129

## 130 (A) METHODS

### 131 (B) Study species

132 Our 23 study species are all trees that largely occur in tropical and sub-tropical forest, with some  
133 extending into seasonally dry forests, are taxonomically resolved, and either dioecious or mixed  
134 to strongly outcrossing Neotropical trees (between 60-100% outcrossing Ward *et al.*, 2005),  
135 which limited variation in mating system and plant habit. Mating system and life form are  
136 characters that have been identified as confounding variables in previous studies, as both have  
137 been shown to have strong effects on patterns of neutral genetic diversity (Hamrick & Godt,

138 1996; Duminil *et al.*, 2007). To further minimize confounding effects, we used a consistent  
139 approach to study each species (see Fig. S1 in Supporting Information). Where possible, we  
140 standardized population sampling (mean  $\pm$  SD populations per species =  $3.7 \pm 1.7$ , range = 2 to  
141 9), focusing our efforts on populations of individually mapped trees (one population per species;  
142 mean  $\pm$  SD n =  $67 \pm 18$ , range = 32 to 89), together with one or more populations close to (50-  
143 100 km) and distant from (>500 km) the mapped population, and focusing on a single geographic  
144 area (i.e. the Neotropics) which incorporated a significant proportion of the species' range in each  
145 case (Fig. 1; Table 1). We used standardized laboratory protocols and genetic markers (AFLPs  
146 Vos *et al.*, 1995) (details of laboratory protocols in Methods S1) to achieve consistency and  
147 comparability of the estimates of population genetic parameters (Vekemans & Hardy, 2004;  
148 Cavers *et al.*, 2005; Kremer *et al.*, 2005; Petit *et al.*, 2005; Hardy *et al.*, 2006; Jump & Peñuelas,  
149 2007; Dick *et al.*, 2008).

150 Species were stratified by three variables central to standing hypotheses, based on data  
151 available at the time of our analysis (Loveless & Hamrick, 1984; Hamrick *et al.*, 1992; Hamrick  
152 *et al.*, 1993; Hamrick & Godt, 1996; Duminil *et al.*, 2007): range size, seed dispersal vector and  
153 successional stage (Table 2). Pollination syndrome has been an important factor to consider in  
154 studying genetic diversity, however we had insufficient variation in this parameter to include it in  
155 our study (18 of 23 were insect pollinated). These categories were used as predictor variables of  
156 patterns of variation in population genetic parameters. The 23 study species were from 22  
157 different genera and 15 families, indicating that our species do not share patterns of population  
158 genetic variation due to recent ancestry, as might conceivably be the case for recently diverged  
159 sister species. For all study species, the magnitude and spatial distribution of genetic variation is  
160 independently acquired.

161 Species were defined as having wide (>50,000 km<sup>2</sup>; n = 15) or narrow (<50,000 km<sup>2</sup>; n =  
162 8) ranges (local endemics, sensu Gentry, 1986). In theory, range size should have a positive effect  
163 on genetic diversity because larger ranges should correlate with larger effective population sizes  
164 (assuming effective density is constant) and reduce the influence of random genetic drift  
165 (Loveless & Hamrick, 1984). This hypothesis has been generally supported by empirical data  
166 (Hamrick *et al.*, 1992; Hamrick & Godt, 1996; Broadhurst *et al.*, 2017). Range size has also been  
167 hypothesized to have a negative effect on population differentiation because larger range size  
168 should correlate with greater dispersal ability and hence greater levels of gene flow (Loveless &  
169 Hamrick, 1984; Hamrick *et al.*, 1992). However, several studies found conflicting patterns in  
170 empirical data (Loveless & Hamrick, 1984; Hamrick *et al.*, 1992; Hamrick & Godt, 1996;  
171 Duminil *et al.*, 2007), a pattern that may be explained by sampling over geographic barriers

172 within wider ranging species, or a greater age of some widespread species (Dick & Heuertz,  
173 2008; Dick *et al.*, 2013), allowing time for genetic differentiation to accrue.

174 Species were grouped as either late successional (n = 11) or pioneer (n = 12) based on  
175 functional trait data (traits included wood density, seed size and specific leaf area; see Table S1),  
176 plus field observations reported in primary literature (Forget, 1992; Huc *et al.*, 1994; Jones *et al.*,  
177 2005; Flores *et al.*, 2006; Silva & Pinheiro, 2009). Pioneer species have been hypothesized to  
178 have lower genetic diversity (Loveless & Hamrick, 1984) and stronger spatial genetic structure  
179 (Davies *et al.*, 2010; Harata *et al.*, 2012), reflecting the habit of copious reproductive output and  
180 recruitment following disturbance, with few overlapping generations, which results in elevated  
181 genetic drift and founding of family groups plus a narrower window of opportunity for incoming  
182 gene flow (for exception, see Born *et al.*, 2008). Expectations of successional stage effects on  
183 population differentiation are mixed (Loveless & Hamrick, 1984), but generally, pioneer species  
184 are expected to exhibit higher levels of population differentiation because founder effects and few  
185 overlapping generations increase genetic drift, leading to rapid divergence among populations,  
186 and reduce opportunities for incoming gene flow.

187 We classified species according to their primary seed dispersal vector and sampled 13  
188 animal-dispersed (*e.g.* bird, bat, monkey, rodent) and 10 abiotically dispersed species (*e.g.*  
189 gravity, explosive capsules, water, wind). Two species are known to undergo both abiotic and  
190 biotic seed dispersal (*Araucaria angustifolia*, *Calophyllum brasiliense*) but were grouped into the  
191 abiotically dispersed group in our analysis. Species with abiotically dispersed seeds are generally  
192 expected to have more limited seed dispersal than species with animal dispersed seeds (Howe &  
193 Smallwood, 1982), hence the former have been found to exhibit stronger population  
194 differentiation (Loveless & Hamrick, 1984; Hamrick *et al.*, 1992; Hamrick & Godt, 1996;  
195 Duminil *et al.*, 2007) and stronger spatial genetic structure (Loveless & Hamrick, 1984; Hamrick  
196 *et al.*, 1993; Harata *et al.*, 2012). The same reasoning suggests that population differentiation  
197 should correlate with spatial genetic structure due to the similar influence of seed dispersal (Dick  
198 *et al.*, 2008), but this remains largely untested.

199

## 200 **(B) Genetic analysis**

201 We performed a genome scan of an average of 228 AFLP loci ( $\pm 30$  SE, range = 61 to 673)  
202 across our uniform sampling design of 23 Neotropical tree species from 96 populations, 2966  
203 trees in total (Table 1; for details of AFLP laboratory methods see Methods S1). We estimated  
204 the percentage of polymorphic loci (P; n = 23 species), mean expected heterozygosity across  
205 populations ( $H_E$ ; n = 23 species), and total expected heterozygosity within species ( $H_T$ ; n = 23  
206 species), and differentiation among populations ( $F_{ST}$ ; n = 21 species) in AFLPsurv (Vekemans,

207 2002). Mean and total expected heterozygosity were tightly correlated ( $r^2 = 0.85$ ), and to  
208 minimize redundancy in our results, our analysis will focus on mean expected heterozygosity.

