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## 3 **Dynamic seed zones to guide climate-smart seed sourcing for tropical dry** 4 **forest restoration in Colombia**

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40 **Abstract**

41 Tree-based forest landscape restoration interventions require knowledge on the suitability and origin of  
42 seed sources and planting material. A common recommendation is to select locally sourced material  
43 based on the assumption that it is well adapted to local environmental conditions and to avoid  
44 introduction of maladapted genes. However, faced with accelerating climate change, it may be prudent  
45 to supplement local provenances with ‘climate-matched’ provenances, i.e. where current climate  
46 conditions are similar to those anticipated in the future at the planting site. Restoration practitioners  
47 usually do not have access to the necessary information to implement such climate-smart seed sourcing.  
48 Here, we combine genetic data of 11 socio-economically important tree species of the tropical dry  
49 forests of Colombia with spatial environmental data to inform the delineation of dynamic seed zones  
50 for the restoration of this highly threatened ecosystem. Analysis of Molecular Variance (AMOVA)  
51 indicates significant population genetic differentiation within all 11 species. We fitted linear mixed  
52 effects models to evaluate if the genetic distance between trees was mainly related to geographic  
53 distance (i.e. isolation by distance; IBD), environmental distance (i.e. isolation by environment; IBE),  
54 or both. Observed scales of genetic differentiation were best explained by the model including both  
55 geographic and environmental distance (IBD + IBE) for 6 out of 11 species, and by the IBE model for  
56 the remaining species, suggesting that the observed differentiation is at least partly driven by adaptive  
57 processes. Aiming at capturing as much as possible of the observed genetic differentiation, we propose  
58 a set of 36 provisional seed zones that are applicable across species and dynamic under climate change,  
59 based on the clustering of environmental data and geographical coordinates. We project these seed  
60 zones to future climate conditions using five general circulation models and two emission scenarios,  
61 and discuss how they can be used to implement different climate-smart seed sourcing strategies in a  
62 pragmatic way. The seed zone maps are made available in a user-friendly online tool.

## 63 1 Introduction

64 In times of unprecedented anthropogenic pressure on ecosystems worldwide, the restoration of  
65 degraded lands has become a global priority (Aronson and Alexander, 2013; FAO, 2020; Suding et al.,  
66 2015). Restoration efforts often include tree planting, in which case decisions need to be made on the  
67 species to be planted, but also on the provenance of the planting material (also called forest reproductive  
68 material; typically seeds but may also consist of cuttings, stakes, wildlings, etc.. we use ‘seeds’ here for  
69 simplicity). Many provenance trials, also called common garden trials, have found evidence of adaption  
70 of tree species populations to local environmental conditions, although most have been carried out in  
71 temperate regions (e.g., Isaac-Renton et al., 2018; Kreyling et al., 2014; St. Clair, 2006; Vitasse et al.,  
72 2009) and knowledge on adaptive variation in tropical tree species remains scarce (Alberto et al., 2013,  
73 but see for example Barton et al., 2020). As a result of these findings, it is frequently recommended to  
74 source seeds or seedling planting material for restoration locally to avoid maladaptation, while also  
75 reducing the risk of outbreeding depression or erosion of intraspecific genetic diversity (Bischoff et al.,  
76 2010; Hufford and Mazer, 2003; Krauss and He, 2006; McKay et al., 2005; Montalvo and Ellstrand,  
77 2001; Vander Mijnsbrugge et al., 2010).

78 The importance of using local seed for tree planting activities remains a crucial principle to guide seed  
79 sourcing decisions (Pedrini and Dixon, 2020). However, implementing a seed sourcing strategy purely  
80 focussed on local sources may be problematic for a number of reasons. Populations closest to a planting  
81 site may be small and fragmented, resulting in inbred seeds of low genetic diversity (Aguilar et al.,  
82 2006; Breed et al., 2012; Vranckx et al., 2012), making it difficult or impossible to capture sufficient  
83 genetic diversity to establish self-sustaining populations (Broadhurst et al., 2008). In addition, several  
84 studies have found evidence that the geographic distance between individuals or populations is not  
85 always the best indicator of their genetic dissimilarity, which is sometimes more correlated to  
86 environmental distance (i.e. isolation by environment; IBE) than to geographic distance (i.e. isolation  
87 by distance; IDB) (Montalvo and Ellstrand, 2001; Sexton et al., 2014). Furthermore, the magnitude and  
88 rate of climate change is increasingly raising concerns about the ability of local populations to adapt to  
89 future climate conditions (Aitken et al., 2008; Hancock and Hughes, 2014; Vitt et al., 2010) and has  
90 sparked an intense debate on what constitutes an appropriate seed sourcing strategy.

91 Several alternative ‘climate-smart’ or ‘climate-proof’ seed sourcing approaches have been proposed,  
92 often involving the collection of at least a part of the seeds in areas where current climate conditions  
93 are most similar to those anticipated at the planting site under climate change, i.e. ‘climate matching’  
94 or ‘predictive provenancing’ (e.g., Booth, 2016; Broadmeadow et al., 2005; Crowe and Parker, 2008;  
95 Gray and Hamann, 2011; Harrison et al., 2017; Shryock et al., 2018; Thomson et al., 2010). Other  
96 alternative approaches put more focus on increasing genetic diversity and thereby adaptive potential.  
97 These include the ‘composite provenancing’ approach, in which seeds from several local provenances

98 are mixed with progressively smaller amounts of seeds from more distant provenances (Broadhurst et  
99 al., 2008), the ‘admixture provenancing’ approach, in which seeds from different populations are mixed  
100 without considering the geographic or environmental distance to the planting site (Breed et al., 2013),  
101 the ‘regional admixture’ approach, in which seeds are sourced from different populations within a  
102 biogeographic region (or seed zone) (Bucharova et al., 2018), and the ‘climate-adjusted provenancing’  
103 approach, in which seeds are collected in several localities along a gradient coinciding with the direction  
104 of predicted climatic changes (Prober et al., 2015). Although the debate continues, several authors have  
105 provided guidance on how to select the most appropriate strategy depending on species characteristics,  
106 the expected impacts of climate change and the local context (Breed et al., 2013; Havens et al., 2015;  
107 Ramalho et al., 2017; Sgrò et al., 2011).

108 Most of previously proposed climate-smart seed sourcing strategies developed based on temperate  
109 species are not easily implemented in a tropical context, given the large local species pools, scarcity of  
110 data on scales of adaptive variation, generally limited logistic capacity of restoration practitioners and  
111 lack of government incentives to obtain appropriate planting material (Atkinson et al., 2018; Jalonen et  
112 al., 2018). To support climate-smart seed sourcing for a tropical context, we propose a more pragmatic  
113 approach using dynamic seed zones to guide seed sourcing decisions. Seed zones (also called seed  
114 transfer zones, seed provenance zones, or breeding zones) are geographic areas in which planting  
115 material can be moved freely while minimizing the risk for a loss of population fitness and the disruption  
116 of population genetic patterns (Hufford and Mazer, 2003; Miller et al., 2011). Ideally, they should be  
117 informed by insights on intraspecific adaptation patterns (Hufford and Mazer, 2003), traditionally  
118 obtained using provenance trials (Hamann et al., 2000; Kramer et al., 2015; Miller et al., 2011). While  
119 provenance trials remain crucial to inform seed sourcing decisions (Brancalion et al., 2015; Breed et  
120 al., 2018), they do not exist for most species used in restoration, especially in the tropics, prompting  
121 researchers to delineate provisional seed zones using proxies for local adaptation.

122 Based on the notion that seed zones are assumed to reflect genetic differentiation between populations,  
123 several studies have used genetic markers to inform seed sourcing decisions, including both adaptive  
124 markers (e.g., Hufford et al. 2016, Shryock et al. 2017), which are directly under natural selection, and  
125 neutral markers (e.g., Jorgensen et al. 2016, Durka et al. 2017, Listl et al. 2018). While neutral markers  
126 are not directly linked to adaptation, meta-analyses have found that population differentiation in neutral  
127 markers tends to be correlated with differentiation in adaptive markers (Leinonen et al., 2008; Merilä  
128 and Crnokrak, 2001). Other proxies purely based on environmental characterization data have also been  
129 used, such as ecoregions (e.g., Kramer et al., 2015; Miller et al., 2011), selected climate variables (e.g.,  
130 Bower et al. 2014, Castellanos-Acuña et al. 2018), or a wide range of environmental variables (e.g.,  
131 Potter and Hargrove 2012, Crow et al. 2018). These have the advantage that they do not rely on detailed  
132 species-specific data, but it remains uncertain how representative environmentally delimited seed zones  
133 are for mapping genetic and adaptive variability across a species range. In addition, they are not likely

134 to accurately reflect differences in biotic interactions, which may also drive local adaptation, especially  
135 in the tropics (Hargreaves et al., 2020). Nonetheless, one great advantage of environmentally-based  
136 delimitation of seed zones is that it permits to predict how zones are expected to shift under climate  
137 change, i.e. ‘dynamic’ seed zones (Kramer and Havens, 2009; Vitt et al., 2010). Hence, the combination  
138 of environmentally-based limitation of seed zones and species-specific estimates of population genetic  
139 differentiation based on genetic marker data seems a promising approach to develop such dynamic seed  
140 zones.

141 Tropical dry forests (TDFs), also called seasonally dry tropical forests (SDTFs), are among the most  
142 threatened of ecosystems worldwide (Fremout et al., 2020; Hoekstra et al., 2005; Janzen, 1988). In the  
143 Americas, only around one-third of the original TDF cover remains (Portillo-Quintero and Sánchez-  
144 Azofeifa, 2010). Despite this, TDFs receive less scientific attention than more humid tropical forests  
145 and there is a pressing need to step up science-based restoration and conservation efforts in this  
146 ecosystem (Pennington et al., 2018; Schröder et al., 2021). The situation is especially critical in  
147 Colombia, where only around 8% of the original TDF cover remains (García et al., 2014), making it a  
148 national priority for restoration and conservation (Norden et al., 2020; Vargas and Ramírez, 2014).

