Effects of landscape, resource use, and body size on genetic structure in bee populations

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Abstract

Quantifying genetic structure and levels of genetic variation are fundamentally important to predicting the ability of populations to persist in human-altered landscapes and adapt to future environmental changes. Genetic structure reflects the dispersal of individuals over generations, which can be mediated by species-level traits or environmental factors. Dispersal distances are commonly positively associated with body size and negatively associated with the amount of degraded habitat between sites, motivating investigation of these potential drivers of dispersal concomitantly. We quantified genetic structure and genetic variability within populations of seven Euglossine bee species in the genus Euglossa across fragmented landscapes. We genotyped bees at thousands of SNP loci and tested the following predictions: (1) deforested areas restrict gene flow; (2) larger species have lower genetic structure; (3) species with greater resource specialization have higher genetic structure; and (4) sites surrounded by more intact habitat have higher genetic diversity. Contrasting with previous work on bees, we found no associations of body size and genetic structure. Genetic structure was higher for species with greater resource specialization, and the amount of intact habitat between or surrounding sites was positively associated with parameters reflecting gene flow and genetic diversity. These results challenge the dominant paradigm that individuals of larger species disperse farther. They suggest that landscape and resource requirements are important factors mediating dispersal, and they motivate further work into ecological drivers of gene flow for bees.

- 1 Title: Effects of landscape, resource use, and body size on genetic structure in bee populations
- 2 Running title: Landscape, body size, and dispersal in bees

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6 Abstract:

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24 **Keywords**: genetic structure, dispersal, body size, Euglossine, habitat, deforestation, bee

25 Introduction

26 As much as 75% of the global land surface has been modified by humans (Luyssaert et al., 2014). 27 One of the most concerning forms of land modification is deforestation, which typically leads to 28 fragmented landscapes that are characterized by small, isolated patches of forest surrounded by 29 agriculture or human infrastructure. Deforestation is a leading cause of biodiversity loss worldwide, due 30 to negative effects on abundance, species diversity, and genetic diversity (Schlaepfer et al., 2018). 31 Theory suggests that populations persisting in fragmented areas may experience genetic erosion 32 before changes in abundance can be detected (Pflüger et al., 2019). Therefore, quantifying the genetic 33 variability and genetic structure of populations living in fragmented areas is fundamental to 34 understanding their ability to persist in human-altered landscapes and adapt to future environmental 35 changes. Genetic structure reflects a non-random spatial distribution of genotypes, which occurs when 36 gene flow is limited across space (Wright, 1943). Gene flow occurs via dispersal and maintains genetic 37 diversity within populations (Franklin, Ian Robert, 1980). Spatially limited gene flow often results in a 38 pattern whereby populations become more genetically distinct as the distance between them increases, 39 a pattern termed "isolation by distance" (Wright, 1943). Landscape features such as water bodies or 40 mountains can also impede gene flow, a pattern called "isolation by resistance" (McRae, 2006). 41 Populations that are isolated and for which dispersal is limited may be at higher risk of extinction due to 42 loss of alleles via genetic drift, which lowers evolutionary potential (Frankel, Otto Herzberg & Soulé, 43 Michael E., 1981).

Dispersal distances may be mediated both by individual characteristics and environmental
effects (Baguette et al., 2012). Dispersal scales linearly with body size across many clades, including birds
and mammals (Dawideit et al., 2009; Ottaviani et al., 2006), moths (Beck & Kitching, 2007), plants
(Thomson et al., 2010), butterflies (Stevens et al., 2013), and bees (López-Uribe et al., 2019). However,
dispersal-body size associations often show high variability within the groups assessed, and other

species-level characteristics may also be important such as life history traits (McCoy et al., 2010; Stevens
et al., 2013), dispersal capacity (Hillman et al., 2014), diet breadth (Stevens et al., 2014), and other
resource requirements (Bowler & Benton, 2005).

52 Environmental drivers of dispersal include resource availability (Baguette, Michael et al., 2012) 53 and the extent of landscape connectivity among sites (Baguette et al., 2013). Larger organisms tend to 54 have higher resource requirements than smaller organisms, so resource availability may more strongly 55 influence dispersal propensity of larger organisms than smaller ones (Byers, 2000). In terms of landscape 56 connectivity, physical barriers to movement and habitat quality throughout the landscape can both 57 restrict dispersal (Manel & Holderegger, 2013). Negative effects of anthropogenically-altered habitat on 58 dispersal have been found for a range of species including small mammals (Ribeiro et al., 2021), birds 59 (Björklund et al., 2010), bees (Jha & Kremen, 2013) and butterflies (Crawford et al., 2011; Takami et al., 60 2004). This may be due to higher mortality for animals that travel farther in between habitat fragments 61 (Bonelli et al., 2013; Lucas et al., 1994; Mennechez et al., 2003). Other studies reveal little evidence of 62 restricted dispersal across anthropogenically-altered areas for organisms including bats (Richardson et 63 al., 2021), plants (Culley et al., 2007), and other bee species (S. S. Suni, 2017). Urban areas may even act 64 as a conduit for movement in some species (Ballare & Jha, 2021; Miles et al., 2019). Therefore, 65 understanding interplay among body size, resource requirements, and landscape in mediating dispersal 66 distances is critical given ongoing and projected anthropogenic landscape changes. 67 Bee pollinators may be particularly vulnerable to negative effects of habitat fragmentation due 68 to their haplodiploid genetic systems, which render their effective population sizes no more than 75% 69 that of equally-sized diploid populations (Whiting & Whiting, 1925). Widespread population declines 70 due to habitat loss have been reported for many bee species (LeBuhn & Vargas Luna, 2021; Potts et al., 71 2010), and these may occur via the loss of floral resources or nesting areas (Carvell et al., 2006; Cohen et

al., 2020), greater energetic costs associated with travel (Andrieu et al., 2009), or heat stress (Aguirre-

73 Gutiérrez et al., 2017; S. S. Suni & Dela Cruz, 2021). Body size and resource specialization have been 74 proposed as important traits that may mediate responses of bees to habitat loss. Larger bees are 75 potentially able to cross larger degraded areas, but they also requiring larger areas of forage to persist 76 (Harrison & Winfree, 2015). Meta analyses based on mark-recapture and genetic data suggest larger 77 bees travel farther (Greenleaf et al., 2007; López-Uribe et al., 2019), but explicit tests of how body size 78 and landscape may jointly influence dispersal in bees are lacking. Regarding resource specialization, 79 generalists are predicted to be more resistant to negative effects of habitat loss due to their ability to 80 use resources in more patches (Johnson et al., 2000). However, generalists have been found to be more 81 affected by habitat loss than specialists, but only for small bees (Bommarco et al., 2010). Taken 82 together, this past research motivates the investigation of potential intersections of landscape and 83 species-level traits on parameters that mediate bee dispersal in fragmented landscapes.

84 Here, we examined drivers of genetic structure and genetic diversity for seven species of bees in 85 the tribe Euglossini that vary widely in body size. Euglossine bees (also called Orchid Bees) are important 86 pollinators of over 700 species of orchids and other tropical plants (Roubik & Hanson, 2004). Male 87 Euglossine bees exhibit a unique behavior whereby they visit orchids and other plants to collect volatile 88 compounds that are used in sexual chemical signaling when emitted during courtship behavior (Eltz et 89 al., 2005). Euglossine bees have previously been found to show weak genetic structure over tens to 90 hundreds of kilometers (Boff et al., 2014; da Rocha Filho et al., 2013; Soro et al., 2017; Suni & 91 Hernandez, 2023; Suni, 2017; Suni et al., 2014; Suni & Brosi, 2012; Zimmermann et al., 2011). However, 92 that previous work used microsatellite loci, which may provide less insight into patterns of genetic 93 structure than a large number of SNP loci would (Gärke et al., 2012). To understand if landscape 94 characteristics and species-level traits are associated with genetic structure and diversity, we developed 95 SNP loci for each of seven species in the genus Euglossa that vary in body size. We then tested the 96 following predictions: (1) deforested areas restrict gene flow; (2) larger species have lower genetic

97 structure; (3) species with greater resource specialization have higher genetic structure; and (4) sites
98 surrounded by more intact habitat have higher genetic diversity. Our joint analysis of individual traits
99 with landscape effects on dispersal reveals patterns that contradict the dominant paradigm found for
100 bees regarding body size, and highlight the potential importance of resource specialization in influencing
101 dispersal in fragmented landscapes.