209 We also calculated the standard deviation of  $P$  and  $H_E$  ( $\sigma P$  and  $\sigma H_E$ ) and the coefficient of  
210 variation of  $P$  and  $H_E$  ( $c_v P$  and  $c_v H_E$ ) among populations, which are underutilized metrics to  
211 explore the variance in diversity across populations (and derived from a parameter first proposed  
212 by Brown and Weir in 1983, and further developed by Schoen and Brown 1991). The variance of  
213 population genetic diversity is rarely estimated in tree species because they usually exhibit very  
214 low differentiation for allelic frequencies and correspondingly low differentiation for diversity  
215 across populations. However, the variance in genetic diversity may be an important metric to  
216 observe in trees because it could, for example, be impacted by the strength of founder effects.  
217 Older, better-connected populations would be expected to have higher diversity than recently  
218 founded populations, as the latter may suffer from genetic bottlenecks (Davies *et al.*, 2010).

219 Spatial genetic structure was analysed in SPAGeDi (Hardy & Vekemans, 2002),  
220 following the procedure described in (Vekemans & Hardy, 2004), and using the Loiselle pairwise  
221 kinship coefficients between individuals,  $F_{ij}$  (Loiselle *et al.*, 1995). To define the slope of the  
222 relationship between average  $F_{ij}$  and geographic distance, we defined distance classes following  
223 the authors' recommendations, where, for each distance class, 50% of all individuals were  
224 represented at least once and the coefficient of variation of the number of times each individual  
225 represented was  $< 1$ . Mean  $F_{ij}$  was plotted over the logarithm of the distance class. Pairwise  
226 kinship coefficients were regressed on the logarithm of pairwise distance to estimate the  
227 regression slope,  $b$ , and the significance of this slope was tested with 10,000 permutations. The  
228 strength of spatial genetic structure was then quantified by calculating  $S_p$  (Vekemans & Hardy,  
229 2004).  $S_p = -b/(F_1 - 1)$ , where  $F_1$  was the average kinship coefficient between individuals within  
230 the first distance class (all species: mean  $\pm$  SE =  $316 \pm 137$  m,  $n = 19$ ; pioneer: mean  $\pm$  SE =  $232$   
231  $\pm 130$  m,  $n = 7$ ; late successional: mean  $\pm$  SE =  $364 \pm 206$  m,  $n = 13$ ) and  $b$  was the regression  
232 slope of  $F_{ij}$  regressed on the logarithm of pairwise distance.  $S_p$  is a reciprocal of neighbourhood  
233 size, where low  $S_p$  indicates that the neighbourhood size is large and therefore weaker spatial  
234 genetic structure is observed.

## 235 236 **(B) Statistics**

237 We used general linear models in a maximum likelihood, multi-model inference framework  
238 (Burnham & Andersen, 2002) in R v. 3.4.1 (2017) to test for hypothesized relationships between  
239 the three life history and geographic predictor variables (range size, seed vector, successional  
240 stage) and the eight genetic response variables ( $P$ ,  $\sigma P$ ,  $c_v P$ ,  $H_E$ ,  $\sigma H_E$ ,  $c_v H_E$ ,  $F_{ST}$ ,  $S_p$ ) at the  
241 species level. We estimated Akaike's Information Criterion corrected for small sample sizes

242 (AICc; calculated in the MuMIn package – [https://cran.r-](https://cran.r-project.org/web/packages/MuMIn/index.html)  
243 [project.org/web/packages/MuMIn/index.html](https://cran.r-project.org/web/packages/MuMIn/index.html)) and Akaike weights ( $w_{AIC}$ ) for each model  
244 (Burnham & Andersen, 2002). To select predictor variables of greatest importance to each  
245 response variable, we derived the index of the relative importance of predictor variable  $i$  ( $AICc_i$ ),  
246 the sum of Akaike weights for all models that included parameter  $i$  (Burnham & Andersen, 2002;  
247 Giam & Olden, 2016). We also calculated ratios of the absolute value of the  $t$  statistic for each  
248 variable to judge variable importance, as suggested by Cade (2015).

249 We used a square root transformation for  $F_{ST}$  and  $CVH_E$ , cube root transformation for  $Sp$ ,  
250 and log base 10 transformation for  $\sigma_P$  and  $CV_P$  to meet the assumption of normality of residuals.  
251 We verified that the models met the statistical assumptions of general linear models by (1) testing  
252 the normality of residuals of fitted models by examining quantile-quantile plots (Crawley, 2007)  
253 and running Shapiro-Wilk tests (Shapiro & Wilk, 1965), and (2) checking for heteroscedasticity  
254 by examining plots of the residuals versus fitted values and scale-location (Crawley, 2007) as  
255 well as running Breusch–Pagan tests in the lmtest library ([https://cran.r-](https://cran.r-project.org/web/packages/lmtest/index.html)  
256 [project.org/web/packages/lmtest/index.html](https://cran.r-project.org/web/packages/lmtest/index.html)) (Breusch & Pagan, 1979). None of the top-ranked  
257 models had  $P > 0.05$  for Shapiro-Wilk or Breusch–Pagan tests, but the multivariate  $F_{ST}$  and  $Sp$   
258 models showed signs of heteroscedasticity in the residuals vs. fitted values plots. For  $P$ , we also  
259 used binomial generalized linear models with polymorphic loci as the successes and non-  
260 polymorphic loci as failures. The response variable for  $P$  was created by taking the sum of the  
261 loci that were polymorphic and not polymorphic for each species across all populations.

262 We ran our main analyses with the species that are known to undergo both abiotic and  
263 biotic seed dispersal (*Araucaria angustifolia* and *Calophyllum brasiliense*) classified as biotic  
264 rather than abiotic seed dispersers. In addition to species-level analysis, we also analysed the  
265 effects of the same predictor variables on population-level  $H_E$  and  $P$  data. For  $P$ , we used  
266 binomial generalized linear mixed-effect models with the lme4 package ([https://cran.r-](https://cran.r-project.org/web/packages/lme4/citation.html)  
267 [project.org/web/packages/lme4/citation.html](https://cran.r-project.org/web/packages/lme4/citation.html)) with species as the random effect. For  $H_E$ , we used  
268 Gaussian mixed-effect models with species as the random effect.

## 270 **(B) Data accessibility**

271 The genetic summary statistics supporting the findings of this study are available within the  
272 Supporting Information. The raw AFLP data will be uploaded to a data repository (e.g. Dryad) if  
273 our paper is accepted for publication.

## 275 **(A) RESULTS**



276 We found genetic diversity differences that correlated with range size (large vs. small range:  
277 mean  $P = 88.66$  vs.  $80.09$ , mean  $H_E = 0.31$  vs.  $0.25$ ;  $AICc_i P = 1.00$ ;  $|t|$  ratio  $P = 0.97$ ;  $AICc_i H_E$   
278  $= 0.67$ ;  $|t|$  ratio  $H_E = 1.00$ ) as well as successional stage (late successional vs. pioneer: mean  $P =$   
279  $90.98$  vs.  $80.82$ , mean  $H_E = 0.30$  vs.  $0.28$ ;  $AICc_i P = 1.00$ ;  $|t|$  ratio  $P = 1.00$ ;  $AICc_i H_E = 0.67$ ;  $|t|$   
280 ratio  $H_E = 0.36$ ), where pioneer and range restricted species had lower genetic diversity (Fig. 2;  
281 Table 3; Table S2, S3). These trends were largely consistent when comparisons were run  
282 individually within our three main study regions (south-east Brazil, Costa Rica, and French  
283 Guyana – inset maps in Fig. 1; Table S4), when binomial generalized linear models were used for  
284  $P$  (Table S5), when mixed-effects models at the population-level were run (for  $P$  but not  $H_E$ ;  
285 Table S6), and when univariate models were run (for both  $P$  and  $H_E$ ; Table S7, S8). The  
286 percentage of polymorphic loci was positively correlated with expected heterozygosity (Fig. S2,  
287 S3; coefficient of determination  $r^2 = 0.51$ ).

