

1 **Balancing selection at a wing pattern locus is associated with major shifts in genome-wide**
2 **patterns of diversity and gene flow**

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27 **Abstract:** Selection shapes genetic diversity around target mutations, yet little is known about how
28 selection on specific loci affects the genetic trajectories of populations, including their genome-
29 wide patterns of diversity and demographic responses. Here we study the patterns of genetic
30 variation and geographic structure in a neotropical butterfly, *Heliconius numata*, and its closely
31 related allies in the so-called melpomene-silvaniform clade. *H. numata* is known to have evolved an
32 inversion supergene which controls variation in wing patterns involved in mimicry associations
33 with distinct groups of co-mimics **whereas it is associated to** disassortative mate preferences and
34 heterozygote advantage at this locus. We contrasted patterns of genetic diversity and structure 1)
35 among extant polymorphic and monomorphic populations of *H. numata*, 2) between *H. numata* and
36 its close relatives, and 3) between ancestral lineages. We show that *H. numata* populations which
37 carry the inversions as a balanced polymorphism show markedly distinct patterns of diversity
38 compared to all other taxa. They show the highest genetic diversity and effective population size
39 estimates in the entire clade, as well as a low level of geographic structure and isolation by distance
40 across the entire Amazon basin. By contrast, monomorphic populations of *H. numata* as well as its
41 sister species and their ancestral lineages all show lower effective population sizes and genetic
42 diversity, and higher levels of geographical structure across the continent. One hypothesis is that the
43 large effective population size of polymorphic populations could be caused by the shift to a regime
44 of balancing selection due to the genetic load and disassortative preferences associated with
45 inversions. Testing this hypothesis with forward simulations supported the observation of increased
46 diversity in populations with the supergene. Our results are consistent with the hypothesis that the
47 formation of the supergene triggered a change in gene flow, causing a general increase in genetic
48 diversity and the homogenisation of genomes at the continental scale.

49

50 **Introduction:** Genetic diversity is shaped by selective processes such as stabilizing or disruptive
51 selection, and by demographic processes such as fluctuations in effective population size. Empirical
52 studies on genetic diversity within and among populations abound, fuelled by an increasing
53 availability of whole genome data, and spurred by our interest in understanding the underlying
54 causes of variation in diversity (e.g. Beichmann 2018, Muers 2009; Murray 2017; Nielsen et al.
55 2009). At the locus scale, strong directional or disruptive selection tends to reduce diversity within
56 populations (Mitchell-Olds et al. 2007), while balancing selection tends to enhance diversity
57 (Charlesworth 2006). Genome-wide factors reducing diversity include low effective population
58 sizes, generating drift, while high genetic diversity is enhanced by large population sizes and gene
59 flow. Overall, it is well recognised that demographic changes should have a genome-wide effect on
60 diversity, while positive selection is expected to play a role on the sites within and around the genes
61 involved in trait variation (Glinka et al. 2003, Muers 2009, Nielsen et al. 2009).

62

63 Variation in behaviour and life-history traits, for instance involving changes in offspring viability or
64 dispersal distance, may also affect species demography, and thus whole genome genetic diversity.
65 However, whether and how genetic variability in a population may be driven by phenotypic
66 evolution at certain traits is poorly understood, and confounding effects may affect patterns of
67 genomic diversity, such as variation in census population size or colonization history. Dissecting
68 how selection on a trait may affect genome-wide diversity can be tackled by comparing closely-
69 related populations differing at this trait coupled with knowledge of when the differences evolved.
70 Here, we took advantage of the dated introgressive origin of a chromosomal inversion associated
71 with major life-history variation to study the demographics and whole genome consequences of
72 changes in the selection regime at a major-effect locus.