102 Materials and Methods

103 Field sampling

104 We sampled bees of seven species that range in body length from 9 mm to 15 mm (Figure 1) at 105 six sites throughout southern Costa Rica in May and June of 2019 (Figure 2, Table 1). The sites and dates 106 on which we sampled included the Las Alturas Biological Research Station (5/30/19), the Las Cruces 107 Biological Research Station (5/18/19 & 5/20/19, the La Gamba Biological Research Station (6/3/19 & 108 6/4/19), the Saladero Ecolodge (6/5/19-6/7/19), the Bromelias Ecolodge (6/2/19), and a site at the 109 northern part of the Osa Peninsula at which local landowners provided permission to sample (Agua 110 Buena; 6/1/19; see Figure 2). The species sampled vary in their resource specialization, with the number 111 of orchid morphospecies visited ranging from 6 to 20 (Roubik & Hanson, 2004; Table S1). The landscape 112 in this area is comprised of forest fragments, pastureland, palm oil plantations, and small towns. 113 Extensive deforestation occurred in the 1950s following European settlement and reduced forest cover 114 to 25% by the 1990s, but pollen and charcoal analyses from lake-sediment cores suggest continuous 115 occupation and some forest clearing by indigenous people over a 3,000-year period (Clement & Horn, 116 2001). 117 To attract bees, we used the chemical baits 1,8-cineole and methyl salicylate. These chemical 118 baits mimic the natural fragrances emitted by orchids (Janzen, 1981). We saturated cotton balls with 119 chemical baits, and used thumb tacks to attach them to tree trunks approximately 1.5 m off the ground, 120 between the hours of 9 am and 12 pm on sunny days, and in forest fragments between 0 and 93 m from forest edges. We netted bees as they arrived at baits, and we stopped sampling when no more bees arrived after 15 minutes. Bees were killed using the fumes of ethyl acetate in vials, and then transferred to vials containing 100% ethanol on the same day. Samples were then transported back to the University of San Francisco for curation and DNA extraction. Bees were pinned and then identified by examining the velvet area, a patch of dense hair on the tibial tuft, as well as other species-specific characteristics (Roubik & Hanson, 2004).

127 DNA sequencing and SNP calling

128 Genomic DNA was extracted from one or two middle legs of each specimen (two legs for the 129 smallest species) using DNeasy Blood and Tissue Extraction Kits (Qiagen). DNA concentration was 130 quantified using a Qbit 2.0 fluorometer (Thermo-Fisher) and then 100 ng of DNA per individual was used 131 to prepare ddRADseq libraries using a protocol modified from Poland et al. (2012), as follows. DNA was 132 digested with the enzymes Pstl and Mspl (New England Biolabs), and then unbarcoded adaptors that 133 were synthesized by IDT (Integrated DNA Technologies) were ligated onto the sticky ends. Ligation 134 products were then cleaned with Agencourt Ampure XP beads (Beckman Coulter) and were then used as 135 templates for PCR. PCR was performed in 96 well plates with each well containing one sample and one 136 of 285 uniquely barcoded TrueSeq primer pairs that had been synthesized by the University of California 137 San Francisco Center for Advanced Technology (UCSF CAT). An AccuBlue DNA Concentration Kit 138 (Biotium) was used to quantify DNA, and then 40 ng of each sample was pooled. Pooled DNA was 139 cleaned using Agencourt Ampure XP beads, and it was then size-selected (300-500 bp) using a Blue 140 Pippin (Sage Science). Success in obtaining accurate target fragment size distributions was confirmed 141 using a Tapestation 4200 (Agilent). The pooled, size-selected DNA was then cleaned using a Monarch 142 PCR & DNA cleanup kit (NEB) before 150-bp paired-end sequencing was performed on a NovaSeq 6000 143 (Illumina) at the UCSF CAT. To maximize sequencing coverage, we performed two NovaSeq runs, such 144 that all individuals of a given species were run on the same NovaSeq. The first run consisted of 284

145 samples belonging to Eug. imperialis, Eug. championi, and Eug. dodosni. The second run consisted of 285 146 samples belonging to Eug. flammea, Eug. maculilabris, Eug. mixta, and Eug. sapphirina, and it also 147 included additional Euglossine species of a different genus that were not included in this study. 148 We obtained demultiplexed sequences from the UCSF CAT. We assessed the quality of the 149 sequencing run using the software FastQC v.0.11.8 (Andrews, 2010), and we compared forward (R1) and 150 reverse (R2) raw fastq files for a subset of samples, checking for per base sequence quality, per-151 sequence guanine-cytosine (GC) content, and adapter content. Following the initial quality check, we 152 used the software Stacks v. 2.54 (Catchen et al. 2011, 2013) to process the sequence data. First, we 153 cleaned the raw Illumina reads using the *process* radtags program. We applied filters that discarded 154 reads for which the restriction enzyme cut-site for Mspl or Pstl was not intact, reads with Illumina 155 TruSeq adapter contamination, and reads with quality scores (Phred33) below 10 within a sliding 156 window of 15% of the read length. We then used the *denovo_map.pl* pipeline to identify orthologous 157 loci across individuals for each species separately. We performed STACKS parameter optimization for 158 each species using a small subset of individuals, following (Paris et al. 2017). We chose the following 159 parameter combination: m = 3, M = 2, n = 3 for each species, where *m* is the minimum stack depth 160 parameter that controls the number of raw reads required to form an initial stack, M is the distance 161 allowed between stacks, which represents the number of nucleotides that may be different between 162 two stacks in order to merge them, and *n* is the distance allowed among catalog loci. We also set the 163 following filtering options: --paired to assemble contigs from paired-end reads and --rm-pcr-duplicates 164 to retain a single set of paired-end reads of the same insert length. We set *max-obs-het* to 0 as in 165 Alonso-Garcia et al. (2021), to process only nucleotide sites at loci in which the maximum observed 166 heterozygosity was 0 and to remove paralogous loci. To minimize the number of retained loci that 167 would be missing in some populations, we re-ran the last step of the *denovo_map.pl* pipeline, the 168 populations program, to retain only polymorphic loci present at certain frequencies. We enabled --min169 populations so that a locus had to be present in at least two fewer the number of sampling sites, and we
170 set --min-samples-per-pop to 0.75. We limited analyses to the first SNP per locus using --write-single171 snp, and we used the --fstats option in the populations program to estimate expected heterozygosity,
172 the number of private alleles, and the percent of loci that were polymorphic for each species within
173 each site. As an additional measure of genetic diversity, we calculated allelic richness using the R
174 package Hierfstat (Goudet, 2005).

175 Landscape characterization

176 To estimate the forest percent surrounding each sampling location and between locations we 177 used ArcGIS v.2.4 (Esri, Redlands, CA). We used the Esri 2020 Land Cover dataset that corresponded to 178 scene 17P (Karra et al. 2021) to obtain forest cover of the study region. We quantified the amount of 179 forest cover within a circle of radius 24 km for each sampling location (Figure S1). We chose this radius 180 because Euglossine bees are capable of travel over tens of kilometers in a single day (Janzen, 1971), and 181 because this was the Euclidian geographic distance between the farthest edge of the Las Cruces site to 182 where we sampled at Las Alturas. Those two sites are our longest-term study sites between which we 183 have been monitoring Euglossine bee genetic structure for over 12 years. To estimate the amount of 184 forest between pairs of sampling locations we first used ArcGIS to calculate Euclidian (straight-line) 185 geographic distances between all possible site pairs. Euclidian distances are the shortest distance 186 between sites, and may traverse water. We also calculated "Broken-stick" geographic distances as in 187 Davis et al. (2010), which are the shortest overland distances between two sites. For both types of 188 distances, we overlaid rectangles of width 1000 m and calculated the amount of forest between each 189 pair of sites. We centered rectangles at each pair of sites and quantified the percent of the area that was 190 forested within that rectangle (Figure S1). Many sites are located near the coastlines of the Golfo Dulce 191 or the Pacific Ocean. We did not clip the circular or rectangular buffers to the coastline if they extended 192 into the water, so water was included as deforested area. We did this to obtain a realistic estimate of

the proportion of forest cover relative to other land cover types and to reflect possible Euglossine bee
flight paths, since some Euglossine species seem to have restricted dispersal over large bodies of water
(da Rocha Filho et al., 2013).