288 The standard deviation in the percentage of polymorphic loci ( $\sigma P$ ) and the coefficient of  
289 variation for both percentage of polymorphic loci ( $CV P$ ) and expected heterozygosity ( $CV H_E$ )  
290 were each affected by successional stage (late successional vs. pioneer: mean  $\sigma P = 4.35$  vs.  
291  $10.70$ ;  $AICc_i \sigma P = 0.87$ ;  $|t|$  ratio  $\sigma P = 1.00$ ;  $\sigma H_E$  did not differ; mean  $CV P = 15.30$  vs.  $41.24$ ;  
292  $AICc_i CV P = 0.88$ ;  $|t|$  ratio  $CV P = 1.00$ ; mean  $CV H_E = 0.04$  vs.  $0.01$ ;  $AICc_i CV H_E = 0.98$ ;  $|t|$  ratio  
293  $CV H_E = 1.00$ ), and pioneer species generally exhibited greater variation of genetic diversity  
294 across populations within species than late successional species (Fig. 2; Table 3; Table S2, S3).  
295 These trends were consistent when we ran univariate models (Table S7). Variation in the  
296 percentage of polymorphic loci was correlated with the variance in expected heterozygosity  
297 (coefficient of determination  $r^2 = 0.58$ ), but neither standard deviation metric was correlated with  
298 the corresponding mean estimate ( $\sigma P \sim P$ : coefficient of determination  $r^2 = 0.07$ ;  $\sigma H_E \sim H_E$ :  
299 coefficient of determination  $r^2 = 0.07$ ) or population differentiation ( $\sigma P \sim F_{ST}$ : coefficient of  
300 determination  $r^2 = 0.03$ ;  $\sigma H_E \sim F_{ST}$ : coefficient of determination  $r^2 < 0.01$ ).

301 Population differentiation was associated with range size (large vs. small range: mean  $F_{ST}$   
302  $= 0.126$  vs.  $0.049$ ;  $AICc_i F_{ST} = 0.86$ ;  $|t|$  ratio  $F_{ST} = 1.00$ ) and seed dispersal vector (animal vs.  
303 abiotic dispersal: mean  $F_{ST} = 0.072$  vs.  $0.131$ ;  $AICc_i F_{ST} = 0.65$ ;  $|t|$  ratio  $F_{ST} = 0.83$ ), and animal  
304 dispersed and narrow range species had lower population differentiation (Fig. 2; Table 3; Table  
305 S2, S3). When we ran univariate models, range size remained as a strong predictor whereas seed  
306 dispersal vector was not (Table S7). Population differentiation did not correlate with mean  
307 geographic distance between populations (coefficient of determination  $r^2 = 0.04$ ).

308 We observed marked differences in fine-scale spatial genetic structure associated with  
309 seed dispersal vector (animal vs. abiotic dispersal: mean  $S_p = 0.011$  vs.  $0.028$ ;  $AICc_i S_p = 0.71$ ;  
310  $|t|$  ratio  $S_p = 1.00$ ) as well as successional stage (late successional vs. pioneer: mean  $S_p = 0.010$

311 vs. 0.030; AICc<sub>i</sub> Sp = 0.62; |t| ratio Sp = 0.75), where abiotically dispersed and pioneer species  
312 had stronger fine-scale spatial genetic structure than biotically dispersed and late successional  
313 species (Fig. 2; Table 3; Table S2, S3). These trends were largely consistent when univariate  
314 models were run (Table S7). We also observed that population differentiation and spatial genetic  
315 structure were positively correlated, potentially driven by two species (*Pinus oocarpa* and  
316 *Vochysia ferruginea*), although our results were robust to bootstrapping (Fig. S3, S4; coefficient  
317 of determination  $r^2 = 0.40$ ,  $\beta = 0.133$ ;  $n = 17$ ; 2.5 and 97.5 percentiles of slope distribution of  
318 10,000 bootstrap iterations = 0.003 and 0.232).

319 Our results were generally robust, but were less clear, when the two species that are  
320 known to undergo both abiotic and biotic seed dispersal were switched from abiotic to biotic seed  
321 dispersal classification (*Araucaria angustifolia*, *Calophyllum brasiliense*) (Table S9, S10).

322

### 323 (A) DISCUSSION

324 We show that with consistent sampling and analysis, range size, successional stage and seed  
325 dispersal vector are useful predictors of the magnitude, variance and structuring of genetic  
326 diversity. Our standardized approach included using the same genetic marker type, focusing our  
327 sampling to the same geographic region – the Neotropics – and sampling across a significant  
328 proportion of the species' range, which are factors that have not been controlled in previous  
329 studies (Duminil *et al.*, 2007). Our results should be interpreted with some caution as our study  
330 region does cross known biogeographic areas (Cavers & Dick, 2013), but our results appear  
331 robust to this sampling design. Further, since we analysed all characters together in a multi-  
332 variable, maximum likelihood, multi-model inference framework, which allowed more robust,  
333 ecologically relevant conclusions to be made by decoupling potential correlations among  
334 characters. We used a rarely used population genetic metric – the population genetic diversity  
335 standard deviation ( $\sigma_P$ ,  $\sigma_{H_E}$ ) – that proved sensitive to the successional stage of our study  
336 species. Together, our study provides the first consistently designed, multi-species study to  
337 explore whether species characteristics can predict the magnitude and structuring of genetic  
338 diversity.

339 Among our 23 study species, pioneer species had lower genetic diversity than late  
340 successional species. These findings support the hypothesis that pioneer species colonize gaps in  
341 sibling cohorts, leading to bottlenecks and the loss of genetic diversity (Nybom & Bartish, 2000;  
342 Davies *et al.*, 2010; Harata *et al.*, 2012). These findings indicate that pioneer species either risk  
343 losing adaptive variation during colonization due to genetic drift, which could impact their  
344 adaptive potential, or that these species are intrinsically well equipped to cope with reduced  
345 genetic diversity. Our findings are consistent with the review by Nybom and Bartish (2000), but

346 several other reviews did not observe an effect of successional stage on genetic diversity,  
347 potentially due to the limitations or level of variance of previous studies (Loveless & Hamrick,  
348 1984; Hamrick *et al.*, 1992; Meirmans *et al.*, 2011).