149 In this study, we used neutral genetic marker data of 11 socio-economically important tree species of  
150 the TDFs of Colombia combined with spatial environmental data to inform the delineation of dynamic  
151 seed zones for guiding climate-smart provenance decisions for the restoration of this highly threatened  
152 ecosystem. First, we use these markers to evaluate if tree species populations are genetically  
153 differentiated, and, whether genetic differences between trees are driven by isolation by distance (IBD),  
154 isolation by environment (IBE), or both. Next, we propose a set of seed zones based on the clustering  
155 of environmental data and geographical coordinates, and project these seed zones to future climate  
156 conditions using different climate models and emission scenarios. Finally, we discuss how these seed  
157 zones can be used to implement several previously proposed climate-smart seed sourcing strategies in  
158 a pragmatic way.

## 159 **2 Methods**

### 160 **2.1 Study region**

161 The study region comprises the potential distribution of TDF in Colombia, i.e. all areas that would be  
162 covered by TDF in the absence of human disturbance, as delimited by the Instituto Alexander von  
163 Humboldt (Rodríguez-Buritacá et al., 2016), building on the work of Etter et al. (2008). The Colombian  
164 TDFs are located mainly along the Caribbean coast and the Inter-Andean valleys of the Cauca,  
165 Magdalena, Chicamocha and Patía rivers. Some authors also include a number of deciduous or semi-  
166 deciduous forest types embedded in a mosaic of savannah and grassland of the Los Llanos region in the  
167 Orinoquia department of north-eastern Colombia, which show similarity with other Colombian TDFs

168 due to drought stress caused by topographic and soil conditions (e.g., rocky outcrops and calcareous  
169 soils) (Pizano et al., 2016), but these forests are not included here. The Caribbean and Inter-Andean  
170 TDFs can be considered as floristically distinct from each other (Banda et al., 2016; González-M et al.,  
171 2018).

172 To avoid any possible omission errors, we slightly extended the potential TDF distribution to include  
173 all areas falling within the climatic definition of TDF, for which we considered a maximum annual  
174 precipitation of 1600 mm, at least 5 months with less than 100 mm precipitation, and a mean annual  
175 temperature of at least 17°C (Murphy and Lugo, 1986; Pennington et al., 2000). Average precipitation  
176 and temperature data were obtained from the WorldClim database at a resolution of 30 arcsec (ca. 0.9  
177 km at the equator) (Hijmans et al., 2005). We did not set a lower limit for precipitation, which resulted  
178 in the inclusion of the entire Guajira peninsula, the driest part of the Colombian TDFs, which in reality  
179 consists of a mosaic of TDF, shrublands, and desert. The reason for this is that populations from the  
180 Guajira peninsula may be useful for climate-smart seed sourcing (see below) to restore areas that are  
181 expected to become drier in the future.

## 182 **2.2 Study species**

183 The study species consist of 11 socio-economically important tree species native to the TDFs of  
184 Colombia, comprising 5 species mostly known for their high-quality timber (*Astronium graveolens*,  
185 *Aspidosperma polyneuron*, *Caesalpinia ebano*, *Cedrela odorata*, *Platymiscium pinnatum*) and 6  
186 multipurpose tree species (*Albizia saman*, *Bursera simaruba*, *Ceiba pentandra*, *Enterolobium*  
187 *cyclocarpum*, *Hura crepitans*, and *Hymenaea courbaril*). Information on their life history traits  
188 (dispersal, pollination, sexual system, and mating system) is given in Table 1. All 11 species are diploid.

189 **Table 1:** Study species with their life history traits. Note that ‘monoecious’ is used in the strict sense here, only  
 190 referring to species with separate male and female flowers on the same plant.

Species	Dispersal	Pollination	Sexual system	Mating system
<i>Albizia saman</i> (Jacq.) F. Muell. (Fabaceae)	Terrestrial mammals*	Bees and moths	Hermaphrodite	Outcrossing (self-incompatible) <sup>1</sup>
<i>Aspidosperma polyneuron</i> Müll.Arg. (Apocynaceae)	Wind (samara fruit)	Moths and butterflies	Hermaphrodite	Mostly outcrossing <sup>2</sup>
<i>Astronium graveolens</i> Jacq. (Anacardiaceae)	Wind (samara-like fruit)	Small bees	Dioecious	Outcrossing (dioecious) <sup>3</sup>
<i>Bursera simaruba</i> (L.) Sarg. (Burseraceae)	Birds, bats	Bees, wasps and wind	Dioecious	Outcrossing (dioecious) <sup>1</sup>
<i>Caesalpinia ebano</i> H.Karst (Fabaceae)	Terrestrial mammals*	Insects	Hermaphrodite	Unknown <sup>4</sup>
<i>Cedrela odorata</i> L. (Meliaceae)	Wind (samara fruit)	Bees and moths	Monoecious	Mostly outcrossing <sup>5</sup>
<i>Ceiba pentandra</i> (L.) Gaertn. (Malvaceae)	Wind (cottony seeds)	Bats	Hermaphrodite	Mixed <sup>6</sup>
<i>Enterolobium cyclocarpum</i> (Jacq.) Griseb. (Fabaceae)	Terrestrial mammals*	Bees and moths	Hermaphrodite	Outcrossing (self-incompatible) <sup>1</sup>
<i>Hura crepitans</i> L. (Euphorbiaceae)	Autochory	Bats	Monoecious	Mostly outcrossing <sup>6</sup>
<i>Hymenaea courbaril</i> L. (Fabaceae)	Terrestrial mammals*	Bats	Hermaphrodite	Outcrossing (self-incompatible) <sup>1</sup>
<i>Platymiscium pinnatum</i> (Fabaceae)	Wind (samara fruit)	Bees	Hermaphrodite	Unknown

191 \* *A. saman*, *E. cyclocarpum*, and *H. courbaril* were primarily dispersed by now extinct megafauna (Janzen and  
 192 Martin, 1982), the same is possibly true for *C. ebano*, which fruits are also hard pods. *H. courbaril* and *E.*  
 193 *cyclocarpum* were part of the diet of Precolumbian societies (Iriarte et al., 2020; Zizumbo-Villarreal et al., 2016)  
 194 and may have undergone considerable human dispersal as a consequence, the same is possibly true for *A. saman*  
 195 which also has edible fruits. Similarly, *A. saman* and *E. cyclocarpum* have been used traditionally as pasture trees  
 196 for centuries (Aguirre-Morales et al., 2020; Thomas et al., 2016).

197 <sup>1</sup>(Bawa, 1974)

198 <sup>2</sup>(Chaves et al., 2016)

199 <sup>3</sup>(Sanchez-Gomez et al., 2020)

200 <sup>4</sup>No info on *C. ebano* was found. Bullock (1985) report mating systems of several *Caesalpinia* species in Mexico,  
 201 but these include entirely outcrossing, mostly outcrossing and mixed reproduction (both selfing and outcrossing).

202 <sup>5</sup>(James et al., 1998)

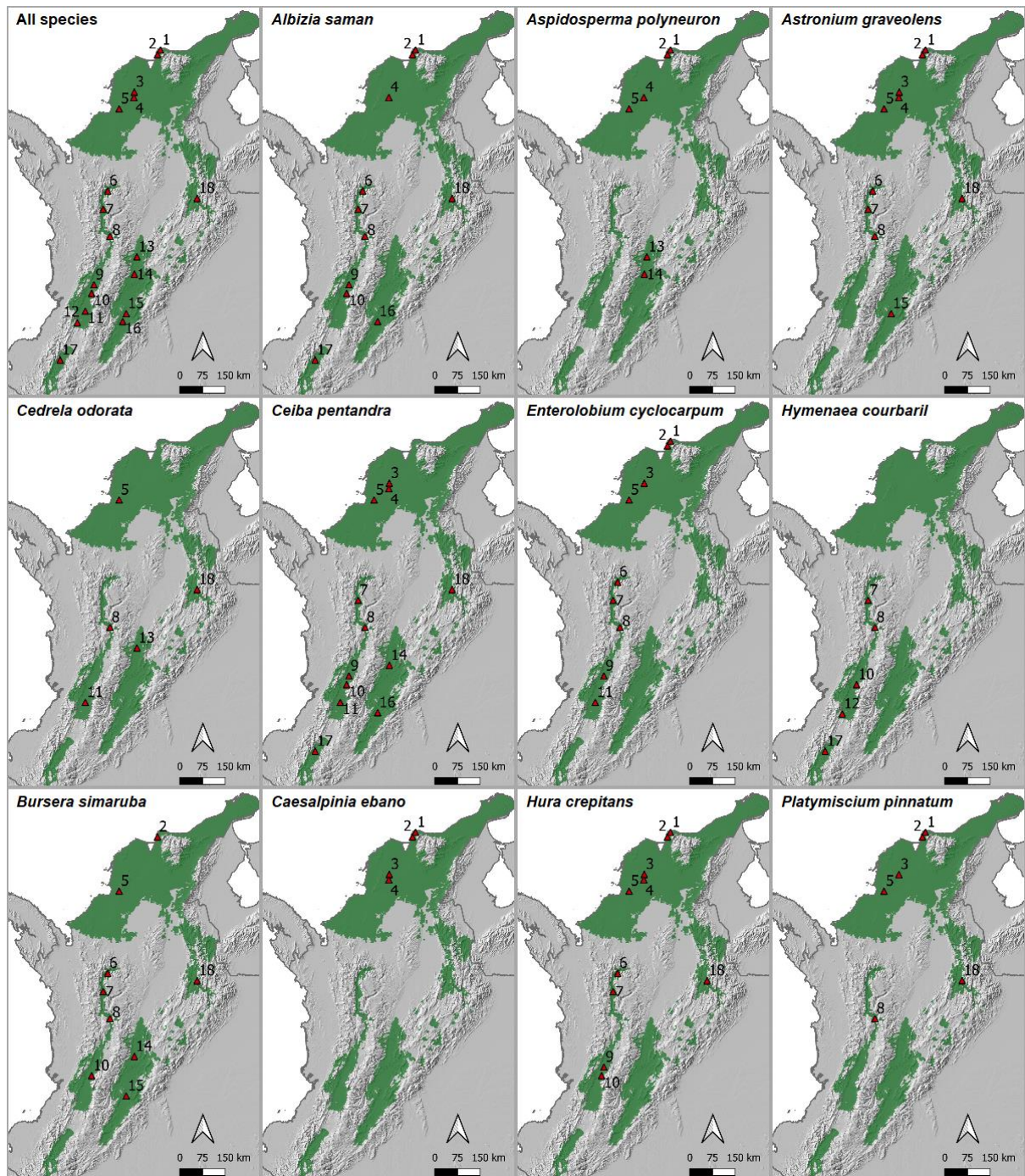
203 <sup>6</sup>(Murawski and Hamrick, 1992)