73 *Heliconius* butterflies are aposematic, chemically-defended butterflies distributed over the
 74 American tropics from Southern Brazil to Southern USA (Emsley 1965; Brown 1979) (Fig 1A).
 75 *Heliconius* butterflies are well-known for visual resemblance among coexisting species, a
 76 relationship called Müllerian mimicry which confers increased protection from bird predators
 77 through the evolution of similar warning signals (Sheppard et al. 1985). Most species are locally
 78 monomorphic, but their mimicry associations vary among regions, and most species display a
 79 geographic mosaic of distinct mimetic “races” through their range. In contrast to most *Heliconius*
 80 species, the tiger-patterned *Heliconius numata* is well-known for maintaining both mimicry
 81 polymorphism within localities, with up to seven differentiated coexisting forms, and extensive
 82 geographic variation in the distribution of wing phenotypes (Brown & Benson 1974; Joron et al.
 83 1999). Forms of *H. numata* combine multiple wing characters conveying resemblance to distinct
 84 sympatric species in the genus *Melinaea* and other local Ithomiini species (Nymphalidae:
 85 Danainae). Polymorphism in *H. numata* is controlled by a supergene, i.e. a group of multiple linked
 86 functional loci segregating together as a single Mendelian locus, coordinating the variation of
 87 distinct elements of phenotype (Brown & Benson 1974; Joron et al. 2006). Supergene alleles are
 88 characterized by rearrangements of the ancestral chromosomal structure, forming three distinct
 89 chromosomal forms with zero (ancestral type, Hn0), one (Hn1) or three chromosomal
 90 rearrangements (Hn123) (Fig 1B). The ancestral arrangement, Hn0, devoid of inversions, is fixed in
 91 most *Heliconius* species (although an inversion in the same region evolved independently in a
 92 distantly-related *Heliconius* lineage (Edelman et al. 2019)). Arrangement Hn1 contains a 400kb
 93 inversion called P₁ originating from an introgression event about 2.2 My ago from *H. pardalinus*, in
 94 which P₁ is fixed (Jay et al. 2018). This introgression is thought to be the founding event triggering
 95 the formation of the supergene and the maintenance of polymorphism in *H. numata* (Jay et al.
 96 2018). Arrangement Hn123 displays two additional inversions, P₂ and P₃, in linkage with P₁, and
 97 therefore originated after the introgression of P₁ into the *H. numata* lineage (Jay et al. 2021).

98
 99 *Heliconius numata* is widespread in the lowland and foothill tropical forests of the Amazon basin,
 100 the Guianas, and the Brazilian Atlantic Forest (Mata Atlântica), but the frequencies of the three
 101 chromosome arrangements vary across the range. Ancestral type Hn0 is fixed in the Atlantic Forest
 102 populations of Brazil (forms *robigus* or *ethra*), but segregates at intermediate frequencies in all
 103 other *H. numata* populations throughout the range (forms *silvana* and *laura*) (Fig 1C). Chromosome
 104 type Hn1 is associated with the Andean mimetic form *bicoloratus* and is found in the Eastern
 105 Andean foothills of Ecuador, Peru, and Bolivia. Chromosome type Hn123 is associated with a large
 106 diversity of wing-pattern forms of intermediate allelic dominance, including *tarapotensis*, *arcuella*
 107 and *aurora*, and is reported from Andean, lowland Amazonian and Guianese populations. Inversion
 108 polymorphism is therefore structured across the range, with populations being fixed for the
 109 ancestral chromosome (Atlantic Forest, see Text S1 & Table S1-2), or displaying a polymorphism
 110 with two (Amazon-Guiana) or three (Andes) chromosomal types in coexistence (Joron et al. 2011).
 111 Monomorphic populations of the Atlantic Forest, devoid of rearrangements at the supergene locus,
 112 might represent the ancestral state displayed by *H. numata* populations before the evolution of the
 113 supergene via introgression (Fig 1C).

114
 115 The wing patterns of *H. numata* are subject to selection on their resemblance to local co-mimics
 116 (Chouteau et al. 2016), but the polymorphism is maintained by balancing selection on the
 117 chromosome types. Balancing selection is indeed mediated by disassortative mating favouring
 118 mixed-form mating (Chouteau et al. 2017) and is likely to have evolved in response to the

119 deleterious mutational load carried by inversions, which causes heterozygous advantage in *H.*
 120 *numata* (Jay et al. 2021, Faria et al. 2019, Maisonneuve et al. 2019). The introgression of P₁ and the
 121 formation of a supergene were associated with a major shift in the selection regime (Jay et al.
 122 2018). The mating system also changed during or after introgression. These events may therefore
 123 have profoundly affected the population biology of the recipient species, *H. numata*. We investigate
 124 here whether the adaptive introgression of a balanced inversion is associated with a signature in
 125 genetic diversity and geographic structure. In particular, we predict that genetic diversity should be
 126 higher in *H. numata* than in closely related taxa. Similarly, nucleotide diversity should be higher in
 127 all polymorphic populations carrying either one segment (Hn1) or two (Hn1,Hn123) compared to
 128 the population that is monomorphic and only carries the non-inverted segment (Hn0) in the
 129 Brazilian Atlantic Forest. We analyse changes in the demographic history of the clade containing *H.*
 130 *numata* and closely related taxa, as well as their current patterns of diversity and demography, using
 131 three well separated populations of *H. numata* representing different states of inversion
 132 polymorphism. Our results suggest that following supergene formation, a change in the selection
 133 regime and mating system may have facilitated gene flow among morphs and had key consequences
 134 in current patterns of genetic structure.