349 Pioneer species also had higher variation in genetic diversity (for  $\sigma P$ , but not  $\sigma H_E$ ). There  
350 has been little discussion in the literature on the drivers of variation in genetic diversity, but our  
351 findings provide justification for further investigation of this parameter, and indicate that  
352 succession and founder effects during gap-colonization are potentially important characters  
353 influencing this variable. This was most likely due to stronger population sampling effects during  
354 gap-colonization and scaling-up of genetic turnover from within-population to inter-population  
355 levels (Dick *et al.*, 2008), as supported by the positive association we observed between  $F_{ST}$  and  
356  $S_p$ . It is perhaps expected that  $F_{ST}$  and  $S_p$  associate as both are measurements of isolation by  
357 distance processes, and as such, both are likely to be impacted by the same factors (e.g. limited  
358 seed dispersal). However, the strength of our conclusions is limited by the variable number of  
359 populations per species, which could adversely affect variance estimates, and we were unable to  
360 test alternative factors that could potentially influence variation in genetic diversity (e.g.  
361 historical demography, asymmetrical gene flow). As such, we suggest that simulation studies  
362 should be undertaken to develop testable hypotheses to better understand the causes and  
363 consequences of variation in genetic diversity, and the associations between fine-scale and  
364 population genetic structure.

365 We observed that range restricted species had lower genetic diversity than wide range  
366 species, which is consistent with the theory that large range sizes buffer genetic diversity  
367 (Loveless & Hamrick, 1984). Species with larger range sizes should also, at least in part, have  
368 greater dispersal capacity or maintain larger effective population sizes, and both would result in  
369 reduced effects of random genetic drift on genetic diversity. Our findings were consistent with  
370 some previous reviews (Hamrick *et al.*, 1992; Hamrick & Godt, 1996; Broadhurst *et al.*, 2017),  
371 but not others (Nybom & Bartish, 2000). As previously reported, we also found redundancy in  
372 the different measures of genetic diversity (Hamrick & Godt, 1990; Meirmans *et al.*, 2011;  
373 Broadhurst *et al.*, 2017), where the percentage of polymorphic loci was highly correlated with  
374  $H_E$ .

375 Population genetic differentiation was strongly associated with seed dispersal vector,  
376 supporting previous theoretical expectations that animals have the capacity to disperse seeds  
377 further, on average, than abiotic means (e.g. wind, water; Loveless & Hamrick, 1984; Hamrick *et*  
378 *al.*, 1992; Hamrick & Godt, 1996; Duminil *et al.*, 2007) (for exceptions, see Nybom & Bartish,  
379 2000; Meirmans *et al.*, 2011). Furthermore, population genetic differentiation was strongly  
380 associated with species range size. Species with wider ranges had stronger population genetic

381 differentiation than species with smaller ranges, which is contrary to the expectation that species  
382 with larger ranges have greater capacity to disperse and thus have lower population genetic  
383 differentiation (Loveless & Hamrick, 1984; Duminil *et al.*, 2007). We suggest that this result  
384 reflects our species-wide sampling efforts, where, despite the absence of an  $F_{ST}$ -geographic  
385 distance correlation, species with wider ranges are likely to also span biogeographic barriers (e.g.  
386 mountains, rivers), increasing isolation by distance. Future studies should explore this result in  
387 more detail by, for example, conducting multi-species studies within areas that do not contain  
388 major dispersal barriers and sampling many populations per species.

389 The strength of spatial genetic structure within populations appeared to be most  
390 influenced by seed dispersal vector and successional stage. Abiotically dispersed plants and  
391 pioneer species had stronger fine-scale spatial genetic structure than biotically dispersed and late  
392 successional species, most likely due to restricted seed dispersal and family cohorts establishing  
393 together. These findings are largely consistent with previous findings (Loveless & Hamrick,  
394 1984; Hamrick *et al.*, 1993; Davies *et al.*, 2010; Harata *et al.*, 2012), and support the use of these  
395 categorical traits to predict levels of gene flow at local scales (Dick *et al.*, 2008).

396

## 397 (A) CONCLUSIONS

398 Protecting and managing forest genetic resources is an urgent priority, particularly as the extent  
399 of forest continues to be reduced and fragmented in the face of ongoing land clearance and  
400 climate change. Forest genetic resources provide the raw material underpinning population  
401 genetic health, adaptive potential, restoration and breeding. A recent international initiative by the  
402 FAO developed the Global Plan of Action on forest genetic resources ([http://www.fao.org/3/a-](http://www.fao.org/3/a-i3849e.pdf)  
403 [i3849e.pdf](http://www.fao.org/3/a-i3849e.pdf)) designed to promote their protection and sustainable management, and regional  
404 consortia such as EUFORGEN (<http://www.euforgen.org/>) have made great strides in identifying  
405 and protecting temperate forest genetic resources. Yet a huge task remains, even in well-  
406 resourced regions such as Western Europe, in finding effective proxies for predicting the levels  
407 and distribution of genetic diversity in tree species as manual characterization of all forest genetic  
408 resources is not tractable. The task, and need, is greatest in the high-diversity forests of the  
409 tropics. Currently, proxy prediction is most commonly done using abiotic environmental  
410 predictors and little biotic knowledge is built in to forecasting where genetic diversity lies.

411 Understanding how ecology relates to genetic diversity can provide important predictive  
412 power for the management of tree species. For example, knowing the relationships between key  
413 characteristics and genetic parameters allows prediction of tree species' capacity to overcome  
414 gaps in distribution or to re-connect fragmented populations (Loveless & Hamrick, 1984), which  
415 could be used to inform the spatial arrangement of connecting corridors. Patterns of neutral

416 genetic diversity can also provide a baseline against which studies of adaptive potential and  
417 adaptation can be set, where populations with higher levels of neutral genetic diversity may also  
418 be those with higher levels of adaptive potential (Sgrò *et al.*, 2011; Broadhurst *et al.*, 2017), and  
419 for seed collections, where diversity sampling can be better targeted (e.g. for seed banking, seed-  
420 based restoration; Broadhurst *et al.*, 2016) should be adjusted based on species characteristics.  
421 While it would be preferable to assign species to continuous character states and to incorporate  
422 phenotypic trait variation for analytical purposes, and new evidence may allow this, using the  
423 categorical assignment and neutral genetic data proved a powerful standpoint on which to make  
424 informed genetic resource management decisions.

425 The relationships we established between species characters and the magnitude, variance  
426 and structure of genetic diversity can be directly used to make much-needed genetic resource  
427 management recommendations (FAO, 2014; IPBES, 2014). Our results on the magnitude of  
428 population genetic diversity indicate that pioneer and narrow range species have lower genetic  
429 diversity, suggesting that species with these characters may either be at risk of poor adaptability  
430 due to low genetic diversity or that they are intrinsically well suited to adapt with low genetic  
431 diversity. It may therefore be required to use multiple seed sources when undertaking seed-based  
432 restoration for these pioneer or narrow range species, to augment their genetic diversity (Breed *et al.*  
433 *et al.*, 2013; Breed *et al.*, 2016). We also implement an infrequently used metric that describes the  
434 variance in genetic diversity across populations, and showed that pioneer species had higher  
435 variance than late successional species. Thus, more populations of pioneer species are likely to be  
436 required if representative species-wide sampling is desired (e.g. for seed banking, seed  
437 production areas; Broadhurst *et al.*, 2016).