### 204 **2.3 Field sampling**

205 Field sampling was carried out across the TDFs of Colombia in a total of 18 localities (Figure 1; Table  
206 S1), covering all Colombian TDF regions except the TDFs of the Orinoquia department. Within each  
207 locality, adult trees were sampled within a circular area with radius of ca. 10 km, with selected trees  
208 standing at least 50 meters apart to avoid sampling closely related individuals. The number of sampled  
209 populations per species varied between 4 (*C. ebano*) and 12 (*C. pentandra*). A total of 725 trees was  
210 sampled, with the number of sampled trees per species varying between 37 (*H. courbaril*) and 96 (*A.*  
211 *saman*, *A. graveolens*, *C. pentandra*) (Table 2). All biological materials were collected in collaboration  
212 with the Instituto Alexander von Humboldt following the Colombian Decreto 302 of 2003.

213 **Table 2:** Number of sampled trees per population for the tree species included in this study. The species codes  
 214 refer to the following species: Asa: *Albizia saman*; Apo: *Aspidosperma polyneuron*; Agr: *Astronium graveolens*;  
 215 Bsi: *Bursera simaruba*, Ceb: *Caesalpinia ebano*, Cod: *Cedrela odorata*, Cpe: *Ceiba pentandra*; Ecy:  
 216 *Enterolobium cyclocarpum*; Hcr: *Hura crepitans*; Hco: *Hymenaea courbaril*, Ppi: *Platymiscium pinnatum*. The  
 217 central coordinates and average climatic conditions of the populations are given in Table A.1 (Supplementary  
 218 material).

		Species										
Region	Population	Asa	Apo	Agr	Bsi	Ceb	Cod	Cpe	Ecy	Hcr	Hco	Ppi
Caribbean Coast	Santa Marta N	3	12	9		9			4	5		9
	Santa Marta S	7	1	9	10	12			4	9		7
	El Guamo			10		13		14	10	10		10
	Zambrano	14	9	10		8		11		11		
	Colosó		9	9	11		10	8	3	15		3
Cauca Valley	Ituango	10		9	9				10	10		
	Santa Fe de Antioquia	7		12	11			10	13	9	11	
	La Pintada	12		9	11		11	11	16		10	10
	La Paila	12						12	3	3		
	Mata de Guadua	10			1			4		3	1	
	CIAT						2	3	2			
	Jamundí										7	
Magdalena Valley	Tolima N		5				4					
	Tolima S		9		1			2				
	Tatacoa N			10	10							
	Tatacoa S	6						3				
Patía Valley	Patía	10						8			8	
Chicamocha Valley	Chicamocha	5		9	10		13	12		10		10
	<b>N° of populations</b>	11	6	10	9	4	5	11	9	10	5	6
	<b>Total n° trees</b>	96	45	96	74	42	40	96	65	85	37	49



219

220 **Figure 1:** Sampling locations for all species (first panel 1) and individual species (remaining panels). The potential  
 221 tropical dry forest (TDF) distribution (see section 2.1) is shown in green. Populations are indicated with red  
 222 triangles and are numbered as follows: 1. Santa Marta Norte; 2. Santa Marta Sur; 3. El Guamo; 4. Zambrano; 5.  
 223 Colosó; 6. Ituango; 7. Santa Fe de Antioquia; 8. La Pintada; 9. La Paila; 10. Mata de Guadua; 11. Centro  
 224 Internacional de Agricultura Tropical (CIAT); 12. Jamundí; 13. Tolima Norte; 14. Tolima Sur; 15. Tatacoa Norte;  
 225 16. Tatacoa Sur; 17. Patía; 18. Chicamocha. Populations 1-5 are located along the Carribean coast, localities 6-12  
 226 in the Cauca Valley, localities 13-16 in the Magdalena Valley, locality 17 in the Patía Valley, and locality 18 in  
 227 the Chicamocha Valley.

## 228 2.4 Genetic markers

229 Total genomic DNA was extracted from leaf material using the CTAB procedure (Doyle, J. J. and  
230 Doyle, 1987), with modifications described by Novaes et al. (2009), and was amplified using  
231 polymerase chain reactions (PCR). Seven of the study species (*A. saman*, *A. polyneuron*, *A. graveolens*,  
232 *C. odorata*, *C. pentandra*, *E. cyclocarpum* and *H. courbaril*) were characterized using simple sequence  
233 repeat markers (SSRs), also called microsatellite markers, with the number of loci per species ranging  
234 from 7 (*A. graveolens*) to 12 (*A. saman*) (Table A.2, Supplementary material), whereas the four  
235 remaining species (*B. simaruba*, *C. ebano*, *H. crepitans*, *P. pinnatum*), were characterized using inter-  
236 simple sequence repeat markers (ISSRs), with the number of loci per species ranging from 44 (*H.*  
237 *crepitans*) to 67 (*B. simaruba*) (Table A.2, Supplementary material). SSR markers are codominantly  
238 inherited whereas ISSR markers are dominantly inherited, which means that ISSR markers do not allow  
239 to distinguish between heterozygosity and homozygosity. The data of the species characterized with  
240 ISSR markers have been published by Bocanegra-González et al. (2019), those of *A. saman* by Aguirre-  
241 Morales et al. (2020), those of *A. graveolens* by Morillo-Paz (2019), those of *C. odorata* by Aguirre-  
242 Morales (2017), those of *C. pentandra* by Bocanegra-González et al. (2018), and those of *E.*  
243 *cyclocarpum* by Thomas et al. (2016). Those of *A. polyneuron* and *H. courbaril* have not yet been  
244 published. Although similar, SSR and ISSR markers may result in different estimates of population  
245 genetic differentiation (Ganopoulos et al., 2011; Li et al., 2017; Rawat et al., 2014), but comparisons  
246 between species were not a primary objective of this study.

## 247 2.5 Environmental variables

248 A set of 30 climate and soil variables were selected to evaluate IBE (see section 2.6) and to construct  
249 seed zones (see section 2.7) (Table A.1, Supplementary material). These comprise 19 bioclimatic  
250 variables from the WorldClim v1.4 database (Hijmans et al., 2005), aridity (i.e. annual precipitation  
251 divided by annual potential evapotranspiration), which was calculated using WorldClim temperature  
252 and precipitation data and the Hargreaves formula for estimating potential evapotranspiration  
253 (Hargreaves and Allen, 2003), and 9 continuous soil variables from the ISRIC SoilGrids250m v1  
254 database (Hengl et al., 2017). All variables were used at a spatial resolution of 30 arcsec (ca. 0.9 km at  
255 the equator).

256 Future climate data were downloaded from the CCAFS Climate Data Portal ([http://www.ccafs-  
258 climate.org/](http://www.ccafs-<br/>257 climate.org/)). Data were downloaded for the 2050s and 2070s time horizons and two greenhouse gas  
259 emission scenarios, using the representative concentration pathway RCP8.5 as the worst-case scenario  
260 and RCP4.5 as a more optimistic scenario, as forecasted by five general circulation models (GCMs).  
261 These GCMs were selected using the same approach as in Fremout et al. (2020), first selecting all GCMs  
262 that perform better than the median GCM performance (against observed temperature and precipitation  
values) following Knutti et al. (2013) and that are available at the CCAFS Climate Data Portal and next

263 maximizing the dissimilarity between GCMs by selecting the GCM with the best performance in each  
264 node of the GCM family tree of Knutti et al. (2013), determined after cutting the tree at level 16  
265 (Schlaepfer et al., 2017). This procedure resulted in the selection of 5 GCMs: CESM1(CAM5), GFDL-  
266 CM3, HADGEM2-ES, MIROC5, and MPI-ESM-LR.