196 *Population and landscape genetics*

197 To determine if deforested areas restrict gene flow (prediction 1), we used Maximum Likelihood 198 of Population Effects (MLPE) mixed models to determine the effects of landscape on genetic structure 199 while taking the geographic distance between pairs of sites into account. MLPE models are emerging as 200 a powerful analytical approach in landscape genetics that permits theoretic model selection (Jha & 201 Kremen, 2013; Row et al., 2017). The MPLE approach uses pairwise individual-based genetic distances as 202 a response variable, landscape resistances and geographic distance as fixed effects, and includes a 203 random effect matrix of pairwise individual comparisons that accounts for the non-independent nature 204 of the pairwise dataset (Clarke et al., 2002). Our models included genetic distance between pairs of 205 individuals as the dependent variable, the amount of forest and geographic distance between sites as 206 independent variables, and the individuals compared as a random effect.

207 We used Hamming distance as our measure of genetic distance between individuals. Hamming 208 distance measures the dissimilarity between two strings of equal length (Hamming, 1950). It has long 209 been used in information theory and it is becoming more widely used in population genetics (Wang et 210 al., 2015). Hamming distance is especially useful when studying haploid organisms (Widhelm et al., 211 2021), such as such as the male bees we used in this study. We calculated the Hamming distance among 212 all pairs of individuals separately for each species. First, we used Stacks to output a genepop file 213 containing SNP genotypes, which we then converted into a genind object using the Adegenet package in 214 R (Jombart, 2008). Then, we used a series of custom scripts that leveraged the R packages Hierfstat, 215 tseries (Trapletti & Hornik, 2022), ResistanceGA (Peterman, 2018), and nlme (Pinheiro et al., 2017) to

calculate genetic distance and implement the MLPE models (see 'Data accessibility', below for how toaccess custom scripts).

218 To implement the MLPE approach, we ran a set of seven generalized least square (GLS) models 219 for each species separately. Code that uses generalized least squares (GLS) models to implement the 220 MLPE covariance structure is available at: <u>https://github.com/nspope/corMLPE</u>. We ranked models 221 according to their Akaike Information Criteria corrected for sample size (AICc), as in (Balbi et al., 2018). 222 We report estimates and P-values for fixed effects for models for which the difference from the model 223 with the greatest negative log likelihood was <2. Our models were as follows: a full model that included 224 Euclidian geographic and forest distances as the independent variables, a model that included only 225 Euclidian geographic distance, a model that included only forest geographic distance, a full model that 226 included broken-stick geographic and forest distances as the independent variables, a model that 227 included only broken-stick geographic distance, a model that included only broken-stick forest distance, 228 and an intercept only model. To understand if male Euglossine bees of some species disperse away from 229 their natal areas, but do not travel across the whole geographic areas sampled, we also ran a second set 230 of models for each species using datasets that included comparisons only between samples from 231 different sites (no within-site comparisons). We then evaluated if the relationship between genetic and 232 geographic distance differed between these two sets of models. We ran MLPE models for species from 233 which at least three individuals had been sampled from at least four sites (Table 1).

To determine if body size or resource generalization predict genetic structure (predictions 2 & 3), we first calculated the average genetic distance between pairs of individuals for each pair of sites, for each species. We then used this as the dependent variable in linear mixed models implemented using the lme4 package in R (Bates et al., 2014). We ran two models, one with body size as the independent variable, and one with diet breadth as the dependent variable, and we included the pair of sites between which average genetic distance was calculated as the random effect. To assess diet breadth, we compiled the number of morphospecies and genera of orchids visited for each species from records
reported in Roubik and Hanson (2004). We tested for statistical significance of the independent variable
of each model using likelihood ratio tests on nested models. In the results section we report estimates
from the best model chosen via backward model selection, and chi-square and associated P-values from
likelihood ratio tests. Table S3 shows the dataset used in this analysis.

245 To determine if sites that were surrounded by more forest had higher genetic diversity 246 (prediction 4), we ran linear mixed models implemented using the *lme4* package in R (Bates et al., 2014; 247 R Core Team, 2019). Either expected heterozygosity, the number of private alleles, or allelic richness was 248 the dependent variable. We modeled those dependent variables as a function of the forest percent 249 surrounding sites at a radius of 24 km, and we included sample size as a covariate and species as a 250 random effect. We used a dataset that included only species-site combinations that had at least four 251 individuals sampled for this analysis, and tested for significance of the independent variables using 252 likelihood ratio tests on nested models.

253 Results

254 The first sequencing run produced 467,504,244 reads (mean per sample = 1,663,716) and the 255 second run produced 679,177,300 reads (mean per sample = 2,451,904). After initial quality filtering, we 256 retained 207,471,708 reads in the first run (mean per sample = 738,333) and 508,060,286 reads in the 257 second run (mean per sample = 1,834,153). After genotyping and quality control, our final sample 258 included 493 bees that represented an average of 15 bees per species per site (Table 1). The de novo 259 assembly generated a mean of $82,670 \pm 35,080$ loci across the Euglossine bee species (Table S2). Of 260 these the mean number of polymorphic loci was $6,998 \pm 4,124$, which represented a mean of $73,656 \pm$ 261 62,300 SNPs per species. After the filtering to require that loci were present in several populations (see 262 methods), the mean number of assembled loci was 8,640 ± 7,329, and the mean number of polymorphic 263 loci was 2,994 ± 2,477 (Table S2).

264 The average genetic distance among individuals between pairs of sites ranged from 0.0017 – 265 0.18 for all species, and the average for each species across all site pairs ranged from 0.034 to 0.1. We 266 found support for prediction (1), that deforested areas restrict gene flow. For all species, there was a 267 significant negative relationship between the amount of forest between pairs of sites and the genetic 268 distance among individuals, when pairwise comparisons among bees within sites were included in MLPE 269 models (Table S4). There was variation across species in whether they exhibited isolation by distance. 270 There was a significant positive relationship between genetic and geographic distance for species with 271 the lowest resource specialization (Eug. sapphirina and Eug. flammea) but not for the more generalized 272 species (Eug. dodsoni, Eug. championi, and Eug. imperialis; Table S4). MLPE models that omitted 273 pairwise comparisons among bees within sites revealed a pattern of isolation by distance for all species 274 (Table S5).

We found no support for prediction (2), that body size predicts genetic structure. The genetic distance among pairs of individuals was not statistically associated with body size ($\chi^2 = 0.77$, P = 0.78, Table S3). However, we found support for prediction (3), that resource specialization predicts genetic structure. The number of orchid morphospecies visited was negatively related to the average genetic distance among individuals within species (Est. = -0.002, $\chi^2 = 5.0$, P = 0.025; Table S3; Figure 3).