438 Our findings for population genetic differentiation indicate that it is possible to predict  
439 species responses to biogeographic barriers based on seed dispersal vector, which can be  
440 integrated with other data to delineate seed zones (Breed *et al.*, 2013), or used to optimize  
441 sampling of database collections for tracking timber stocks (Dormontt *et al.*, 2015). Spatial  
442 genetic structure was most affected by successional stage and seed dispersal vector, and this  
443 knowledge can be used to inform seed collection strategies on how to avoid closely related  
444 individuals and to ensure representative sampling of population-level variation (Lowe *et al.*,  
445 2015). Our findings can also help advance species distribution models by allowing the  
446 incorporation of these population genetic functional group classifications into existing simulation  
447 frameworks (Fordham *et al.*, 2014; McCallum *et al.*, 2014), which are now an important basis for  
448 improving predictions of how land-use changes alter biodiversity and ecosystem services for  
449 forest tree species more generally (IPBES, 2014).

450

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456  
457 **(A) BIOSKETCH**

458 The authors have an interest in the genetic management of Neotropical tree species for  
459 conservation and restoration. AJL, AK, BF, CD, RG, ML, RM, CN proposed the funded project;  
460 AJL, SC, AK designed the study; AJL and SC coordinated field and lab work; HC, CD, BF, RG,  
461 ML, RM, CN, FS, HMV-B undertook field work; SC, HC, NC, GG, MG, RG, ML, RM, CMN,  
462 FS, HMV-B generated data; MFB, CD, BF, JBCH did analyses; MFB, AFL wrote the first draft  
463 of the manuscript, all authors contributed substantially to revisions. The authors declare no  
464 conflicts of interest.

465 **(A) REFERENCES**

- 466 Born, C., Kjellberg, F., Chevallier, M.-H., Vignes, H., Dikangadissi, J.-T., Sanguié, J., Wickings,  
467 E.J. & Hossaert-McKey, M. (2008) Colonization processes and the maintenance of  
468 genetic diversity: insights from a pioneer rainforest tree, *Aucoumea klaineana*.  
469 *Proceedings of the Royal Society Biological Sciences Series B*, **275**, 2171-2179.
- 470 Breed, M.F., Mortimer, P.E. & Lowe, A.J. (2016) Restoration: 'Garden of Eden' unrealistic.  
471 *Nature*, **533**, 469.
- 472 Breed, M.F., Stead, M.G., Ottewell, K.M., Gardner, M.G. & Lowe, A.J. (2013) Which  
473 provenance and where? Seed sourcing strategies for revegetation in a changing  
474 environment. *Conservation Genetics*, **14**, 1-10.
- 475 Breed, M.F., Ottewell, K.M., Gardner, M.G., Marklund, M.H.K., Dormontt, E.D. & Lowe, A.J.  
476 (2015) Mating patterns and pollinator mobility are critical traits in forest fragmentation  
477 genetics. *Heredity*, **115**, 108-114.
- 478 Breusch, T.S. & Pagan, A.R. (1979) A simple test for heteroscedasticity and random coefficient  
479 variation. *Econometrica*, 1287-1294.
- 480 Broadhurst, L.M., Jones, T.A., Smith, F.S., North, T. & Guja, L. (2016) Maximizing seed  
481 resources for restoration in an uncertain future. *BioScience*, **66**, 73-79.

- 482 Broadhurst, L.M., Breed, M.F., Lowe, A.J., Bragg, J., Catullo, R., Coates, D., Encinas-Viso, F.,  
483 Gellie, N., James, E., Krauss, S. & Byrne, M. (2017) Genetic diversity and structure of the  
484 Australian flora. *Diversity and Distributions*, **23**, 41-52.
- 485 Brown, A. & Weir, B. (1983) Measuring genetic variability in plant populations. *Isozymes in*  
486 *plant genetics and breeding, part A* (ed. by S.D. Tanksley and T.J. Orton), pp. 219-239.  
487 Elsevier Science Publishing, Amsterdam.
- 488 Burnham, K.P. & Andersen, D.R. (2002) *Model selection and multimodel inference*, 2nd edn.  
489 Springer, New York.
- 490 Cade, B.S. (2015) Model averaging and muddled multimodel inferences. *Ecology*, **96**, 2370-  
491 2382.
- 492 Cavers, S. & Dick, C.W. (2013) Phylogeography of Neotropical trees. *Journal of Biogeography*,  
493 **40**, 615-617.
- 494 Cavers, S., Degen, B., Caron, H., Lemes, M.R., Margis, R., Salgueiro, F. & Lowe, A.J. (2005)  
495 Optimal sampling strategy for estimation of spatial genetic structure in tree populations.  
496 *Heredity*, **95**, 281-289.
- 497 Crawley, M. (2007) *The R Book*. John Wiley & Sons, Ltd, Chichester, UK.
- 498 Davies, S., Cavers, S., Finegan, B., Navarro, C. & Lowe, A. (2010) Genetic consequences of  
499 multigenerational and landscape colonisation bottlenecks for a Neotropical forest pioneer  
500 tree, *Vochysia ferruginea*. *Tropical Plant Biology*, **3**, 14-27.
- 501 Degen, B., Caron, H., Bandou, E., Maggia, L., Chevallier, M.H., Leveau, A. & Kremer, A.  
502 (2001) Fine-scale spatial genetic structure of eight tropical tree species as analysed by  
503 RAPDs. *Heredity*, **87**, 497-507.
- 504 Dick, C., Hardy, O., Jones, F. & Petit, R. (2008) Spatial scales of pollen and seed-mediated gene  
505 flow in tropical rain forest trees. *Tropical Plant Biology*, **1**, 20-33.
- 506 Dick, C.W. & Heuertz, M. (2008) The complex biogeographic history of a widespread tropical  
507 tree species. *Evolution*, **62**, 2760-2774.
- 508 Dick, C.W., Lewis, S.L., Maslin, M. & Bermingham, E. (2013) Neogene origins and implied  
509 warmth tolerance of Amazon tree species. *Ecology and evolution*, **3**, 162-169.

- 510 Dormontt, E.E., Boner, M., Braun, B., Breulmann, G., Degen, B., Espinoza, E., Gardner, S.,  
511 Guillery, P., Hermanson, J.C. & Koch, G. (2015) Forensic timber identification: It's time  
512 to integrate disciplines to combat illegal logging. *Biological Conservation*, **191**, 790-798.
- 513 Duminil, J., Fineschi, S., Hampe, A., Jordano, P., Salvini, D., Vendramin, G.G. & Petit, R.J.  
514 (2007) Can population genetic structure be predicted from life-history traits? *American*  
515 *Naturalist*, **169**, 662-672.
- 516 FAO (2014) The state of the world's forest genetic resources. In. Commission on Genetic  
517 Resources for Food and Agriculture Organization of the United Nations, Rome.
- 518 Flores, O., Gourlet-Fleury, S. & Picard, N. (2006) Local disturbance, forest structure and  
519 dispersal effects on sapling distribution of light-demanding and shade-tolerant species in a  
520 French Guianian forest. *Acta Oecologica*, **29**, 141-154.
- 521 Fordham, D.A., Brook, B.W., Moritz, C. & Nogués-Bravo, D. (2014) Better forecasts of range  
522 dynamics using genetic data. *Trends in ecology & evolution*, **29**, 436-443.
- 523 Forget, P.-M. (1992) Regeneration ecology of *Eperua grandiflora* (Caesalpiniaceae), a large-  
524 seeded tree in French Guiana. *Biotropica*, **24**, 146-156.
- 525 Gentry, A.H. (1986) Endemism in tropical versus temperate plant communities. *Conservation*  
526 *biology: The science of scarcity and diversity* (ed. by M. Soule), pp. 153-181. Sinauer,  
527 Sunderland, MA.
- 528 Giam, X. & Olden, J.D. (2016) Quantifying variable importance in a multimodel inference  
529 framework. *Methods in Ecology and Evolution*, **7**, 388-397.
- 530 Hamrick, J. & Godt, M. (1990) Allozyme diversity in plant species. *Plant population genetics,*  
531 *breeding, and genetic resources.* (ed. by A.H.D. Brown, M.T. Clegg, A.L. Kahler and  
532 B.S. Weir), pp. 43-63. Sinauer Associates Inc., Sunderland, MA.
- 533 Hamrick, J.L. & Godt, M.J.W. (1996) Effects of life history traits on genetic diversity in plant  
534 species. *Philosophical Transactions of the Royal Society B-Biological Sciences*, **351**,  
535 1291-1298.
- 536 Hamrick, J.L., Godt, M.J.W. & Sherman-Broyles, S.L. (1992) Factors influencing levels of  
537 genetic diversity in woody plant species. *New Forests*, **6**, 95-124.