## 267 **2.6 Population genetic structure and isolation by distance and environment**

268 To assess whether populations were genetically differentiated, we first calculated the genetic distance  
269 between sampled trees. The simple matching coefficient was used to express genetic distance for the  
270 species characterized by ISSR markers, whereas the Kosman-Leonard distance was used for the species  
271 characterized by the codominant SSR markers, for which traditional genetic distance measures are not  
272 appropriate (Kosman and Leonard, 2005). We calculated these genetic distances using the ‘ade4’ and  
273 ‘mmod’ packages for R (Dray and Dufour, 2007; Winter, 2012). Next, these distances were used to test  
274 for genetic differentiation between the sampled populations by carrying out an Analysis of Molecular  
275 Variance (AMOVA; Excoffier et al. 1992) for each of the species, using the ‘adegenet’ and ‘ade4’  
276 packages for R (Dray and Dufour, 2007; Jombart and Ahmed, 2011). AMOVA, which results in  $\Phi_{ST}$   
277 statistics analogous to  $F_{ST}$  statistics (Wright, 1949), accommodates both dominant and codominant  
278 marker data and works relatively well with limited numbers of loci (Nelson and Anderson, 2013).  
279 Statistical significance of population genetic differentiation was assessed with a permutation test ( $n =$   
280 1000) using the ‘randtest’ function of the ‘ade4’ package, which involves the random permutation of  
281 the rows of the genetic distance matrix (Excoffier et al., 1992). When overall differentiation was  
282 significant, pairwise AMOVA analyses were carried out to evaluate differentiation between individual  
283 populations.

284 To evaluate IBD and IBE, we compared the pairwise genetic distances between trees described above  
285 with the corresponding geographical and environmental distances. To remove multicollinearity between  
286 environmental variables, environmental distances between trees were calculated as the Euclidean  
287 distance in the multi-dimensional space made up by the first  $n$  PCs (principal components) derived from  
288 a principal component analysis (PCA) on the entire set of environmental variables, with  $n$  the number  
289 of PCs having an eigenvalue higher than one (i.e. the Kaiser-Guttman criterion; Guttman, 1954).  
290 Geographical distance was calculated as the Euclidean distance between tree locations. Next, we used  
291 linear mixed effects models (LMMs) to regress genetic distance against geographic distance and  
292 environmental distance. LMMs have been found to outperform other regression methods for landscape  
293 genetic analysis (Shirk et al., 2018), including Mantel tests, which have been criticized in the context  
294 of spatial analysis (Guillot and Rousset, 2013; Legendre et al., 2015). Following the maximum  
295 likelihood population effects framework (MLPE; Clarke et al. 2002), we accounted for the non-  
296 independence of pairwise comparisons using random effects, with tree identity and population as nested  
297 random effects. Geographic distance and environmental distance were included as fixed effects.

298 For each species, three LMMs were fit using the ‘lme4’ package for R (Bates et al., 2015), using (i)  
299 geographic distance, (ii) environmental distance, and (iii) both geographic and environmental distance  
300 as fixed effects. Each of these LMMs was then compared with a null model including only random  
301 effects, using the Akaike Information Criterion (AIC) corrected for small sample sizes (Hurvich and  
302 Tsai, 1989). Models were with fit with maximum likelihood (ML) rather than REML (restricted  
303 maximum likelihood) to obtain valid AIC scores (Clarke et al., 2002; Shirk et al., 2018). Models with  
304 AIC scores less than ten units lower than the AIC score of the null model were considered to have  
305 insufficient support (Burnham and Anderson, 2004). For each species, the best performing model (IBD,  
306 IBE, or IBD + IBE) was selected based on the AIC scores. In addition, the marginal R-squared (i.e. the  
307 proportion of variance explained by the fixed effects) was estimated for each of the fitted LMMs  
308 following Nakagawa et al. (2017), using the ‘MuMIn’ package for R (Barton, 2016).

## 309 **2.7 Seed zones**

310 To guide seed sourcing decision in the TDFs of Colombia, we aimed at constructing seed zones that are  
311 (i) applicable across tree species, (ii) dynamic in the light of ongoing climate change, and (iii) able to  
312 capture as much as possible of the observed population genetic differentiation of the study species, but  
313 large enough to keep it practically feasible to implement the seed zones on the ground. We constructed  
314 geo-environmental seed zones with the CLARA (Clustering Large Applications) approach, an extension  
315 for large datasets of the partitioning-around-medoids (PAM) algorithm, which represents a more robust  
316 alternative of the k-means algorithm (Kaufman and Rousseeuw, 1990). To reflect both environmental  
317 conditions and geographical location, the environmental variables, which were represented by the same  
318 PCs described in the previous section, were clustered along with longitude and latitude. Longitude and  
319 latitude were rescaled to have a variance equal to the average variance of the first six PCs. The number  
320 of clusters (i.e. number of seed zones) was varied between 5 and 50. To evaluate the performance of  
321 the resulting seed zone maps, we carried out additional AMOVAs to estimate the percentage of  
322 molecular variance explained by each of these seed zone scenarios. The optimal number of seed zones  
323 for each of the species was estimated as the number that explained most of the molecular variance. In  
324 order to obtain a seed zone map applicable across species, we selected the seed zone map with the  
325 highest number of zones among these optima. To reduce the occurrence of isolated patches in this  
326 proposed seed zone map, it was subject to a modal filter with a window of 3x3 grid cells (i.e. each grid  
327 cell was assigned to the most common seed zone in the 3x3 window).

328 To project the current seed zones to future climate conditions, the predicted future environmental  
329 conditions of every grid cell within the potential TDF distribution were projected to the same PCs as  
330 those used for the clustering of the current seed zones (i.e. the variables describing the future  
331 environmental conditions were recombined using the same linear combinations of variables as defined  
332 by the PCA that was carried out based on the current environmental variables), after which each grid

333 cell was assigned to the closest cluster medoid, using the ‘cl\_predict’ function of the ‘clue’ package for  
 334 R (Hornik, 2019). In this way, the current seed zones were projected to future climate conditions as  
 335 predicted by each of the selected GCMs under emission scenarios RCP4.5 and RCP8.5 for the 2050s  
 336 and 2070s (section 2.5).

### 337 **3 Results**

338 We found significant genetic differentiation between populations for all 11 species, with the percentage  
 339 of genetic variance between populations ranging from only 4% for *E. cyclocarpum* to 28% for *P.*  
 340 *pinnatum*. Overall population differentiation was highly significant ( $P < 0.001$ ) for all species except for  
 341 *E. cyclocarpum* ( $P = 0.02$ ). For most species, at least half of pairwise comparisons between populations  
 342 was significant, except for *E. cyclocarpum* (5 out of 36 comparisons) and *A. saman* (23 out of 55  
 343 comparisons). *C. ebano* was the only species for which all pairwise combinations of populations were  
 344 significantly differentiated (6 out of 6 comparisons) (Table 3).

345 **Table 3:** AMOVA (Analysis of Molecular Variance) partitioning of genetic variance for the tree species included  
 346 in this study, showing the results of the overall AMOVA in the second column and results of the pairwise AMOVA  
 347 analyses in the third column. All overall AMOVA analyses were statistically significant ( $P < 0.05$ ). SSR and ISSR  
 348 stand for simple sequence repeats and inter-simple sequence repeats, respectively.

Species	% Variance between populations ( $\Phi_{ST}$ )	Number of significant pairwise comparisons
<b>SSR markers</b>		
<i>Albizia saman</i> (B)	14	24/55
<i>Aspidosperma polyneuron</i> (A)	15	16/28
<i>Astronium graveolens</i> (A)	7	26/45
<i>Cedrela odorata</i> (A)	18	5/10
<i>Ceiba pentandra</i> (A)	10	38/66
<i>Enterolobium cyclocarpum</i> (B)	4	6/36
<i>Hymenea courbaril</i> (B)	10	5/10
<b>ISSR markers</b>		
<i>Bursera simaruba</i> (B)	19	18/21
<i>Caesalpinia ebano</i> (B)	19	6/6
<i>Hura crepitans</i> (A)	25	31/45
<i>Platymiscium pinnatum</i> (A)	27	11/15

349 Table 4 presents the results of the pairwise AMOVA analyses in more detail. Genetic differentiation  
 350 not only occurred between the five biogeographic regions where sampling was carried out but also  
 351 within these regions, with 47% of the pairwise population combinations within regions being  
 352 significantly differentiated. For example, the Santa Fe de Antioquia and Ituango populations, despite  
 353 being separated by only ca. 100 km and being both located in the Cauca river valley, were genetically

354 differentiated for 3 out of 5 species (Table 4). This is further illustrated by the significant differentiation  
355 between all four populations of *C. ebano*, which are all located in the same region (Caribbean coast).  
356 Further, Table 4 shows that the Patía population was clearly the most differentiated from the other  
357 sampled populations, with significant pairwise differentiation for all but two of the pairwise  
358 comparisons, and high percentages of genetic variance between populations, as high as 50% and higher  
359 (Table 4). While genetic differentiation between the five sampling regions was often significant, this  
360 was not always the case. For example, the populations in the south of the Magdalena valley (TAN:  
361 Tatacoa N; TAS: Tatacoa) were only differentiated from the populations in the Cauca valley in 50% of  
362 the comparisons.

363 **Table 4:** Pairwise AMOVA (Analysis of Molecular Variance) results. The lower triangular matrix shows the number of species for which each of the pairwise comparisons  
364 indicated significant genetic differentiation ( $P < 0.05$ ), as the fraction of the total number of species for which pairwise comparisons were made. The upper triangular matrix  
365 indicates the average percentage of variance situated between the two populations (calculated from all the pairwise comparisons, not only those that were statistically significant).  
366 Dashes (“-”) indicate pairs of populations for which no comparisons could be made. Populations are abbreviated as follows: SMN: Santa Marta N; SMS: Santa Marta S; GUA:  
367 El Guamo; ZAM: Zambrano; COL: Colosó; ITU: Ituango; SFE: Santa Fe de Antioquia; PIN: La Pintada; PAI: La Paila; MAT: Mata Guadua; CIA: Centro Internacional de  
368 Agricultura Tropical; JAM: Jamundí; TON: Tolima N; TOS: Tolima S; TAN: Tatacoa N; TAS: Tatacoa S; PAT: Patía; CHI: Chicamocha.