135

136 **Material and Methods**

137 We used here whole genome resequencing from 137 specimens of *Heliconius*, including 68 *H.*
 138 *numata*. Sampling included specimens from populations in the Andean foothills (3 chromosome
 139 types), from the upper Amazon (2 chromosome types), from French Guiana (2 chromosome types)
 140 and from the Brazilian Atlantic Forests (1 chromosome type) (Fig 1C; Table S3). Related taxa were
 141 represented by the sister species *H. ismenius*, found west of the Andes (parapatric to *H. numata*), by
 142 Amazonian representatives of the lineage *H. pardalinus* (donor of the inversion), *H. elevatus*, *H.*
 143 *ethilla*, *H. besckei* as well as *H. hecale*, and by *H. melpomene* and *H. cydno* as outgroups. Only
 144 Andean, Amazonian and Guianese populations of *H. numata* display chromosomal polymorphism,
 145 all other taxa being fixed for the standard gene arrangement (Hn0), or for the inverted arrangement
 146 Hn1 (*H. pardalinus*) (Jay et al. 2018). Hereafter, *H. numata* populations from the Andes, Amazon
 147 and French Guiana will be collectively referred to as “Amazonian”, and populations from the
 148 Atlantic Forest as “Atlantic”. Butterfly bodies were preserved in NaCl saturated DMSO solution at
 149 20°C and DNA was extracted using QIAGEN DNeasy blood and tissue kits according to the
 150 manufacturer’s instructions with RNase treatment. Illumina Truseq paired-end whole genome
 151 libraries were prepared and 2x100bp reads were sequenced on the Illumina HiSeq 2000 platform.

152 Reads were mapped to the *H. melpomene* Hmel2 reference genome (Davey et al., 2016) using
 153 Stampy (version 1.0.28; Lunter and Goodson, 2011) with default settings except for the substitution
 154 rate, which was set to 0.05 to allow for the expected divergence from the reference of individuals in
 155 the so-called silvaniform clade (*H. numata*, *H. pardalinus*, *H. elevatus*, *H. hecale*, *H. ismenius*, *H.*
 156 *besckei* and *H. ethilla*). For *H. melpomene* and *H. cydno* belonging to the so-called *melpomene*
 157 clade, their genomes were mapped with a substitution rate of 0.02. Alignment file manipulations
 158 were performed using SAMtools v0.1.3 (Li et al. 2009). After mapping, duplicate reads were
 159 excluded using the *MarkDuplicates* tool in Picard (v1.1125; <http://broadinstitute.github.io/picard>)
 160 and local indel realignment using IndelRealigner was performed with GATK (v3.5; DePristo et al.
 161 2011). Invariant and polymorphic sites were called with GATK HaplotypeCaller, with options --
 162 min_base_quality_score 25 --min_mapping_quality_score 25 -stand_emit_conf 20 --heterozygosity
 163 0.015.

164 F_{ST} , d_{XY} and π were calculated in overlapping windows of 25 kb based on linkage disequilibrium

165 decay (*Heliconius* Genome Consortium 2012) using custom scripts provided by Simon H. Martin
 166 (<https://github.com/simonhmartin>), and the genome-wide average was calculated using our own
 167 scripts (available from <https://github.com/angelesdecarra>). Distance in km between sampling sites
 168 was measured along a straight line, not taking into account potential physical barriers. Following
 169 Rousset (1997), in a 2-dimensional habitat, under a model of isolation by distance (IBD)
 170 differentiation, measured as $F_{ST}/(1-F_{ST})$, should increase as a function of the logarithm of distance.
 171 Therefore, we tested for the existence and intensity of an IBD signal among species and between
 172 populations of *H. numata* using a linear model. If IBD is stronger in species not polymorphic for the
 173 inversion we should observed significantly steeper slopes in these species. To test this, we measured
 174 IBD (i) within populations of each species separately, (ii) for all *H. numata* within the Amazonian
 175 region (excluding Atlantic forest populations) and (iii) for all *H. numata* including the Atlantic
 176 region. The slopes of $F_{ST}/(1-F_{ST})$ versus $\log(\text{distance})$ were calculated using the R package *lsmeans*
 177 (Lenth 2016); the slope difference among species or between populations within species was
 178 estimated with an ANOVA and its significance evaluated with function pairs of this package (Text
 179 S1 and see example script on github.com/angelesdecarra).
 180