- 538 Hamrick, J.L., Murawski, D.A. & Nason, J.D. (1993) The influence of seed dispersal  
539 mechanisms on the genetic structure of tropical tree populations. *Vegetatio*, **107**, 281-297.
- 540 Harata, T., Nanami, S., Yamakura, T., Matsuyama, S., Chong, L., Diway, B.M., Tan, S. & Itoh,  
541 A. (2012) Fine-scale spatial genetic structure of ten dipterocarp tree species in a Bornean  
542 rain forest. *Biotropica*, **44**, 586-594.
- 543 Hardy, O.J. & Vekemans, X. (2002) SPAGeDi: a versatile computer program to analyse spatial  
544 genetic structure at the individual or population levels. *Molecular Ecology Notes*, **2**, 618-  
545 620.
- 546 Hardy, O.J., Maggia, L., Bandou, E., Breyne, P., Caron, H., Chevallier, M.H., Doligez, A.,  
547 Dutech, C., Kremer, A., Latouche-HallÉ, C., Troispoux, V., Veron, V. & Degen, B.  
548 (2006) Fine-scale genetic structure and gene dispersal inferences in 10 Neotropical tree  
549 species. *Molecular Ecology*, **15**, 559-571.
- 550 Howe, H.F. & Smallwood, J. (1982) Ecology of seed dispersal. *Annual Review of Ecology and*  
551 *Systematics*, **13**, 201-228.
- 552 Huc, R., Ferhi, A. & Guehl, J.M. (1994) Pioneer and late stage tropical rainforest tree species  
553 (French Guiana) growing under common conditions differ in leaf gas exchange  
554 regulation, carbon isotope discrimination and leaf water potential. *Oecologia*, **99**, 297-  
555 305.
- 556 IPBES (2014) Plenary of the Intergovernmental Science-Policy Platform on Biodiversity and  
557 Ecosystem Services. *Antalya, Turkey, 9-14 December 2013*,
- 558 Jones, F., Chen, J., Weng, G. & Hubbell, S. (2005) A genetic evaluation of seed dispersal in the  
559 neotropical tree *Jacaranda copaia* (Bignoniaceae). *American Naturalist*, **166**, 543-555.
- 560 Jump, A. & Peñuelas, J. (2007) Extensive spatial genetic structure revealed by AFLP but not SSR  
561 molecular markers in the wind-pollinated tree, *Fagus sylvatica*. *Molecular Ecology*, **16**,  
562 925-936.
- 563 Kremer, A., Caron, H., Cavers, S., Colpaert, N., Gheysen, G., Gribel, R., Lemes, M., Lowe, A.J.,  
564 Margis, R., Navarro, C. & Salgueiro, F. (2005) Monitoring genetic diversity in tropical  
565 trees with multilocus dominant markers. *Heredity*, **95**, 274-280.

- 566 Loiselle, B.A., Sork, V.L., Nason, J. & Graham, C. (1995) Spatial genetic structure of a tropical  
567 understory shrub, *Psychotria officinalis* (Rubiaceae). *American Journal of Botany*, **82**,  
568 1420-1425.
- 569 Loveless, M.D. & Hamrick, J.L. (1984) Ecological determinants of genetic structure in plant  
570 populations. *Annual Review of Ecology and Systematics*, **15**, 65-95.
- 571 Lowe, A.J., Cavers, S., Boshier, D., Breed, M.F. & Hollingsworth, P.M. (2015) The resilience of  
572 forest fragmentation genetics - no longer a paradox - we were just looking in the wrong  
573 place. *Heredity*, **115**, 97-99.
- 574 McCallum, K.P., Guerin, G.R., Breed, M.F. & Lowe, A.J. (2014) Combining population genetics,  
575 species distribution modelling and field assessments to understand a species vulnerability  
576 to climate change. *Austral Ecology*, **39**, 17-28.
- 577 Meirmans, P.G., Goudet, J., IntraBioDiv, C. & Gaggiotti, O.E. (2011) Ecology and life history  
578 affect different aspects of the population structure of 27 high-alpine plants. *Molecular  
579 Ecology*, **20**, 3144-3155.
- 580 Montoya, D., Zavala, M.A., Rodríguez, M.A. & Purves, D.W. (2008) Animal versus wind  
581 dispersal and the robustness of tree species to deforestation. *Science*, **320**, 1502-1504.
- 582 Nybom, H. & Bartish, I.V. (2000) Effects of life history traits and sampling strategies on genetic  
583 diversity estimates obtained with RAPD markers in plants. *Perspectives in Plant Ecology,  
584 Evolution and Systematics*, **3**, 93-114.
- 585 Petit, R.J., Duminil, J., Fineschi, S., Hampe, A., Salvini, D. & Vendramin, G.G. (2005)  
586 Comparative organization of chloroplast, mitochondrial and nuclear diversity in plant  
587 populations. *Molecular Ecology*, **14**, 689-701.
- 588 R Core Team (2017) *R: A language and environment for statistical computing*. R Foundation for  
589 Statistical Computing.
- 590 Schoen, D.J. & Brown, A. (1991) Intraspecific variation in population gene diversity and  
591 effective population size correlates with the mating system in plants. *Proceedings of the  
592 National Academy of Sciences*, **88**, 4494-4497.
- 593 Sgrò, C.M., Lowe, A.J. & Hoffmann, A.A. (2011) Building evolutionary resilience for  
594 conserving biodiversity under climate change. *Evolutionary Applications*, **4**, 326-337.