Biogeographic region		SMN	SMS	GUA	ZAM	COL	ITU	SFE	PIN	PAI	MAT	CIA	JAM	TON	TOS	TAN	TAS	PAT	CHI
Caribbean coast	SMN		11	12	15	15	17	8	14	21	35	0	-	27	22	2	4	59	20
	SMS	3/7		10	10	13	15	11	11	9	15	0	-	23	11	10	5	49	19
	GUA	2/5	3/5		10	5	13	2	11	18	13	9	-	-	26	0	1	35	13
	ZAM	4/5	2/5	4/4		11	15	10	12	15	16	16	-	22	22	4	1	44	18
	COL	2/5	4/6	1/5	2/4		21	8	14	13	6	3	-	18	12	7	0	28	19
Cauca valley	ITU	2/4	4/5	2/3	3/3	3/4		13	11	17	14	0	-	-	-	19	11	50	25
	SFE	0/4	3/5	0/4	2/4	2/5	3/5		7	11	-7	5	15	-	17	4	-3	29	15
	PIN	2/4	2/5	3/4	3/3	3/6	1/4	2/6		14	12	9	20	5	13	14	5	31	18
	PAI	1/3	1/3	2/3	3/3	2/3	3/3	2/4	3/3		0	0	-	-	24	-	12	45	16
	MAT	1/2	0/2	1/2	2/3	1/2	1/2	1/4	0/3	0/3		0	35	-	21	-	13	25	15
Magdalena valley	CIA	0/1	0/1	1/2	1/1	0/3	0/1	0/2	0/3	0/2	0/1		-	2	18	-	10	34	7
	JAM	-	-	-	-	-	-	1/1	1/1	-	0/1	-		-	-	-	-	12	-
	TON	1/1	1/1	-	1/1	2/2	-	-	0/1	-	-	0/1			6	-	-	-	13
	TOS	1/1	0/1	1/1	2/2	1/2	-	1/1	0/1	1/1	1/1	0/1	-	0/1		29	21	29	20
	TAN	0/1	1/2	0/1	0/1	1/2	2/2	0/2	1/2	-	-	-	-	-	-	-	-	-	8
Patía valley	TAS	0/1	0/1	0/1	0/2	0/1	1/1	0/2	1/2	2/2	0/2	0/1	-	-	1/1	-		50	7
	PAT	1/1	1/1	1/1	2/2	1/1	1/1	2/3	3/3	2/2	2/3	1/1	1/1	-	1/1	-	2/2		43
Chicamocha valley	CHI	2/4	4/5	3/4	3/4	4/6	3/4	4/5	4/6	3/3	1/3	0/2	-	1/1	0/1	1/2	0/2	2/2	

369

370 The first six PCs resulting from the PCA of the environmental variables (n = 30) together explained  
 371 85% of the total variance. Already 60% was explained by the first two PCs only, mainly reflecting  
 372 variation in temperature and precipitation, respectively. The loadings of the environmental variables of  
 373 the first six PCs are given in Table A.4 and the scree plot is given in Figure A.1 (Supplementary  
 374 material).

375 The results of the LMMs that were fit to evaluate IBD and IBE are given in Table 5. For 6 out of 11  
 376 species, the model with the lowest AIC was the model including both geographic distance and  
 377 environmental distance (IBD + IBE) as explanatory variables, whereas the IBE model had the lowest  
 378 AIC for the 5 remaining species (Table 5). On average, environmental distance was a better predictor  
 379 of genetic distance, according to both the AIC and R<sup>2</sup> values (Table 5).

380 **Table 5:** Results of the linear mixed effects models (LMMs) evaluating the relationship between genetic distances  
 381 between trees and corresponding geographical distances (isolation by distance; IBD) and environmental distances  
 382 (isolation by environment; IBE). LMMs were fitted within a maximum likelihood population effects (MLPE)  
 383 framework, using trees and population as nested random effects and geographical and environmental distance as  
 384 fixed effects. The reported R<sup>2</sup> values are marginal R<sup>2</sup> values only reflecting the variance explained by the fixed  
 385 effects. ΔAIC scores refer to the difference between the AIC score of the fitted model and the AIC score of a null  
 386 model with only random effects. All AIC scores were corrected for small sample sizes. Models with ΔAIC scores  
 387 higher than -10 were not considered to have sufficient support and are not shown. For each species, the best model  
 388 (last column) was identified based on the lowest ΔAIC.

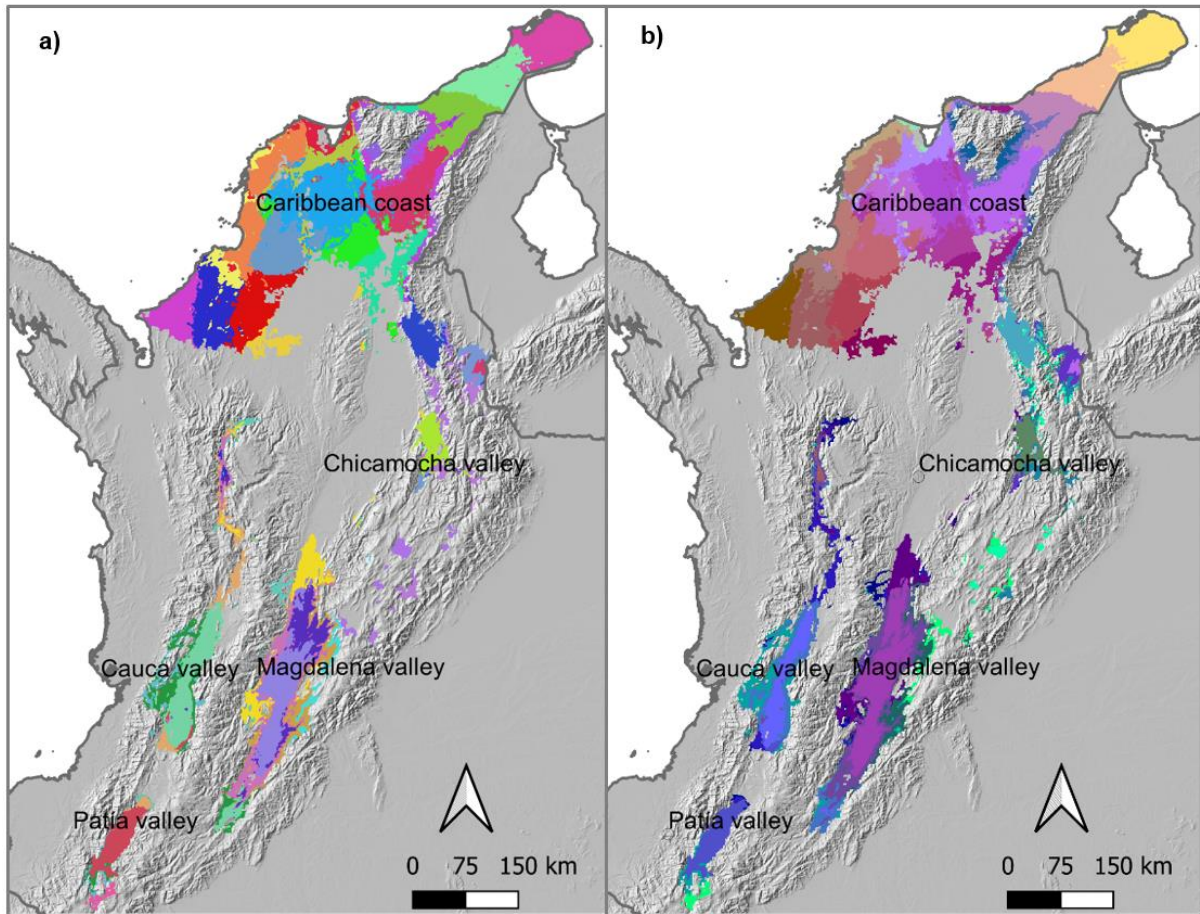
	Isolation by distance (IBD)		Isolation by environment (IBE)		IBD + IBE		Best model
	ΔAIC	R <sup>2</sup>	ΔAIC	R <sup>2</sup>	ΔAIC	R <sup>2</sup>	
<b>SSR markers</b>							
<i>Albizia saman</i>	-118.7	0.03	-12.3	0.00	-156.1	0.05	IBD + IBE
<i>Aspidosperma polyneuron</i>	-79.0	0.10	-20.2	0.03	-79.2	0.10	IBD + IBE
<i>Astronium graveolens</i>	-11.3	0.00			-13.3	0.00	IBD + IBE
<i>Cedrela odorata</i>	-36.5	0.06	-77.3	0.11	-75.7	0.11	IBE
<i>Ceiba pentandra</i>	-101.7	0.02	-12.3	0.00	-113.1	0.03	IBD + IBE
<i>Enterolobium cyclocarpum</i>	-38.1	0.02	-34.0	0.02	-41.8	0.02	IBD + IBE
<i>Hymenea courbaril</i>			-18.8	0.03	-17.5	0.03	IBE
<b>ISSR markers</b>							
<i>Bursera simaruba</i>	-11.4	0.01	-198.5	0.17	-196.9	0.17	IBE
<i>Caesalpinia ebano</i>	-60.4	0.11	-65.2	0.13	-64.2	0.13	IBE
<i>Hura crepitans</i>	-62.4	0.04	-118.7	0.07	-116.8	0.07	IBE
<i>Platymiscium pinnatum</i>	-85.3	0.19	-93.0	0.20	-98.0	0.22	IBD + IBE
<b>Average</b>	-55.7	0.05	-60.0	0.07	-88.4	0.08	

389 The performance of the seed zone maps with different number of seed zones was evaluated by means  
 390 of AMOVAs (Table 6; full results in Table A.5 in Supplementary material). The optimal number of  
 391 seed zones, determined as the number of seed zones explaining most of the genetic variance, ranged  
 392 from 5 (*A. polyneuron*) to 36 (*C. pentandra*). In order to obtain a conservative, risk-minimizing seed  
 393 zone map applicable across species, we selected the base map with 36 zones for further analysis. In  
 394 addition, this approach has the advantage that seed zones could be merged for species with known  
 395 genetic variance which can be explained by fewer zones.