181 Admixture (Alexander et al. 2009) analyses were run on a subset of the 68 *H. numata* genomes,
 182 keeping only 15 individuals from Peru to have a more balanced representation of individuals across
 183 the geographic distribution. Filters were applied to keep biallelic sites with minimum mean depth of
 184 8, maximum mean depth of 200 and at most 50% genotypes missing. We only kept 1 SNP per
 185 kilobase to remove linked variants using the thinning function in *vcftools*, and we obtained the
 186 optimal number of clusters using cross-validation for values of K from 1 to 10 (Alexander et Lange,
 187 2011). Principal component analyses (PCA) were performed with the same filters as for admixture,
 188 using the same *H. numata* genomes as for the admixture analyses, using *plink2* (Chang et al. 2015).
 189

190 In order to estimate demographic parameters independently of the effect of selection on diversity,
 191 we performed stringent filtering on the dataset. We removed all predicted genes and their 10,000
 192 base-pair flanking regions, before performing G-PhoCS (Gronau et al. 2011) analyses as detailed
 193 below. Repetitive regions were masked using RepeatMasker and Tandem Repeat Finder (Benson
 194 1999). GC islands detected with CpGcluster.pl with parameters 50 and 1E-5 (Hackenberg et al.,
 195 2006) were also masked. Scaffolds carrying the supergene rearrangements (Hmel215006 to
 196 Hmel215028) were excluded, as were scaffolds from the sex chromosome (Z) and mtDNA, since
 197 those are expected to show unusual patterns of diversity due to selection and different effective
 198 population sizes.
 199

200 We analysed the demographic history of *H. cydno*, *H. numata*, *H. ismenius*, *H. pardalinus* and *H.*
 201 *elevatus* with G-PhoCS, which allows for the joint inference of divergence time, effective
 202 population sizes and gene flow. In order to detect differences in demography correlating with the
 203 presence of the supergene in *H. numata*, we conducted analyses separating the Atlantic population
 204 of *H. numata* from Amazonian populations. G-PhoCS is an inference method based on a full
 205 coalescent isolation-with-migration model. Inferences are conditioned on a given population
 206 phylogeny (based on Kozak et al. 2015) with migration bands (i.e. priors in the migration rates) that
 207 describe allowed scenarios of post-divergence gene flow. The model assumes distinct migration rate
 208 parameters associated with each pair of populations, and allows for asymmetric gene flow. Given
 209 the computational burden of G-PhoCS, we selected two individuals per taxon or population,
 210 retaining those with the highest sequencing depth (see Table S3), taking into account their relative

211 **abundance**. The input dataset consisted of 4092 genomic regions, each 1kb in length and spaced at
 212 approximately 30kb intervals (above the value at which LD decays at more than half of its value)
 213 and with genotypes in at least one of the two samples of each taxon We used as priors for
 214 coalescence times (τ) and genetic diversity (θ), Gamma functions with $\alpha=1$ and $\beta=100$, and for
 215 migration rates $\alpha=0.002$ and $\beta=0.00001$. These priors were chosen to allow good convergence while
 216 also ensuring non informativity. In order to calculate the highest posterior density interval, we used
 217 the library HDInterval in R, and to integrate such posterior densities we used the library sfsmisc in
 218 R. We rescaled the results using a mutation rate of $1.9E-9$ (Martin et al. 2016) and 4 generations per
 219 year (i.e., $g=0.25$). Migration bands were considered significant following the criteria of Freedman
 220 et al. (2012): if the 95% HPD interval did not include 0 or if the total migration rate (i.e. migration
 221 rate times the duration of the migration band) was larger than 0.03 with posterior probability larger
 222 than 0.5.