- 595 Shapiro, S.S. & Wilk, M.B. (1965) An analysis of variance test for normality (complete samples).  
596 *Biometrika*, **52**, 591-611.
- 597 Silva, A.L.G.d. & Pinheiro, M.C.B. (2009) Reproductive success of four species of *Eugenia* L.  
598 (Myrtaceae). *Acta Botanica Brasilica*, **23**, 526-534.
- 599 Suding, K., Higgs, E., Palmer, M., Callicott, J.B., Anderson, C.B., Baker, M., Gutrich, J.J.,  
600 Hondula, K.L., LaFevor, M.C., Larson, B.M.H., Randall, A., Ruhl, J.B. & Schwartz,  
601 K.Z.S. (2015) Committing to ecological restoration. *Science*, **348**, 638-640.
- 602 Swab, R.M., Regan, H.M., Keith, D.A., Regan, T.J. & Ooi, M.K.J. (2012) Niche models tell half  
603 the story: spatial context and life-history traits influence species responses to global  
604 change. *Journal of Biogeography*, **39**, 1266-1277.
- 605 Vekemans, X. (2002) AFLP-SURV version 1.0. Laboratoire de Genetique et Ecologie Vegetale.  
606 *Universite Libre de Bruxelles, Belgium*,
- 607 Vekemans, X. & Hardy, O.J. (2004) New insights from fine-scale spatial genetic structure  
608 analyses in plant populations. *Molecular Ecology*, **13**, 921-935.
- 609 Vos, P., Hogers, R., Bleeker, M., Reijans, M., van de Lee, T., Hornes, M., Frijters, A., Pot, J.,  
610 Peleman, J., Kuiper, M. & Zabeau, M. (1995) AFLP: A new technique for DNA  
611 fingerprinting. *Nucleic Acids Research*, **23**, 4407-4414.
- 612 Ward, M., Dick, C.W., Gribel, R. & Lowe, A.J. (2005) To self, or not to self ... A review of  
613 outcrossing and pollen-mediated gene flow in neotropical trees. *Heredity*, **95**, 246-254.

614  
615

## 616 (A) SUPPORTING INFORMATION

617 Additional Supporting Information may be found in the online version of this article

618

619 **Methods S1** AFLP methods

620 **Table S1** Functional trait data (sourced from TRY) by succession category

621 **Table S2** Genetic diversity, population genetic differentiation and fine-scale spatial genetic  
622 structure data for the study species

623 **Table S3** Population genetic patterns investigated with general linear models

624 **Table S4** Mean population genetic diversity in the three main regions of our study

625 **Table S5** Binomial generalized linear model results for the effects of the species characters on P

626 **Table S6** Population genetic patterns investigated at the population level with generalized mixed  
627 effects models

628 **Table S7** Univariate population genetic patterns investigated with general linear models

629 **Table S8** Univariate binomial generalized linear model results for the effects of species  
630 characters on P

631 **Table S9** Population genetic patterns investigated with general linear models with the two  
632 species that are known to undergo both abiotic and biotic seed dispersal classified as biotic rather  
633 than abiotic

634 **Table S10** Binomial generalized linear model results for the effects of the species characters on P  
635 with the two species that are known to undergo both abiotic and biotic seed dispersal classified as  
636 biotic rather than abiotic

637 **Figure S1** We used a consistent study design, including species selection, population sampling  
638 and the genetic marker used

639 **Figure S2** Plot of percentage of polymorphic loci against mean expected heterozygosity ( $H_E$ )

640 **Figure S3** Plot of first two principal components of a PCA of the genetic response variables,  
641 showing the associations of the five main population genetic parameters

642 **Figure S4** Plot of population differentiation ( $F_{ST}$ ) estimates against fine-scale spatial genetic  
643 structure ( $S_p$ ) for each species

644  
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648 (other than missing files) should be addressed to the authors.

649

650 **Table 1** Family, range size, seed dispersal vector, successional stage, number of AFLP loci scored, number of populations sampled and total number  
 651 of samples across all populations of the study species.

<b>Species</b>	<b>Family</b>	<b>Range size</b>	<b>Seed dispersal vector</b>	<b>Successional stage</b>	<b>Loci</b>	<b><i>n</i> populations (<i>n</i> total samples)</b>
<i>Anacardium occidentale</i>	Anacardiaceae	Wide	Biotic (birds)	Pioneer	181	2 (89)
<i>Araucaria angustifolia</i>	Araucariaceae	Wide	Mixed (gravity, birds)	Shade tolerant	673	9 (190)*
<i>Bocoa prouacensis</i>	Fabaceae	Narrow	Biotic (monkeys, bats)	Shade tolerant	88	2 (123)*
<i>Calophyllum brasiliense</i>	Clusiaceae	Wide	Mixed (gravity, water, bats)	Shade tolerant	519	4 (159)*
<i>Chrysophyllum sanguinolentum</i>	Sapotaceae	Wide	Biotic (monkeys)	Shade tolerant	149	3 (121)*
<i>Dicorynia guianensis</i>	Fabaceae	Narrow	Abiotic (gravity)	Shade tolerant	134	3 (92)*
<i>Eperua falcata</i>	Fabaceae	Narrow	Abiotic (gravity)	Shade tolerant	107	4 (169)*
<i>Eperua grandiflora</i>	Fabaceae	Narrow	Abiotic (gravity)	Shade tolerant	173	3 (113)*
<i>Eugenia uniflora</i>	Myrtaceae	Wide	Biotic (birds)	Pioneer	205	5 (71)*
<i>Hyeronima alchorneoides</i>	Euphorbiaceae	Wide	Biotic (birds)	Shade tolerant	213	5 (244)*
<i>Jacaranda copaia</i>	Bignoniaceae	Wide	Abiotic (wind)	Pioneer	125	3 (92)
<i>Lecythis ampla</i>	Lecythidaceae	Wide	Biotic (rodents)	Shade tolerant	242	6 (157)*
<i>Lonchocarpus costaricensis</i>	Fabaceae	Narrow	Abiotic (wind)	Pioneer	487	6 (114)
<i>Pinus oocarpa</i>	Pinaceae	Wide	Abiotic (wind)	Pioneer	383	3 (132)*
<i>Sideroxylon capiri</i>	Sapotaceae	Narrow	Biotic (monkeys, bats)	Pioneer	254	4 (86)*
<i>Simarouba amara</i>	Simaroubaceae	Wide	Biotic (monkeys, birds)	Pioneer	157	5 (136)*
<i>Swietenia macrophylla</i>	Meliaceae	Wide	Abiotic (wind)	Pioneer	242	2 (106)*
<i>Symphonia globulifera</i>	Clusiaceae	Wide	Biotic (monkeys, bats)	Shade tolerant	184	3 (153)*
<i>Tapirira guianensis</i>	Anacardiaceae	Wide	Biotic (monkeys, birds)	Pioneer	198	4 (173)*
<i>Tetragastris panamensis</i>	Burseraceae	Wide	Biotic (monkeys, birds)	Shade tolerant	208	2 (115)*

<i>Virola michelii</i>	Myristicaceae	Narrow	Biotic (monkeys, birds)	Pioneer	240	2 (55)
<i>Vochysia ferruginea</i>	Vochysiaceae	Wide	Abiotic (wind)	Pioneer	61	4 (183)*
<i>Vouacapoua americana</i>	Fabaceae	Narrow	Biotic (rodents)	Shade tolerant	92	2 (93)*

652 \*The larger population was spatially mapped for fine-scale spatial genetic structure analysis **Table 2** Predicted effects of three species characteristics (range size,  
653 seed dispersal, succession stage) on the levels, variance and structure of population genetic diversity. The process, support for and against these  
654 predictions from the literature are indicated, as are the findings from our study.