396 **Table 6:** Optimal number of seed zones per species as estimated by the AMOVAs, also showing the percentage  
 397 of genetic variance situated between zones at these numbers, and the percentage of genetic variance situated  
 398 between populations for comparison.

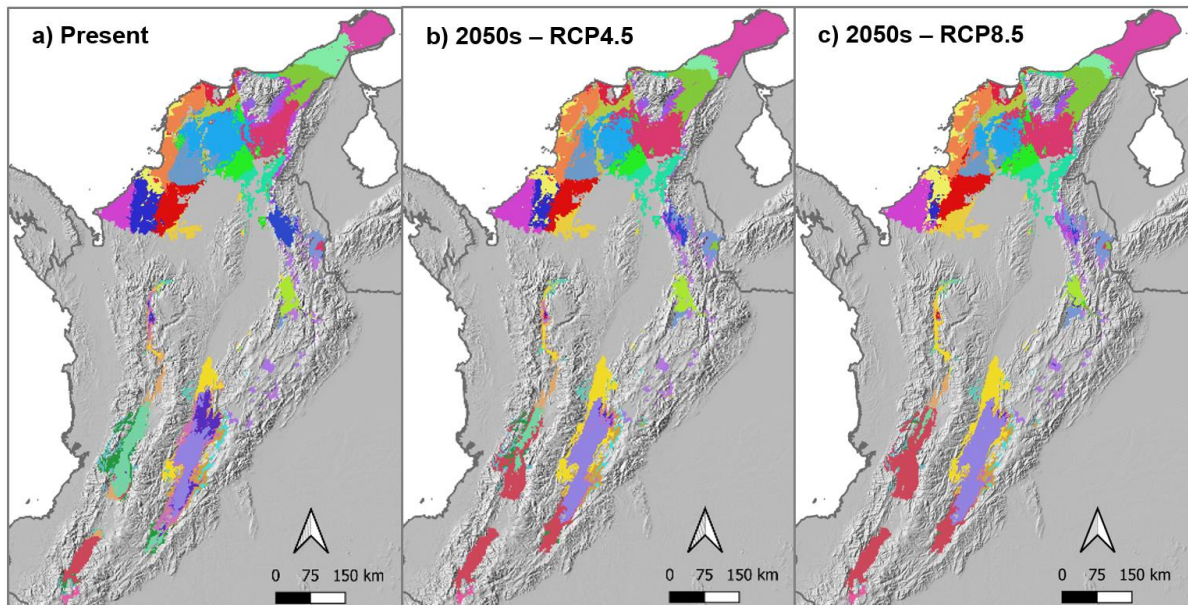
	Optimal n° seed zones	% variance between seed zones	% variance between populations
<b>SSR markers</b>			
<i>Albizia saman</i>	26	17	14
<i>Aspidosperma polyneuron</i>	5	16	15
<i>Astronium graveolens</i>	27	7	7
<i>Cedrela odorata</i>	8	18	18
<i>Ceiba pentandra</i>	36	11	10
<i>Enterolobium cyclocarpum</i>	26	5	4
<i>Hymenea courbaril</i>	20	10	10
<b>ISSR markers</b>			
<i>Bursera simaruba</i>	24	19	19
<i>Caesalpinia ebano</i>	19	18	19
<i>Hura crepitans</i>	27	23	25
<i>Platymiscium pinnatum</i>	27	27	27

399 The seed zones under present climatic conditions are shown in Figure 2. Panel b, depicting the seed  
 400 zones with colours reflecting environmental similarity, shows that some seed zones located far apart  
 401 have similar environmental conditions, as illustrated for example by the purple colors in both the central  
 402 part of the Caribbean TDFs and the Magdalena valley. However, by adding longitude and latitude as  
 403 input variables in the clustering, these areas were assigned to different seed zones, as would be expected  
 404 under an IBD scenario. Figure 3 depicts the seed zones under both present (panel a) and future climate  
 405 conditions (2050s; panels b and c), showing the expected expansion and contraction of seed zones under  
 406 future climatic conditions, as illustrated for example by the predicted southward expansion of the pink  
 407 seed zone of the Guajira peninsula in the northernmost part of Colombia.



408

**Figure 2:** Seed zones ( $n = 36$ ) for the tropical dry forests (TDFs) of Colombia under present climate conditions. The colours of the seed zones in panel a were randomly chosen, whereas the colours in panel b reflect the environmental similarity between seed zones, using a red-green-blue colouring scheme determined by the average scores of the three first principal components (PCs) of the environmental variables, as proposed by Hargrove and Hoffman (2005).



409

410 **Figure 3:** Seed zones ( $n = 36$ ) for the tropical dry forests (TDFs) of Colombia under present climate conditions  
 411 (panel a) and future climate conditions for the 2050s under the representative concentration pathways RCP4.5 and  
 412 RCP8.5 (panel b and c). The future seed zones shown here are those predicted by the HadGEM2-ES model, one  
 413 of the five selected GCMs (section 2.5), but note that the tool in which the seeds zones are integrated  
 414 ([www.diversityforrestoration.org/](http://www.diversityforrestoration.org/)) combines the predictions of all five GCMs.

## 415 **4 Discussion**

### 416 **4.1 Population genetic differentiation**

417 All of the 11 study species showed significant genetic differentiation between populations, with the  
 418 percentage of variance situated between populations ( $\Phi_{ST}$ ) ranging from 4% to 28% (Table 3). These  
 419 figures are similar to other neutral marker characterizations of Neotropical tree species at similar spatial  
 420 scales (e.g., Chase et al. 1995, Lacerda et al. 2001, Lowe et al. 2003, Cerón-Souza et al. 2005) and  
 421 within the range of expected values for outcrossing long-lived species, which tend to retain most of the  
 422 genetic variability within populations (Hamrick and Godt, 1996; Nybom, 2004; Reisch and Bernhardt-  
 423 Römermann, 2014). The marked differences between species, also between species characterized by  
 424 the same type of markers, suggest that species-specific seed zones may be more appropriate than general  
 425 seed zones. However, collecting genetic data on all the species used in restoration of the Colombian  
 426 TDFs, which are home to several hundreds of tree species, would be a daunting task, and it may be  
 427 unrealistic to expect restoration practitioners to implement different seed zones for each of the species  
 428 used, given the generally limited logistic capacity and lack of incentives to obtain appropriate planting  
 429 material (Atkinson et al., 2018; Jalonen et al., 2018). Trait-based generalizations provide one possible  
 430 solution, but previous research has shown that predicting the degree of population genetic  
 431 differentiation using life history traits is far from an easy undertaking, as relationships between both are

432 generally weak (Duminil et al., 2007). We therefore adopted a more pragmatic approach and aimed to  
433 construct a single set of seed zones applicable across species, drawing on genetic marker data of 11  
434 species with differing life history traits (Table 1).

435 The current difference in population genetic structure between the study species are the result of a  
436 complex interplay of different factors. Past climatic changes are likely to have led to different degrees  
437 of range contraction and expansion and possibly convergence of different genetic groups in some areas  
438 (Aguirre-Morales et al., 2020; Bocanegra-González et al., 2018; Thomas et al., 2016). Using Bayesian  
439 genetic clustering, these studies furthermore suggest that exchange of tree genetic resources across the  
440 Inter-Andean valleys may have taken place during or prior to the Last Glacial Maximum (LGM)  
441 (Aguirre-Morales et al., 2020; Bocanegra-González et al., 2018), which may explain why genetic  
442 differentiation between populations in the Magdalena valley and those in the Cauca valley was often  
443 not significant (Table 4).

444 Other important factors influencing the genetic differentiation between populations are species' seed  
445 dispersal modes, pollination modes, and mating systems (Ballesteros-Mejia et al., 2016; Hamrick et al.,  
446 1993; Lowe et al., 2018), which result in differences in the degree of gene flow. For example, our  
447 findings partly followed the expected pattern that populations of animal-dispersed species tend to be  
448 less genetically differentiated (average  $\Phi_{ST}$  = 13.2%) than those of abiotically dispersed species (e.g.  
449 wind, autochory; average  $\Phi_{ST}$  = 15.0%) as a consequence of gene flow over larger distances in animal-  
450 dispersed species (Hamrick et al. 1993, Lowe et al. 2018, but see Duminil et al. 2007), although some  
451 wind-dispersed seeds can be dispersed over extremely large distances (e.g., *C. pentandra*; Dick et al.,  
452 2007). The species with the lowest degree of differentiation, *E. cyclocarpum* (Table 3), used to be  
453 dispersed by Pleistocene horses (Janzen and Martin, 1982), but has long been widely planted in pasture  
454 lands and is now mainly dispersed by introduced cattle and horses (Gonzales et al., 2010; Janzen, 1982),  
455 resulting in large distance dispersal when these animals are moved around by humans. Human-mediated  
456 dispersal may also have become the dominant dispersal mode of other species originally dispersed by  
457 megafauna, including *H. courbaril* and *A. saman* (Janzen and Martin, 1982). *E. cyclocarpum* and *H.*  
458 *courbaril* also have a long history as human food (Iriarte et al., 2020; Zizumbo-Villarreal et al., 2016)  
459 and the same is possibly true for *A. saman*. However, as each of the factors mentioned above may have  
460 affected the genetic structure of each of the species differently, it is difficult to explain the findings of  
461 the AMOVAs (Table 3). In addition, the differences between species in our study should be interpreted  
462 with care because not all species were present in all sampling sites and because two different marker  
463 types were used. Rigorous hypothesis testing on the relationship between life history traits and  
464 population genetic structure was therefore not part of the scope of the present study. Rather, the value  
465 of the selected set of species lies in the broad range of life history traits they represent, increasing the  
466 likelihood that the obtained results are applicable across many tree species of the Colombian TDFs.