280 Euglossine bee species varied in their genetic diversity (Table 1; Figure 4). Across species and 281 sites, means (\pm SD) were as follows: 0.19 \pm 0.057 for expected heterozygosity, 234 \pm 427 for private 282 alleles, and 1.4 ± 0.18 for allelic richness. We found support for prediction (4), that the amount of intact 283 habitat around sites positively affected genetic diversity. There was a trend towards increased expected 284 heterozygosity in sites surrounded by more forest ($X^2 = 3.0$, P = 0.084, Table 1, Figure 4a), although this 285 trend was not significant. Sites that were surrounded by more forest had more private alleles (Est. = 286 12.9, X^2 = 4.44, P = 0.035; Table 1; Figure 4b). Allelic richness did not vary with the amount of forest 287 surrounding sites ($X^2 = 1.9$, P = 0.17, Table 1).

289 Discussion

290 We present a systematic investigation of morphological and landscape drivers of genetic 291 structure for seven bee species within a clade, as well as an assessment of how genetic diversity varies 292 with the amount of intact habitat surrounding sites. We found evidence that forested landscape 293 facilitates gene flow, as genetic distances among pairs of bees were higher between sites separated by 294 less forest. We also found that genetic structure was not related to body size, but that it was related to 295 resource specialization. Bee species that were more specialized in the orchid morphospecies from which 296 they collected floral fragrances had higher genetic structure. Finally, we found evidence that the amount 297 of forested area surrounding sites was positively associated with the genetic variability of bees in those 298 sites. 299 The movement of animals can be altered in landscapes that have been fragmented (Fahrig,

300 2007). This includes the movement of flying organisms that may not be impeded by physical barriers but 301 that may still experience risks associated with travel over degraded or open areas (Caizergues et al., 302 2003; Vidal & Rendón-Salinas, 2014). For Euglossine bees, dispersal over deforested areas may be 303 influenced by the extent to which they are heat-tolerant (Roubik, 1993), as deforested areas may be 304 much hotter than intact forest (Mantyka-pringle et al., 2012). Deforested or open areas may also pose 305 greater predation risks if it compromises the ability to camouflage (Coker et al., 2009). Past work has 306 revealed restricted dispersal across water for some bee species in the genus Euglossa (Boff et. al., 2014; 307 da Rocha Filho et al., 2013). Therefore, it is not surprising that distances that traced water bodies better 308 explained genetic structure for most species, and especially for the species with the highest gene flow 309 across the landscape.

Our finding positive associations between genetic and geographic distance is somewhat
 consistent with past work. Mark-recapture observations of bees in the genus *Euglossa* have

312 documented high recapture rates over monthly time periods (T. Eltz et. al., 1999; López-Uribe et. al., 313 2008). However, other mark-recapture efforts documented male bees traveling tens of kilometers 314 within a period of days through intact forest (Pokorny et al., 2015). In addition, past population genetic 315 studies have typically found evidence of restricted dispersal for species in Euglossa only for island 316 populations (Boff et. al., 2014; da Rocha Filho et al., 2013). For populations separated by land, 317 mitochondrial COI genotyping found identical haplotypes on both sides of the Andes mountains for bees 318 in Euglossa (Dick et al., 2004). Microsatellite genotyping found low genetic structure for Eug. dilemma 319 across 130 km (Zimmermann et al., 2011), Eug. dilemma and Eug. viridissima across 114 km (Soro et al., 320 2017), Eug. imperialis across 226 km (S. S. Suni, 2017), and Eug. championi across 14 km (Suni & Brosi, 321 2012) and across 80 km (Suni et al., 2014). Our work differs from past work in that it leverages hundreds 322 to thousands of SNP loci per species to assess genetic structure. The use of more powerful markers may 323 explain our ability to detect significant isolation by distance and a signal that forest promotes dispersal. 324 This discrepancy between microsatellite and SNP-based results is consistent with past work that found 325 higher sensitivity of SNPs for detection of genetic structure using the same DNA (Zimmerman et al., 326 2020).

327 The lack of an association between body size and genetic structure contrasts with what has 328 been found previously for bees. A significant positive relationship was found between body size and 329 homing or foraging distance for 62 bee species from six families (Greenleaf et al., 2007). That study 330 compiled observational data of short-term movement patterns, and did not include estimates of 331 realized dispersal. A meta-analysis that examined associations between body size, and estimates of 332 genetic structure based on microsatellites, found an overall negative relationship between body size and 333 genetic differentiation across 42 species of bees (López-Uribe et al., 2019). Despite that negative 334 relationship overall, there was high variation in that dataset, suggesting traits other than body size are 335 also likely important drivers of genetic structure. Indeed, social species exhibited lower genetic structure than solitary species, which could be due to higher levels of kin competition for social species when
compared to solitary species (West et al., 2002). In our case, reports of nest sharing have been reported
for species within the genus *Euglossa* (Augusto & Garófalo, 2004). We therefore posit that the
avoidance of kin competition may not be a strong driver of genetic structure, although specific work
testing this hypothesis would be worthwhile.

341 Our data suggest that species that are more generalized in their resource use either disperse 342 farther or travel farther when foraging. This is consistent with some other work showing that resource 343 specialization is associated with lower gene flow. For example, species that are more generalized in their 344 resource requirements are expected to be able to disperse farther due to their ability to refuel en route 345 (Bowler & Benton, 2005). However, an empirical survey of 740 species of varying tropic levels found no 346 association between resource specialization and dispersal (Stevens et al., 2014). In addition, work 347 specifically on bees also found no evidence that genetic structure is associated with the degree of diet 348 specialization across 42 species (López-Uribe et al., 2019). Though diet specialization is commonly used 349 as a measure of niche breadth, resource requirements other than dietary requirements may also be 350 important drivers of dispersal (Bowler & Benton, 2005). Our examination of the extent of floral 351 generalization for fragrance collection revealed a positive association between the number of orchid 352 morphospecies visited and gene flow. Many tropical plants are locally rare (Wills et al., 2006), and it is 353 possible that species that are more generalized in the orchids they visit travel farther distances to 354 acquire diverse bouquets of fragrances.

355 It is worth noting that bees vary in their nesting behavior, with some species building aerial 356 nests and others using pre-existing cavities. Work on non-Euglossine bees suggests that intact habitat 357 may be particularly important for cavity nesters (Lima et al., 2020; Neame et al., 2013). However, some 358 species of cavity nesters such as carpenter bees in the genus *Xylocopa* seem to be able to thrive in urban 359 areas where human-made cavities are present (Cane et al., 2006). For Euglossine bees, past work

360 suggested that the costs of habitat destruction may be low for aerial nesters in previously deforested 361 areas, if subsequent reforestation occurs. Abundances of Euglossine bees in Brazil were found to be high 362 in secondary forest, which was attributed to there being more resin for nest construction (Becker et al., 363 1991). Regarding the species used in this study, there is variation in their nesting behavior (Table S1), 364 and no apparent associations between nesting behavior and genetic structure. For example, there is 365 variation in the nesting behavior among species that show lower genetic structure. Euglossa dodsoni 366 and Eug. championi construct aerial nests (Eberhard, 1988; Riveros et al., 2009), while Eug. imperialis 367 constructs nests in cavities that may be in the ground (Roberts & Dodson, 1967). This suggests nesting 368 behavior may not be a strong driver of genetic structure for the bees examined here, but additional 369 work on intersections between nesting behavior and deforestation on bee movement would be useful 370 to strengthen any conclusions that can be drawn.

371 There was evidence that sites that were surrounded by less forest had lower genetic diversity. 372 The susceptibility of populations to negative effects of habitat fragmentation depends on species-373 specific characteristics, such as habitat specialization and dispersal capacity (Sekar, 2012; Slade et al., 374 2013), as well as habitat availability in the surrounding area (Peakall & Lindenmayer, 2006). Species with 375 high dispersal capacity may be less likely to suffer from negative effects of fragmentation if they can 376 utilize other habitat patches. This should result in the maintenance of gene flow among patches and 377 genetic diversity within patches. Lower dispersal capacity but a network of accessible patches should 378 result in a pattern of isolation by distance, as we found in this study. Low dispersal capacity and isolated 379 fragments should lead to high genetic drift within patches and the loss of genetic diversity (Louy et al., 380 2007). With limited dispersal among fragments, genetic drift may quickly cause the loss of rare alleles in 381 small populations (Allendorf, 1986). Our finding significantly more private alleles in sites with more 382 forest suggests that drift may be lower and effective population sizes higher in fragments surrounded by 383 greater amounts of habitat. This supports other work that has documented decreases in genetic

diversity with habitat loss across diverse taxa including mammals (Lino et al., 2019), plants (González et
al., 2020), amphibians (Dixo et al., 2009), and insects (Bickel et al., 2006).