Characteristic	Prediction	Process	Support for	Support against	This study
<b>Range size</b>	Species with larger ranges have higher genetic diversity	Weaker genetic drift	(Hamrick & Godt, 1990; Hamrick <i>et al.</i> , 1992; Hamrick & Godt, 1996)	(Nybom & Bartish, 2000)	Species with larger ranges had higher genetic diversity
	No predicted effect on genetic diversity standard deviation				No effect detected
	Species with larger ranges have weaker population genetic differentiation	Greater colonizing ability connects populations	(Hamrick & Godt, 1990; Hamrick <i>et al.</i> , 1992; Hamrick & Godt, 1996)	(Loveless & Hamrick, 1984; Duminil <i>et al.</i> , 2007)	Species with larger ranges had stronger population genetic differentiation
	No predicted effect on spatial genetic structure				No effect detected
<b>Seed dispersal</b>	No predicted effect on genetic diversity				No effect detected
	No predicted effect on genetic diversity standard deviation				No effect detected
	Species with biotically dispersed seeds have weaker population genetic differentiation	Wider seed dispersal	(Loveless & Hamrick, 1984; Hamrick <i>et al.</i> , 1992; Hamrick & Godt, 1996; Duminil <i>et al.</i> , 2007)	(Nybom & Bartish, 2000; Meirmans <i>et al.</i> , 2011)	Species with biotically dispersed seeds had weaker population genetic differentiation
	Species with biotically dispersed seeds	Wider seed dispersal	(Loveless & Hamrick,		Species with biotically

	have weaker spatial genetic structure		1984; Hamrick <i>et al.</i> , 1993; Harata <i>et al.</i> , 2012)		dispersed seeds had weaker spatial genetic structure
<b>Successional stage</b>	Pioneer species have lower genetic diversity	Founder effects leading to genetic bottlenecks	(Nybom & Bartish, 2000; Davies <i>et al.</i> , 2010; Harata <i>et al.</i> , 2012)	(Loveless & Hamrick, 1984; Hamrick <i>et al.</i> , 1992; Meirmans <i>et al.</i> , 2011)	Pioneer species had lower genetic diversity
	Pioneer species have larger genetic diversity standard deviations	Stronger population sampling effects during colonization	(Dick <i>et al.</i> , 2008)		Pioneer species had larger variance in genetic diversity
	Pioneer species have stronger population genetic differentiation	Founder effects increase genetic drift, leading to rapid differentiation			No effect detected
	Pioneer species have stronger spatial genetic structure	Founder effects leading to family group establishment	(Davies <i>et al.</i> , 2010; Harata <i>et al.</i> , 2012)	(Born <i>et al.</i> , 2008)	Pioneer species had stronger spatial genetic structure

655 **Table 3** Population genetic patterns investigated with general linear models. % DE, percentage  
656 deviance explained by the model;  $\Delta\text{AICc}$ , indicator of difference between model Akaike's  
657 Information Criterion corrected for small samples sizes (AICc) and the minimum AICc in the  
658 model set;  $w\text{AICc}$ , weight that show the relative likelihood of model  $j$ ;  $k$ , the number of parameters;  
659 only models with a  $\Delta\text{AICc}$  less than the null model ( $\sim 1$ ) are shown.

Model	% DE	$\Delta\text{AICc}$	$w\text{AICc}$	$k$
Population expected heterozygosity ( $H_E$ )				
$H_E \sim \text{range}$	29.53	0.00	0.39	2
$H_E \sim \text{range} + \text{succession}$	38.02	0.01	0.39	3
$H_E \sim \text{range} + \text{seed}$	29.74	2.89	0.09	3
$H_E \sim \text{range} + \text{seed} + \text{succession}$	38.19	3.25	0.08	4
$H_E \sim 1$	0.00	5.39	0.03	1
Expected heterozygosity variance ( $\sigma_{H_E}$ )				
$\sigma_{H_E} \sim 1$	0.00	0.00	0.32	1
Expected heterozygosity coefficient of variation ( $cv_{H_E}$ )				
$cv_{H_E} \sim \text{succession}$	37.48	0.00	0.63	2
$cv_{H_E} \sim \text{seed} + \text{succession}$	38.61	2.54	0.18	3
$cv_{H_E} \sim \text{range} + \text{succession}$	37.48	2.96	0.14	3
$cv_{H_E} \sim \text{range} + \text{seed} + \text{succession}$	38.63	5.84	0.03	4
$cv_{H_E} \sim 1$	0.00	8.14	0.01	1
Percentage of polymorphic loci variance ( $\sigma_P$ )				
$\sigma_P \sim \text{succession}$	24.56	0.00	0.43	2
$\sigma_P \sim \text{seed} + \text{succession}$	30.81	0.97	0.27	3
$\sigma_P \sim \text{range} + \text{succession}$	25.04	2.81	0.11	3
$\sigma_P \sim 1$	0.00	3.82	0.06	1
Percentage of polymorphic loci coefficient of variation ( $cv_P$ )				
$cv_P \sim \text{succession}$	24.37	0	0.47	2
$cv_P \sim \text{seed} + \text{succession}$	29.79	1.25	0.25	3
$cv_P \sim \text{range} + \text{succession}$	24.45	2.94	0.11	3
$cv_P \sim 1$	0	3.76	0.07	1



Population differentiation ( $F_{ST}$ )				
$F_{ST} \sim \text{range} + \text{seed}$	38.52	0.00	0.48	3
$F_{ST} \sim \text{range}$	23.35	1.54	0.22	2
$F_{ST} \sim \text{range} + \text{seed} + \text{succession}$	39.97	3.00	0.11	4
$F_{ST} \sim 1$	0.00	4.38	0.05	1
Fine-scale spatial genetic structure (Sp)				
Sp $\sim \text{succession} + \text{seed}$	38.30	0.00	0.29	3
Sp $\sim \text{range} + \text{seed} + \text{succession}$	46.62	1.01	0.17	4
Sp $\sim \text{range} + \text{seed}$	34.77	1.06	0.17	3
Sp $\sim \text{succession}$	19.29	1.84	0.11	2
Sp $\sim \text{seed}$	15.97	2.61	0.08	2
Sp $\sim \text{range}$	15.02	2.82	0.07	2
Sp $\sim 1$	0.00	3.07	0.06	1

660 NB: Model results for effects of the species characters on P are in Table S8 since we ran binomial  
661 generalized linear models.

## 662 Figure Legends

663 **Fig. 1 Maps showing the location of sampled populations for all species.** Inset maps show  
664 greater detail of Costa Rica (CR), French Guyana (FG) and southeast Brazil (SEB). Populations of  
665 each species are represented by unique symbols, and the population in which trees are individually  
666 mapped is underlined.

667  
668 **Fig. 2 Partitioning of population genetic metrics for Neotropical trees across life history traits  
669 and geographic distribution.** In plots A-C and D-F, two parameters per plot are shown for each  
670 column: A-C - percentage of polymorphic loci (P, filled squares, on left) and expected  
671 heterozygosity ( $H_E$ , open squares, on right); D-F - standard deviation of polymorphic loci ( $\sigma_P$ ,  
672 filled squares, on left) and expected heterozygosity ( $\sigma H_E$ , open squares, on right). In plots G-I and  
673 J-L a single parameter per plot is shown for each column: G-I = population differentiation ( $F_{ST}$ ); J-  
674 L = spatial genetic structure (Sp). Range size shown in columns A, D, G, J: seed dispersal vector in  
675 columns B, E, H, K: and successional stage in C, F, I, L. The index of the relative importance of  
676 each predictor variable ( $AICc_i$ ) is shown. All samples sizes are in Table 1.