467 For more than half (6 out of 11) of the study species, genetic differences between trees was best  
468 explained using both geographical and environmental distance, while they were best explained by  
469 environmental distance for the remaining species (Table 5), suggesting that the observed genetic  
470 differentiation is a consequence of both neutral and adaptive processes, in line with similar findings  
471 elsewhere (Sexton et al. 2014). The obtained  $R^2$  values were relatively low for most species, which was  
472 expected since the studied species retain the majority of genetic variability within populations, within  
473 which both geographic and environmental distances are limited. As discussed by Sexton et al. (2014),  
474 IBE results from reduced gene flow between populations growing under divergent environmental  
475 conditions, which may be the consequence of natural selection, with maladapted immigrants or  
476 offspring of local and immigrant parents failing to germinate, survive or reproduce, and hence failing  
477 to introduce their genes in the local population. It may also be caused by nonrandom mating resulting  
478 from different environmental conditions, both as a consequence of genetic adaptation (e.g., flowering  
479 period mismatch between immigrants and locals) or phenotypic plasticity (e.g., plants growing at  
480 different altitudes flowering at different times). Without provenance trials, it is difficult to determine  
481 the relative importance of these different mechanisms (Sexton et al., 2014).

482 Neutral markers have been used extensively to inform provenancing decisions (Durka et al., 2017;  
483 Jorgensen et al., 2016; Krauss et al., 2013; Krauss and He, 2006; Listl et al., 2018; Malaval et al., 2010;  
484 Massatti et al., 2020). However, they are not directly linked to local adaptation (Holderegger et al.,  
485 2006), which has led to criticism on their use in guiding seed provenancing (McKay et al., 2005).  
486 Nevertheless, meta-analyses have found that differentiation in adaptive and neutral markers tends to be  
487 correlated (Leinonen et al., 2008; Merilä and Crnokrak, 2001). Divergent natural selection (i.e. selection  
488 acting in different directions on different populations) can not only lead to population differentiation by  
489 acting on specific loci or those physically associated with them, but can also promote barriers to gene  
490 flow as discussed above. This results in genome-wide neutral divergence via genetic drift, which is  
491 readily detectable by neutral markers (Nosil et al., 2009).

## 492 **4.2 Seed zones**

493 The population genetic differentiation we found in all the study species (Table 3) supports the need for  
494 establishing seed zones to promote adaptedness of seeds used in tree planting activities and to avoid  
495 disruptions of genetic patterns. Pairwise genetic differentiation between populations within the five  
496 sampled biogeographic regions was often significant (47% of within-region comparisons were  
497 significant; Table 4), indicating that population differentiation occurs at a finer scale than these five  
498 regions, which should be reflected in a sufficiently high number of seed zones. Further, as our results  
499 indicate that the genetic differences between trees are related to both geographic distance and  
500 environmental differences (Table 5), we included longitude and latitude along with the PCs derived  
501 from environmental variables in the PAM clustering, seeking to construct seed zones that are

502 environmentally homogeneous while also avoiding large geographic distances between locations within  
503 the same seed zone. This reduces the probability that isolated areas with similar environmental  
504 conditions would be grouped in the same seed zone, which is a disadvantage of seed zones based on  
505 clustering environmental variables only (Potter and Hargrove, 2012).

506 Non-hierarchical clustering methods such as k-means clustering or PAM have been used before to  
507 delineate seed zones (Potter and Hargrove, 2012; Shryock et al., 2018) or ecoregions (Hargrove and  
508 Hoffman, 2005). Compared to hierarchical clustering, these methods have the advantage that they result  
509 in similar environmental heterogeneity within different seed zones (Hargrove and Hoffman, 2005). We  
510 followed Shryock et al. (2018) by submitting the environmental variables to a PCA prior to clustering,  
511 which is more robust against the inclusion of correlated variables, and does not require prior knowledge  
512 about which environmental variables are most closely linked to local adaptation of tree species  
513 populations. We generated seed zone maps with differing number of seed zones, ranging from 5 to 50,  
514 and determined the optimal number of zones by carrying out an AMOVA for each of these numbers of  
515 zones. However, it should be noted that the sampling density was relatively low, and that the number  
516 of seed zones proposed here should be seen as the optimal scenario based on currently available  
517 information. In order to more thoroughly study the spatial scale of genetic differentiation, a more dense  
518 sampling scheme is recommended, in analogy with Malaval et al. (2010) or Michalski and Durka  
519 (2012).

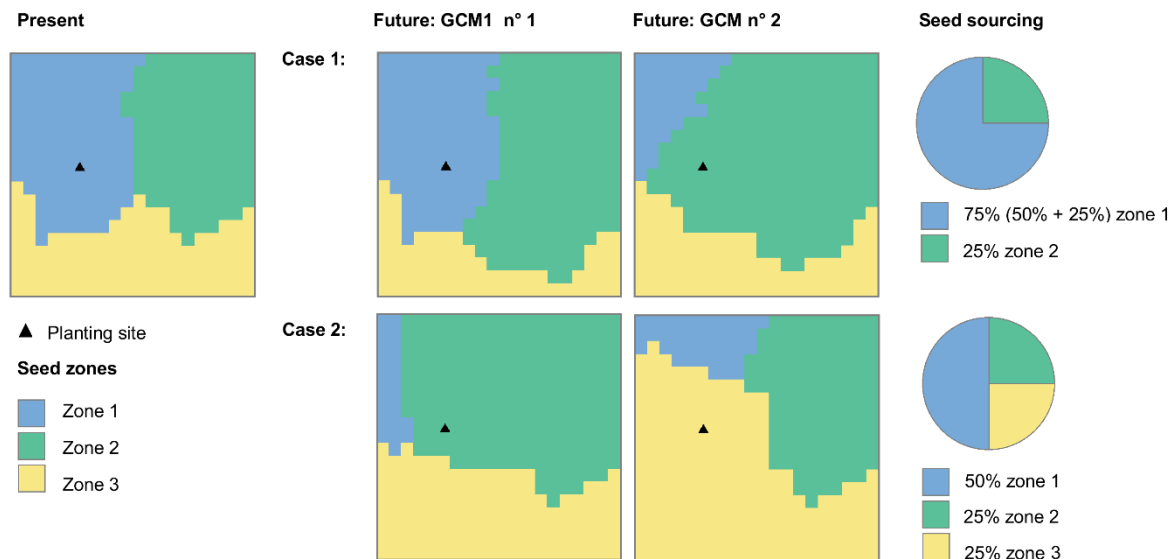
520 We projected the proposed set of seed zones to the climate conditions anticipated in the 2050s and  
521 2070s by assigning the future conditions of each of the grid cells to the closest cluster medoid resulting  
522 from the CLARA algorithm using the present environmental conditions as an input. As such, a grid cell  
523 is assigned to another seed zone under future climate conditions only if its distance in PCA space to the  
524 cluster medoid of that other seed zone is shorter than to the medoid of the present seed zone. By also  
525 including longitude and latitude in the clustering, the probability that grid cells are assigned to another  
526 seed zone under future climate conditions decreases with the geographic distance to the grid cell  
527 representing the cluster medoid of the latter zone. Grid cells changing from one seed zone to another  
528 result in the contraction and expansion of seed zones under future climate conditions, for example seed  
529 zones that are currently already characterized by high temperatures are likely to expand under future  
530 conditions.

531 Our approach to predict shifts in seed zones from present to future climate condition is similar to the  
532 multivariate spatiotemporal clustering (MSTC) approach proposed by Hargrove and Hoffman (2005),  
533 with the difference that in MSTC, predicted future environmental conditions are used directly as input  
534 variables in the clustering along with the present environmental conditions. Consequently, when some  
535 of the predicted future conditions are very dissimilar to any of the current conditions, some clusters in  
536 MSTC may only exist under future conditions (i.e. represent novel climates). While such novel climates

537 are likely to occur in the future, in the context climate-smart seed sourcing, we considered it more  
538 important to identify the current seed zone where conditions are most similar to the expected novel  
539 climate.

540 The aim of the seed zones presented here is to promote the use of planting material that is well-adapted  
541 to present and future climate conditions in the TDFs of Colombia. To facilitate their use, the zones have  
542 been integrated into a map-based online decision-support tool (DiversityForRestoration, available at  
543 [www.diversityforrestoration.org](http://www.diversityforrestoration.org); Thomas et al. 2017a). The tool recommends to supplement planting  
544 material from the present seed zone with material from areas currently located in the seed zone that is  
545 anticipated at the planting site under future climate conditions, i.e. combining local provenancing with  
546 predictive provenancing. While the tool recommends a 50/50 ratio, some restoration planners may  
547 consider the risk of introducing 50% material from a non-local seed zone too high, and may opt for only  
548 introducing only a smaller amount of non-local seeds, coinciding with the ‘genetic enrichment’  
549 approach proposed by Lefèvre et al. (2013). Both local and predictive provenancing may include one  
550 or more source populations (more is better to increase adaptive potential, but logistical costs may be a  
551 constraint).

552 While all five GCMs predict marked increases in temperature in the TDFs of Colombia, the direction  
553 of predicted precipitation changes is not always consistent. As a result, we found that different GCMs  
554 did not always coincide in the future seed zone projections, in which case the tool recommends sourcing  
555 proportional parts of the planting material in each of the future seed zones as predicted by different  
556 GCMs. This approach coincides with the risk-minimizing portfolio approach proposed by Crowe and  
557 Parker (2008), directly incorporating the uncertainty of future climate predictions, as illustrated in  
558 Figure 4.