386 To our knowledge, this work is the first SNP-based assessment of genetic structure in Euglossine 387 bees, and our results highlight risks to populations associated with habitat fragmentation. In particular, 388 genetic diversity was lower in areas with less intact forest, suggesting that these bee species may be at 389 risk of further genetic erosion as habitat fragmentation continues. Indeed, a study that monitored 390 genetic diversity over time for a species used in the current study, *Eug. championi*, found striking 391 declines in genetic diversity over an 11-year period (Suni & Hernandez, 2023). Our findings reveal new 392 patterns than those found previously for Euglossine bees, which employed mitochondrial haplotypes or 393 microsatellite loci to characterize genetic structure (Boff et al., 2014; da Rocha Filho et al., 2013; (Dick et 394 al., 2004; Soro et al., 2017; Suni & Hernandez, 2023; Suni, 2017; Suni et al., 2014; Suni & Brosi, 2012; 395 Zimmermann et al., 2011). This is consistent with what has been found for bumble bees in temperate 396 areas, where investigations of dispersal distances found discrepancies between patterns emerging from 397 microsatellite versus SNP data (Lozier, 2014; Lozier et al., 2016). The inconsistency found across studies 398 employing different markers therefore motivates investigation into additional population genetic 399 studies in Euglossine bees, and investigations into the extent to which ecological specialization mediates 400 dispersal in bees more generally.

401

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740 Data accessibility

- 741 *Genetic data:* Datasets and code used in to produce statistical results and figures, as well as individual
- 742 genotype data are available at: <u>https://zenodo.org/records/10345245</u>. Individual raw sequence reads
- 743 are deposited in the SRA (BioProject ID: PRJNA880925). Sample metadata: Sample metadata, including
- 744 georeferences in decimal degrees and dates of sampling events are in Table 1.

745 Benefit-sharing

- 746 *Benefits generated*: Permission of local landowners was obtained prior to sampling. Results of scientific
- 747 enterprises are being shared with landowners, including biological research stations and ecolodges that
- promote scientific research and engage with local communities. The contributions of local individuals to
- research are described in *Methods* and *Acknowledgements*.

750 Author contributions

- 751 MH and SS designed the study, SS collected the specimens, MH curated the specimens, extracted DNA
- and performed genomic, bioinformatic, and statistical analyses with guidance from SS, and SS wrote the
- 753 manuscript with critical input from MH.

755 Tables & Figures

756

Site	Lat & Lon	MAT	MAP	Tree	Species	Body size	Orchids	Ν	H_{e}	% Poly
Agua Buena	8.694056 -83.521707	25.8	4108	67.4	Eug. sapphirina	9	6	8	0.15	0.12
Bromelias	8.685824 -83.662379	25.8	4460	44.7	Eug. sapphirina	9	6	14	0.15	0.14
La Gamba	8.702278 -83.203795	25.7	3959	61.1	Eug. sapphirina	9	6	20	0.13	0.14
Las Alturas	8.9453785 -82.833405	19.3	2997	76.0	Eug. sapphirina	9	6	8	0.15	0.11
Las Cruces	8.7875442 -82.964662	20.2	3283	64.3	Eug. sapphirina	9	6	4	0.13	0.08
Saladero	8.697707 -83.330522	25.9	4374	64.2	Eug. sapphirina	9	6	53	0.17	0.22
Agua Buena	8.694056 -83.521707	25.8	4108	67.4	Eug. dodsoni	10	14	4	0.18	0.08
Bromelias	8.685824 -83.662379	25.8	4460	44.7	Eug. dodsoni	10	14	5	0.18	0.09
La Gamba	8.702278 -83.203795	25.7	3959	61.1	Eug. dodsoni	10	14	25	0.20	0.16
Las Cruces	8.7875442 -82.964662	20.2	3283	64.3	Eug. dodsoni	10	14	7	0.21	0.13
Saladero	8.697707 -83.330522	25.9	4374	64.2	Eug. dodsoni	10	14	24	0.22	0.16
La Gamba	8.702278 -83.203795	25.7	3959	61.1	Eug. mixta	11	18	2	0.10	0.02
Las Alturas	8.9453785 -82.833405	19.3	2997	76.0	Eug. mixta	11	18	23	0.27	0.12
Las Cruces	8.7875442 -82.964662	20.2	3283	64.3	Eug. mixta	11	18	23	0.23	0.10
Saladero	8.697707 -83.330522	25.9	4374	64.2	Eug. mixta	11	18	2	0.17	0.04
Las Alturas	8.9453785 -82.833405	19.3	2997	76.0	Eug. maculilabris	12	9	32	0.28	0.09
Las Cruces	8.7875442 -82.964662	20.2	3283	64.3	Eug. maculilabris	12	9	5	0.25	0.05
Agua Buena	8.694056 -83.521707	25.8	4108	67.4	Eug. championi	13	11	6	0.15	0.12
Bromelias	8.685824 -83.662379	25.8	4460	44.7	Eug. championi	13	11	22	0.13	0.14
La Gamba	8.702278 -83.203795	25.7	3959	61.1	Eug. championi	13	11	25	0.15	0.21
Las Alturas	8.9453785 -82.833405	19.3	2997	76.0	Eug. championi	13	11	18	0.15	0.20
Las Cruces	8.7875442 -82.964662	20.2	3283	64.3	Eug. championi	13	11	26	0.14	0.22
Saladero	8.697707 -83.330522	25.9	4374	64.2	Eug. championi	13	11	24	0.15	0.23
La Gamba	8.702278 -83.203795	25.7	3959	61.1	Eug. flammea	14	8	4	0.28	0.06
Las Alturas	8.9453785 -82.833405	19.3	2997	76.0	Eug. flammea	14	8	8	0.31	0.07
Las Cruces	8.7875442 -82.964662	20.2	3283	64.3	Eug. flammea	14	8	10	0.28	0.07
Saladero	8.697707 -83.330522	25.9	4374	64.2	Eug. flammea	14	8	3	0.21	0.04
Agua Buena	8.694056 -83.521707	25.8	4108	67.4	Eug. imperialis	15	20	8	0.16	0.08
Bromelias	8.685824 -83.662379	25.8	4460	44.7	Eug. imperialis	15	20	26	0.17	0.14
La Gamba	8.702278 -83.203795	25.7	3959	61.1	Eug. imperialis	15	20	25	0.13	0.11
Las Alturas	8.9453785 -82.833405	19.3	2997	76.0	Eug. imperialis	15	20	2	0.09	0.03
Las Cruces	8.7875442 -82.964662	20.2	3283	64.3	Eug. imperialis	15	20	1	NA	NA
Saladero	8.697707 -83.330522	25.9	4374	64.2	Eug. imperialis	15	20	26	0.14	0.11

⁷⁵⁷

758 **Table 1.** For each site at which bee species in the genus *Euglossa* were sampled in southern Costa Rica,

the GPS coordinates, the mean annual temperature (MAT) in Celsius, the mean annual precipitation

760 (MAT) in mm, percent of the landscape within a circle of radius 24km that was forested, species

sampled, the body size of the species in mm, the number of specimens, and the expected heterozygosity

- 762 (H_e), and percent of loci that were polymorphic (% Poly). Sampling dates include 5/20/2019 for Las
- 763 Alturas, 5/31/2019 for Las Cruces, 6/1/2019 for Agua Buena, 6/2/19 for Bromelias, 6/3 & 6/4/2019 for
- La Gamba, and 6/6 and 6/7/2019 for Saladero. Temperature and precipitation data for each site were
- 765 obtained from <u>www.worldclim.org</u> at a spatial resolution of 2.5 minutes.



Figure 1. The seven Euglossine species sampled, along with their body sizes. From left: *Euglossa*

- imperialis (15 mm), Euglossa flammea (14 mm), Euglossa championi (13 mm), Euglossa maculilabris (12
- mm), Euglossa mixta (11 mm), Euglossa dodsoni (10 mm), and Euglossa sapphirina (9 mm).



777

Figure 1. Study area of the Osa Peninsula in southern Costa Rica. Specimens were collected **Figure 2.** Study area in Southern Costa Rica, at which seven bee species in the genus *Euglossa* were

obtained for an analysis of their genetic structure. Sites extend from costal sites on the Osa Peninsula (୫୫୮୯୪୮୬୯୫୩)^t tଡ଼ିଂଶ୍ୱେଶ୍ୱତା ୧୫୮୧୯୯୫୫୮ ଅନେ ଅନେକ୍ଷର ଅନେକ୍ଷର ଅନେକ୍ଷର ଜଣା ଅନେକ୍ଷର ଜଣା ନାର୍ଚ୍ଚ ଅନେକ୍ଷର ଜଣା ନାର୍ଚ୍ଚ ଅନେକ v. 7.3.4.8248.





788 Figure 3. For each species, genetic distance averaged across individuals within sites and then averaged 789 across sites is plotted against the number of orchid morphospecies visited by that species. Error bars 790 represent standard errors calculated from within site-averages. Colors represent different species and 791 the size of the points reflects differences in body size. 792





Figure 4. For each species, expected heterozygosity within sites (panel A) or the number of private
 alleles (panel B) is plotted against the percent of forest surrounding sites at a radius of 24 km from the
 sampling location. Colors represent different species of Euglossine bees (genus *Euglossa*) sampled from
 six sites in southern Costa Rica.



Supporting Information for Online Publication for:

Effects of landscape, resource use, and body size on genetic structure in bee populations

Melissa Hernandez & Sevan Suni

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Species	Body size	Orchids	Orchid species	Nesting
Eug. sapphirina	9	6	Houlletia odoratissima, Mormodes, Notylia barkeri, Sievekingia fimbriata, Stanhope ecornuta, Trichocentrum capistratum	Wood cavity
Eug. dodsoni	10	14	Catasetum bicolor, Coeliopis hyacinthosma, Cycnoches guttulatum, Dressleria, Gongora horichiana, G. maculata, G. quinquenervis, Kefersteinia lacteal, Kegeliella, Mormodes igneum, Notylia linearis, Notylia sp, Peristeria, Sievekingia suavis	Hard, nut shaped, on twig or branch
Eug. mixta	11	18	Catasetum bicolor, C. thompsonii, Coryanthes speciosa, C. trifoliata, Cycnoches, Dichaea panamensis, Gongora quinquenervis, Kefersteinia costaricensis, Kegeliella, Mormodes atropurpureum, M. cartonii, M. igneum, M. colossus, M. maculatum, M. powellii, Notylia, Peristeria pendula, Sievekingia fimbriata	Hollow stem or branch
Eug. maculilabris	12	9	Coryanthes, Cycnoches, Dichaea, Kefersteinia, Lacaena spectabilis, Lycaste, Mormodes, Notylia, Peristeria	Nest unknown
Eug. championi	13	10	Cycnoches, Dichaea, Dressleria dilecta, D. eburnean, D. kerryae, Mormodes atropupureun, Noylia, Peristeria, Sobralia, Stanhopea cirrhata	Dome under a leaf or in epiphyte
Eug. flammea	14	8	Catasetum maculatum, Cycnoches egertonianum, Gongora, Peristeria leucoxantha, Sievekingia fimbriata, Stanhopea cirrhata, S. oculate, S. panamensis	Ground cavity
Eug. imperialis	15	20	Catasetum macrocarpum, C. saccatum, Coryanthes trifoliata, Cycnoches egertonianum, Dichaea, Gongora maculate, G. quinquenervis, Kefersteinia, Kegeliella kupperi, Mormodes, Notylia buchtienii, Peristeria, Polycycnis muscifera, Sobralia, Stanhopea candida, S. cirrhata, S. ecornuta, S. costaricensis, Trichocentrum maculatum, Trichopilia maculata	Ground or rock cavity

Table S1. For each species, the body size, number and names of the orchid morphospecies visited, and the nesting habitat, as reported in Roubik & Hanson (2004).

Species	N	Mean depth of coverage	Retained reads	Assembled loci (pre- filtering)	Polymorphic loci (pre- filtering)	SNPs (pre- filtering)	Filtering	Assembled loci (post- filtering)	Polymorphic loci (post- filtering)
E. sapphirina	107	47.27x	296,581,295	153,924	14,229	193,697	р=4, r=0.75	10,485	7,025
E. dodsoni	65	11.26x	32,220,986	51,257	948	9,434	р=3, r=0.75	292	124
E. mixta	51	19.55x	99,403,380	92,712	8,927	82,742	р=3, r=0.75	13,138	4,296
E. maculilabris	37	14.72x	64,371,382	85,557	7,532	26,620	р=1, r=0.75	19,355	4,448
E. championi	121	10.3x	92,578,500	66,226	5,326	88,882	р=4, r=0.75	2,238	1,298
E. flammea	25	18.35x	36,453,361	52,669	4,525	28,132	p=2, r=0.75	13,423	3,141
E. imperialis	88	10.3x	81,510,644	76,344	7,496	86,086	<i>p</i> =4, <i>r</i> =0.75	1,546	626

Table S2. Summary of Stacks output generated using the *process_radtags* and *denovo_map.pl* pipelines. Stacks was run for each orchid bee species separately. For each species, the sample size (N), the mean depth of coverage, number of reads retained after cleaning the raw genomic data using *process_radtags*, and the output from *denovo_map.pl*, including the number of assembled loci, polymorphic loci, and SNPs prior to filtering, the chosen parameter values, and the number of assembled loci and polymorphic loci post filtering. Polymorphic loci were filtered using the populations program and loci were processed if they were present in at least two fewer than the number of sampled sites (*p*) and at least in 75% of individuals (*r*). Species are listed from smallest to largest body size. Post filtering, there was one SNP per polymorphic locus. The *populations* program, which is embedded in the *denovo_map.pl* pipeline, used the parameter values m-3, M=1, n=2, where m is the minimum stack depth parameter that controls the number of raw reads required to form an initial stack, M is the distance allowed between stacks, which represents the number of nucleotides that may be different between two stacks in order to merge them, and n is the distance allowed among catalog loci.