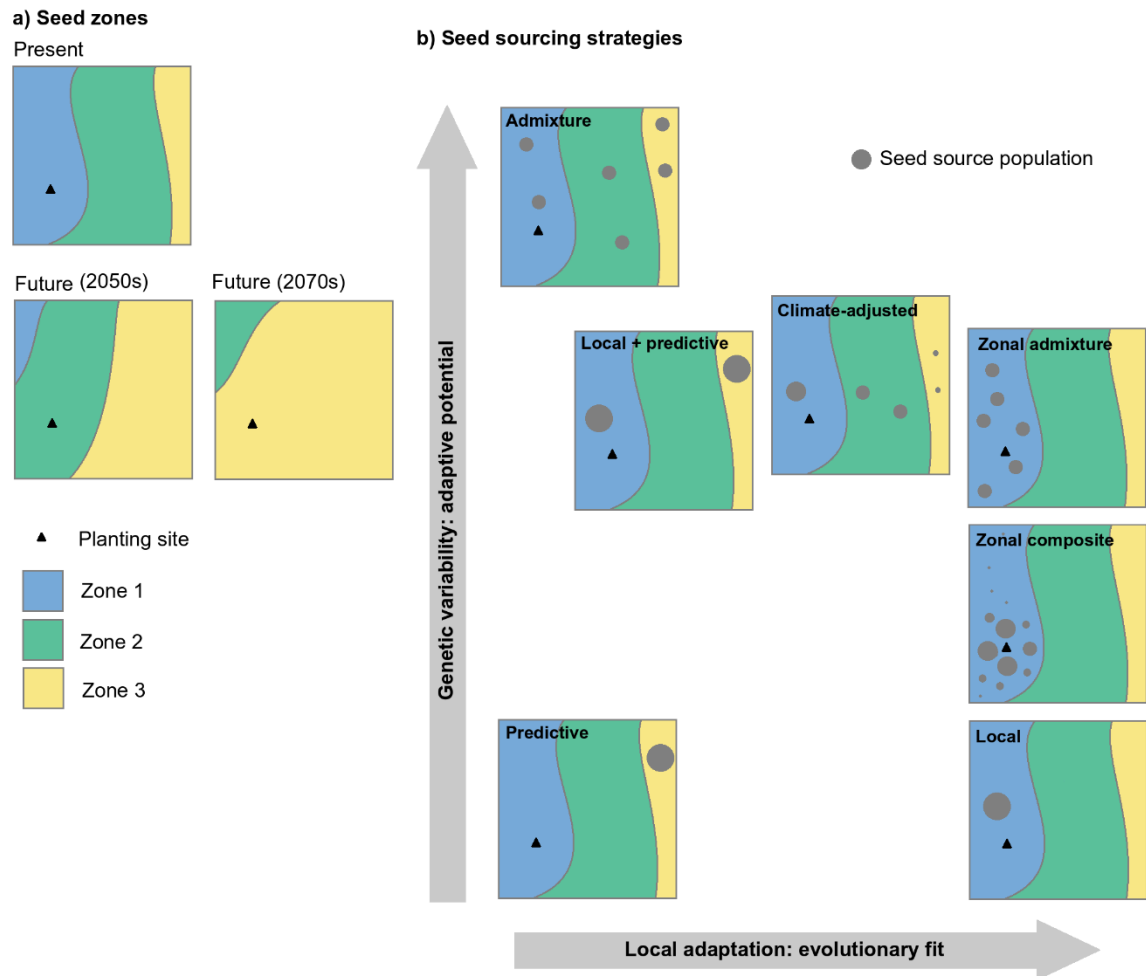
559

560 **Figure 4:** Illustration of the risk-minimizing seed sourcing strategy (‘portfolio approach’) proposed in the  
 561 DiversityForRestoration decision support tool ([www.diversityforrestoration.org](http://www.diversityforrestoration.org)), which involves sourcing 50%  
 562 of the seeds in the present seed zone, and 50% in the future seed zones as predicted by different General  
 563 Circulation Models (GCMs), recommending to source from different future seed zones if different GCMs do not  
 564 coincide in their predictions of future seed zones. This is illustrated here with two GCMs for simplicity, but note  
 565 that 5 GCMs were used (section 2.5).

### 566 4.3 Final considerations and prospects

567 Climate-smart seed sourcing is a complex issue and the ‘local + predictive’ strategy we adopted in the  
 568 DiversityForRestoration tool is only one of several possible strategies. The seed zones presented here  
 569 can be used to implement other strategies too, as illustrated in Figure 5. While the inclusion of longitude  
 570 and latitude in the clustering ensures that seed zones do not expand over very large geographic distances,  
 571 limiting the distances over which predictive provenancing is recommended, some restoration planners  
 572 may prefer to not carry out any assisted gene flow at all, as experimental evidence of its potential  
 573 remains limited and some concerns remain about the risk of outbreeding depression and mismatches in  
 574 biotic interactions (Aitken and Whitlock, 2013; Bucharova, 2017; Bucharova et al., 2018). To increase  
 575 adaptive capacity, these restoration planners may opt to only focus on the collection of genetically  
 576 diverse and locally adapted planting material instead. In this case, the proposed seed zones can be used  
 577 as boundaries within which seeds of several populations may be mixed, coinciding with the regional  
 578 admixture approach proposed by Bucharova et al. (2018), which we termed ‘zonal admixture’ for  
 579 consistency with the seed zones terminology. Similarly, seed zones can be used as boundaries for  
 580 composite provenancing (Breed et al., 2013), which we termed ‘zonal composite’ provenancing, while  
 581 combining seeds from different seed zones coincides with an admixture provenancing strategy  
 582 (Broadhurst et al., 2008). Lastly, when sourcing seeds for restoration in the long term (many restoration  
 583 objectives will typically go beyond the 2050s; e.g. biodiversity conservation, slow-growing timber

584 species), an approach similar to the climate-adjusted provenancing approach (Prober et al., 2015;  
585 Ramalho et al., 2017) could be implemented by sourcing from both the seed zone predicted for the  
586 2050s and the seed zone predicted for the 2070s, if different (Figure 5). In this way, seed zones provide  
587 a useful tool to implement different previously proposed climate-smart seed sourcing strategies in a  
588 pragmatic way. It is important to note that restoration should not be considered a onetime activity, and  
589 that adaptive management may include assisted migration only at a later stage, for example when it has  
590 become more clear that local populations are not able to adapt fast enough, or when the uncertainty  
591 about future climatic changes has become lower. Other climate-smart adaptive management strategies  
592 are discussed in Lefèvre et al. (2013).



593

594 **Figure 5:** Use of dynamic seed zones for different seed sourcing strategies, in relation to the trade-off between local adaptation  
 595 vs. genetic variability, modified from Bucharova et al. (2018) (climate gradient replaced by seed zones and some seed sourcing  
 596 strategies added or modified). The sizes of the grey circles reflect the relative contributions of different seed source populations.  
 597 The ‘regional admixture’ approach proposed by Bucharova et al. (2018) is termed ‘zonal admixture’ here to fit with the seed  
 598 zone terminology. The climate-adjusted provenancing follows Ramalho et al. (2017) by making the amount of seed smaller  
 599 further along the climate change gradient and was therefore placed higher on the local adaptation axis than in Bucharova et al.  
 600 (2018). Note that with ‘local’ seed sourcing we refer to any seed source population within the local seed zone under current  
 601 climate conditions. We assume that all populations in the local seed zone are equally well adapted to local conditions; the local  
 602 seed sourcing strategy is therefore put at the same level of local adaptation as the ‘zonal admixture’ and ‘zonal composite’  
 603 seed sourcing strategies. Similarly, we assume that the ‘local + predictive’ and the ‘zonal admixture’ approaches result in a  
 604 similar level of total genetic variability (the ‘local + predictive’ approach consists of a lower number of provenances but from  
 605 more different environmental conditions). Note that the seed sourcing strategies shown here are not exhaustive and that more  
 606 variations are possible, for example the predictive provenancing approach may involve sourcing seeds from more than one  
 607 source population.

608 Once restoration practitioners have identified the appropriate seed zones, they need to identify at least  
 609 one seed source or seed provider in each of these zones. Hence, it is clear that any seed sourcing strategy  
 610 should go hand in hand with a strategy for the conservation of viable seed sources. The seed zone

611 approach we presented here can also be used as a proxy for the identification of ‘management units’ or  
612 ‘evolutionary significant units’ that should be subject of gene conservation efforts (Azpilicueta et al.,  
613 2013; Potter and Hargrove, 2012; Soliani et al., 2017). Identifying protected areas in each of the seed  
614 zones provides one way forward, but only 5% of the remaining Colombian TDFs is currently protected  
615 (i.e. less than 1% of the original TDF cover; García et al., 2014). While this underlines the pressing  
616 need to step up conservation efforts, it also indicates that protected areas alone will not be sufficient to  
617 underpin an efficient and climate-smart national seed sourcing strategy in Colombia. Experiences from  
618 the restoration of the Atlantic Forest in Brazil have shown that involving local communities and private  
619 landowners in seed collection is a promising way forward for large-scale seed collection (Brancalion et  
620 al., 2012; Schmidt et al., 2019). This can provide local communities and private landowners with  
621 alternative income sources and serve as an economic incentive for conserving local tree populations.  
622 However, it is important that seed providers are trained in proper seed collection practices (Basey et al.,  
623 2015; Thomas et al., 2017), to ensure that the planting material has a sufficient broad genetic basis  
624 promoting the evolutionary potential of established populations (Broadhurst et al., 2008; Thomas et al.,  
625 2014).

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