Species	Size (mm)	Orchids	Site 1	Site 2	For (E)	For (BS)	Km (E)	Km (BS)	GD
Eug. sapphirina	9	6	Agua Buena	Agua Buena	1	1	0	0	0.083
Eug. sapphirina	9	6	Agua Buena	Las Alturas	69.53	72.57	80.7	81.9	0.076
Eug. sapphirina	9	6	Agua Buena	Saladero	23.06	92.42	21	33.9	0.075
Eug. sapphirina	9	6	Bromelias	Agua Buena	94	94	15.4	15.4	0.054
Eug. sapphirina	9	6	Bromelias	Bromelias	1	1	0	0	0.049
Eug. sapphirina	9	6	Bromelias	Las Alturas	74.75	74.75	95.5	95.5	0.051
Eug. sapphirina	9	6	Bromelias	Las Cruces	71.83	89.23	77.5	78.1	0.054
Eug. sapphirina	9	6	Bromelias	Saladero	53.96	92.89	36.1	47.6	0.050
Eug. sapphirina	9	6	La Gamba	Agua Buena	53.6	94.95	34.9	39.5	0.027
Eug. sapphirina	9	6	La Gamba	Bromelias	66.65	94.54	50.4	53.2	0.026
Eug. sapphirina	9	6	La Gamba	La Gamba	1	1	0	0	0.023
Eug. sapphirina	9	6	La Gamba	Las Alturas	69.5	69.5	48.5	48.5	0.027
Eug. sapphirina	9	6	La Gamba	Las Cruces	81.66	81.66	27.7	27.7	0.028
Eug. sapphirina	9	6	La Gamba	Saladero	99.81	99.81	13.8	13.8	0.026
Eug. sapphirina	9	6	Las Alturas	Las Alturas	1	1	0	0	0.072
Eug. sapphirina	9	6	Las Alturas	Saladero	76.01	76.01	61	61	0.070
Eug. sapphirina	9	6	Las Cruces	Agua Buena	62.05	88.2	62	64.4	0.083
Eug. sapphirina	9	6	Las Cruces	Las Alturas	62.09	62.09	22.5	22.5	0.077
Eug. sapphirina	9	6	Las Cruces	Las Cruces	1	1	0	0	0.084
Eug. sapphirina	9	6	Las Cruces	Saladero	90.31	90.31	41.7	41.7	0.076
Eug. sapphirina	9	6	Saladero	La Gamba	99.81	99.81	13.8	13.8	0.024
Eug. sapphirina	9	6	Saladero	Saladero	1	1	0	0	0.069
Eug. dodsoni	10	14	Agua Buena	Agua Buena	1	1	0	0	0.069
Eug. dodsoni	10	14	Agua Buena	Saladero	23.06	92.42	21	33.9	0.048
Eug. dodsoni	10	14	Bromelias	Agua Buena	94	94	15.4	15.4	0.045
Eug. dodsoni	10	14	Bromelias	Bromelias	1	1	0	0	0.062
Eug. dodsoni	10	14	Bromelias	Las Cruces	71.83	89.23	77.5	78.1	0.064
Eug. dodsoni	10	14	La Gamba	Agua Buena	53.6	94.95	34.9	39.5	0.032
Eug. dodsoni	10	14	La Gamba	Bromelias	66.65	94.54	50.4	53.2	0.035
Eug. dodsoni	10	14	La Gamba	La Gamba	1	1	0	0	0.045
Eug. dodsoni	10	14	La Gamba	Las Cruces	81.66	81.66	27.7	27.7	0.044
Eug. dodsoni	10	14	Las Cruces	Agua Buena	62.05	88.2	62	64.4	0.061
Eug. dodsoni	10	14	Las Cruces	Las Cruces	1	1	0	0	0.097
Eug. dodsoni	10	14	Saladero	Agua Buena	23.06	92.42	21	33.9	0.045
Eug. dodsoni	10	14	Saladero	Bromelias	53.96	92.89	36.1	47.6	0.045
Eug. dodsoni	10	14	Saladero	La Gamba	99.81	99.81	13.8	13.8	0.039

Eug. dodsoni	10	14	Saladero	Las Cruces	90.31	90.31	41.7	41.7	0.060
Eug. dodsoni	10	14	Saladero	Saladero	1	1	0	0	0.056
Eug. championi	13	11	Agua Buena	Agua Buena	1	1	0	0	0.074
Eug. championi	13	11	Agua Buena	Las Alturas	69.53	72.57	80.7	81.9	0.059
Eug. championi	13	11	Bromelias	Agua Buena	94	94	15.4	15.4	0.002
Eug. championi	13	11	Bromelias	Bromelias	1	1	0	0	0.002
Eug. championi	13	11	Bromelias	Las Cruces	71.83	89.23	77.5	78.1	0.002
Eug. championi	13	11	La Gamba	Agua Buena	53.6	94.95	34.9	39.5	0.036
Eug. championi	13	11	La Gamba	Bromelias	66.65	94.54	50.4	53.2	0.002
Eug. championi	13	11	La Gamba	La Gamba	1	1	0	0	0.032
Eug. championi	13	11	La Gamba	Las Cruces	81.66	81.66	27.7	27.7	0.031
Eug. championi	13	11	La Gamba	Saladero	99.81	99.81	13.8	13.8	0.034
Eug. championi	13	11	Las Alturas	Agua Buena	69.53	72.57	80.7	81.9	0.067
Eug. championi	13	11	Las Alturas	Bromelias	74.75	74.75	95.5	95.5	0.002
Eug. championi	13	11	Las Alturas	La Gamba	69.5	69.5	48.5	48.5	0.032
Eug. championi	13	11	Las Alturas	Las Alturas	1	1	0	0	0.060
Eug. championi	13	11	Las Alturas	Las Cruces	62.09	62.09	22.5	22.5	0.056
Eug. championi	13	11	Las Alturas	Saladero	76.01	76.01	61	61	0.059
Eug. championi	13	11	Las Cruces	Agua Buena	62.05	88.2	62	64.4	0.064
Eug. championi	13	11	Las Cruces	Las Cruces	1	1	0	0	0.055
Eug. championi	13	11	Saladero	Agua Buena	23.06	92.42	21	33.9	0.067
Eug. championi	13	11	Saladero	Bromelias	53.96	92.89	36.1	47.6	0.002
Eug. championi	13	11	Saladero	Las Cruces	90.31	90.31	41.7	41.7	0.056
Eug. championi	13	11	Saladero	Saladero	1	1	0	0	0.061
Eug. flammea	14	8	La Gamba	La Gamba	1	1	0	0	0.167
Eug. flammea	14	8	La Gamba	Las Alturas	69.5	69.5	48.5	48.5	0.175
Eug. flammea	14	8	Las Alturas	La Gamba	69.5	69.5	48.5	48.5	0.163
Eug. flammea	14	8	Las Alturas	Las Alturas	1	1	0	0	0.163
Eug. flammea	14	8	Las Alturas	Las Cruces	62.09	62.09	22.5	22.5	0.054
Eug. flammea	14	8	Las Cruces	La Gamba	81.66	81.66	27.7	27.7	0.055
Eug. flammea	14	8	Las Cruces	Las Cruces	1	1	0	0	0.045
Eug. flammea	14	8	Las Cruces	Saladero	90.31	90.31	41.7	41.7	0.013
Eug. imperialis	15	20	Agua Buena	Agua Buena	1	1	0	0	0.064
Eug. imperialis	15	20	Bromelias	Agua Buena	94	94	15.4	15.4	0.057
Eug. imperialis	15	20	Bromelias	Bromelias	1	1	0	0	0.054
Eug. imperialis	15	20	La Gamba	Agua Buena	53.6	94.95	34.9	39.5	0.025
Eug. imperialis	15	20	La Gamba	Bromelias	66.65	94.54	50.4	53.2	0.020

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Euro importatio	15	20	La Camba	la Camba	1	1	0	0	
Eug. Imperialis Eua. imperialis	15 15	20 20	La Gamba La Gamba	La Gamba Saladero	1 99.81	1 99.81	U 13.8	0 13.8	
Eug. imperialis	15	20	Saladero	Agua Buena	23.06	92.42	21	33.9	
Eug. imperialis	15	20	Saladero	Bromelias	53.96	92.89	36.1	47.6	
Eug. imperialis	15	20	Saladero	Saladero	1	1	0	0	

Table S3. For each species, its body size, the number of orchid morphospecies visited, the site pairs between which genetic distances were calculated (Site 1 & Site 2), the percent of the distance between them that was forested when calculated using Euclidian paths (For. (E)), the percent of the distance between them that was forested when calculated using Broken-stick paths (For. (BS)), the Euclidian geographic distance between them (Km (E)), the Broken-stick geographic distance between them (Km (BS)), and the average genetic distance among individuals between those pairs (Hamming genetic distance).

Species	Model	AIC	AICc	ΔAICc	Model Summary
	Full (E)	-39633.7	-39633.69	0	
	Km (E)	-37599.99	-37599.99	-2033.7	
	For (E)	-39574.77	-39574.77	-58.92	Km(E) = 0.000058
Eug. sapphirina	Intercept	-37355.1	-37355.09	-2278.6	101 (L) = -0.00017
	Full (BS)	-39365.78	-39365.77	-267.92	<i>t</i> = 7.82, <i>P</i> < 0.001 <i>t</i> = -49.5, <i>P</i> < 0.001
	Km (BS)	-37584.11	-37584.1	-2049.59	
	For (BS)	-39077.91	-39077.9	-555.79	
	Full (E)	-11777.71	-11777.68	-27.88	
	Km (E)	-11639.4	-11639.38	-166.18	
	For (E)	-11761.27	-11761.25	-44.31	Km (BS) = -0.000067
Eug. dodsoni	Intercept	-11436.86	-11436.84	-368.72	101 (83)0.00012
	Full (BS)	-11805.59	-11805.56	0	<i>t</i> = -2.12, <i>P</i> = 0.036 <i>t</i> = -13.0, <i>P</i> < 0.001
	Km (BS)	-11645.41	-11645.39	-160.17	
	For (BS)	-11803.22	-11803.2	-2.36	
	Full (E)	-47630.4	-47595.95	-47630.4	
	Km (E)	-46906.2	-46878.64	-46906.2	
	For (E)	-47417.6	-47390.04	-47417.6	Km (BS) = -0.000064 For (BS) = -0.00013
Eug. championi	Intercept	-45447.31	-45426.64	-45447.31	
	Full (BS)	-48031.57	-47997.12	-48031.56	t = -11.1, P < 0.001 t = -31.3, P < 0.001
	Km (BS)	-47116.26	-47088.7	-47116.25	
	TreeBS	-47912.54	-47884.98	-47912.53	
	Full (E)	-1438.61	-1438.406	0	
	Km (E)	-1256.073	-1255.937	-182.469	
	For (E)	-1406.346	-1406.21	-32.196	Km (E) = 0.00057 For (E) = -0.00085
Eug. flammea	Intercept	-1223.928	-1223.847	-214.559	
	Full (BS)	-1438.61	-1438.406	0	<i>t</i> = 6.0, <i>P</i> < 0.001 <i>t</i> = -16.1, <i>P</i> < 0.001
	Km (BS)	-1256.073	-1255.937	-182.469	
	TreeBS	-1406.346	-1406.21	-32.196	

Eug. imperialis	Full (E)	-26737.65	-26737.63	-338.3	
	Km (E)	-26728.77	-26728.76	-347.17	
	For (E)	-25295.59	-25295.58	-1780.35	Km (BS) = -0.00021
	Intercept	-24822.72	-24822.71	-2253.22	Km (BS) = -0.000049
	Full (BS)	-27075.95	-27075.93	0	<i>t</i> = -26.3, <i>P</i> < 0.001
	Km (BS)	-26936.14	-26936.13	-139.8	<i>t</i> = -12.0, <i>P</i> < 0.001
	For (BS)	-26444.21	-26444.19	-631.74	

Table S4. Results from Maximum Likelihood of Population Effects (MLPE) models assessing the joint effects of the amount of land that was forested and geographic distance among site pairs on genetic distance among pairs of individuals. For each species, seven models were compared, a full model (Full (E)) that included as fixed effects the Euclidian geographic distance and amount of land that was forested along that path among site pairs, a model that included only Euclidian geographic distance (Geo (E)), a model that included only the amount of land that was forested (For (E)), a full model (Full (BS)) that included as fixed effects the Broken-stick geographic distance and amount of land that was forested along that path among site pairs, a model that included only Broken-stick geographic distance (Geo (BS)), a model that included only the amount of land that was forested (For (E)), and an intercept-only model (Intercept). Columns 3-6 show AIC and sample-size corrected AIC (AICc) values, the difference in AICc from the best model, and model results, including estimates for fixed effects and associated t and P-values.

Species	Model	AIC	AICc	ΔAICc	Model Summary			
	Full (E)	-27921.08	-27921.07	0				
	Km (E)	-27827.86	-27827.85	-93.22	Km(E) = 0.00012			
	For (E)	-27785.01	-27785	-136.07	For (E) = -0.000065			
Eug. sapphirina	Intercept	-27645.89	-27645.89	-275.18	t = 11 9 P < 0 001			
	Full (BS)	-27826.33	-27826.32	-94.75	t = -9.82, P < 0.001			
	Km (BS)	-27828.12	-27828.11	-92.96				
	TreeBS	-27644.9	-27644.89	-276.18				
	Full (E)	-8385.668	-8385.625	0				
	Km (E)	-8370.469	-8370.441	-15.184	Km(E) = 0.00016			
	For (E)	-8371.935	-8371.907	-13.718	For (E) = 0.00010			
Eug. dodsoni	Intercept	-8357.691	-8357.674	-27.951	t = 4.0 P < 0.001			
	Full (BS)	-8383.545	-8383.502	-2.123	t = 4.2, P < 0.001			
	Km (BS)	-8361.176	-8361.147	-24.478				
	TreeBS	-8385.064	-8385.036	-0.589				
	Full (E)	-40852.16	-40852.15	0				
	Km (E)	-40589.04	-40589.04	-263.11	Km(E) = 0.000068			
	For (E)	-40728.96	-40728.95	-123.2	For (E) = -0.00012			
Eug. championi	Intercept	-40529.75	-40529.74	-322.41	t = 11.3 P < 0.001			
	Full (BS)	-40617.49	-40617.48	-234.67	t = -16.5, P < 0.001			
	Km (BS)	-40593.78	-40593.77	-258.38				
	TreeBS	-40588.87	-40588.86	-263.29				
	Full (E)	-1266.913	-1266.63	0	Km (E) = 0.0032			
	Km (E)	-1000.5586	-1000.3709	-266.2591	For (E) = 0.04			
	For (E)	-950.5323	-950.3445	-316.2855	<i>t</i> = 40.4, <i>P</i> < 0.001			
Eug. flammea	Intercept	-949.1541	-949.0419	-317.5881	<i>t</i> = -35.2, <i>P</i> < 0.001			
	Full (BS)*	-1266.913	-1266.63	0	*No paths went over water for this species, so			
	Km (BS)	-1000.5586	-1000.3709	-266.2591	Full (E) and Full (BS) are identical			
	TreeBS	-950.5323	-950.3445	-316.2855				

	Full (E)	-20769.57	-20769.55	0	
	Km (E)	-20074.27	-20074.26	-695.29	Km(E) = 0.000075
Fug	For (E)	-20752.27	-20752.26	-17.29	For (E) = 0.00024
imperialis	Intercept	-19647.13	-19647.12	-1122.43	t = 4.4 P < 0.001
	Full (BS)	-20754.82	-20754.79	-14.76	<i>t</i> = 28.1, <i>P</i> < 0.001
	Km (BS)	-20291.12	-20291.1	-478.45	
	For (BS)	-20455.06	-20455.04	-314.51	

Table S5. Results from Maximum Likelihood of Population Effects (MLPE) models assessing the joint effects of the amount of land that was forested and geographic distance among site pairs on genetic distance among pairs of individuals, using a dataset that included only genetic distances estimated from individuals at different sites. For each species, seven models were compared, a full model (Full (E)) that included as fixed effects the Euclidian geographic distance and amount of land that was forested along that path among site pairs, a model that included only Euclidian geographic distance (Geo (E)), a model that included only the amount of land that was forested (For (E)), a full model (Full (BS)) that included as fixed effects the Broken-stick geographic distance and amount of land that was forested along that path among site pairs, a model that included only Broken-stick geographic distance (Geo (BS)), a model that included only the amount of land that was forested (For (E)), and an intercept-only model (Intercept). Columns 3-6 show AIC and sample-size corrected AIC (AICc) values, the difference in AICc from the best model, and model results, including estimates for fixed effects and associated t and P-values.



Figure 5. A circular buffer with a radius of 24 km was created to estimate the percent of forest **Figure S1.** *Left panel:* The percent forest within a circle of radius 24 km was calculated using GIS. Light green = pastureland; dark green = forest; blue = water. *Right panel:* The percent forest between pairs of sites was calculated using both Euclidian (yellow) and Broken-stick (red) paths.