223 **Demographic Reconstruction of population size changes, split and mixtures**

224 G-phocs provides useful information across all species but i) does not allow to quantify the time
 225 scale of population size change, ii) is limited in the number of individuals it can handle and iii)
 226 displayed limited accuracy to distinguish N_e and m in a simulation study (Gronau et al. 2011). We
 227 thus constructed additional models to test the hypothesis that *H. numata* populations with inversion
 228 polymorphism display an increased effective population size due to disassortative mating. To test
 229 this, we used $\partial a \partial i$ to reconstruct the demographic history of *H. numata* individuals from the
 230 Amazonian forest, quantify their historical changes in effective population size and test their
 231 divergence history from 1) *H. numata* from the Brazilian area, which do not carry the inversion; and
 232 2) *H. pardalinus* individuals. We allowed for change in effective population size in both the
 233 ancestral populations. The change in effective population size in *H. numata* associated with the
 234 change in mating system should be more recent than the time of introgression of the inversion into
 235 *H. numata*. To verify this hypothesis, we allowed for change in N_e of the daughter population at any
 236 time after the split. We tested different models of divergence with and without (asymmetric) migration
 237 and included the effect of linked selection (e.g. Roux et al. 2016).

238 Since the conditions of historical divergence are not known, we tested a model of divergence with
 239 ongoing migration (IM) a model of divergence with ancient migration if gene-flow has stopped
 240 recently (AM) and, in the case of divergence into multiple refugia, a model of secondary contact
 241 (SC). We also included a model of Strict Isolation (SI) as a null model.

242 The models shared the following parameters: the ancestral populations of size N_{anc} can grow or
 243 shrink to a size N_{anc2} between T_{anc} and up until its splits at time T_{split} into two daughter populations of
 244 size N_1 and N_2 . Under the SI model, no gene flow occurs between the two populations. Under AM,
 245 gene flow occurred between T_{split} and T_{am} and is followed by a period of strict isolation. Under IM,
 246 gene flow occurs at a constant rate at each generation between the two populations. Gene flow can
 247 be asymmetric, so that two independent migration rates m_{12} (from population 2 to 1) and m_{21} (from
 248 population 1 to 2) were modeled. Under the SC model, the population evolved in strict isolation
 249 between T_{split} and until T_{sc} where a secondary contact occurs continuously up to present time. Gene
 250 flow is modeled as $M = 2N_{REF}.m$. In $\partial a \partial i$, heterogeneity in effective population size was used to
 251 account for linked selection by defining two categories of loci with varying effective population
 252 sizes (proportion $1-Q$ of loci with a “neutral N_e ” and a proportion Q of loci with a reduced effective
 253 population size due to either selection at linked site). To quantify how linked selection affects
 254 reduced N_e , we used a Hill-Robertson scaling factor (Hrf) to relate the effective population size of
 255 loci influenced by selection ($N_r = Hrf * N_e$) to that of neutral loci (N_e). A hierarchical approach was
 256

257 used to avoid over-fitting: first, we compared models assuming constant effective population size.
 258 Second, the best identified models were modified to incorporate population expansion or decline, as
 259 expected given the observed distribution of genetic diversity. Population expansion was
 260 implemented using two additional parameters for population 1 and population 2, allowing each
 261 population to either grow or decline exponentially at any time after their split from the ancestral
 262 population (controlled by parameters s_1 and s_2 for population 1 and 2 respectively) **which defined**
 263 **the new time parameters (Tp1 and Tp2), which indicate the time of exponential change).**
 264 Models were fitted using the diffusion theory implemented in $\partial a \partial i$ (Gutenkunst et al. 2009) and
 265 includes the effect of linked selection. $\partial a \partial i$ uses the SFS as a summary of the data. For a given
 266 demographic model, the SFS is computed using diffusion approximation and compared to the
 267 empirical SFS using AIC.
 268 We used stringent filtering ($GQ > 30$, $4 < \text{mean depth} < 80$) and no missing data to keep high quality
 269 sites and remove potential paralogs or PCR duplicates exhibiting excessive read depth. To minimize
 270 linkage we subset our data to keep one SNP every 5kb. No MAF filter was used and singletons
 271 were kept to avoid ascertainment bias in estimates of demographic parameters. For each model, 32
 272 independent replicate runs were performed and **for each model, the run** with the lowest AIC and
 273 ΔAIC was kept.

274

275 **Forward Simulations**

276 In order to better understand the nature of the processes that generate higher genetic diversity in *H.*
 277 *numata* compared to closely related taxa, we used simulations to test the hypothesis that
 278 disassortative mating generates an increase in levels of genetic diversity at a genome-wide scale.
 279 We hypothesized that such level of genetic diversity is higher than expected under i) **random mating**
 280 **(a model similar to panmixia)** or ii) **assortative mating, as observed between divergently-coloured**
 281 **geographic variants (“races”) in *H. melpomene* (Jiggins et al. 2004) or in *H. timareta* (Sanchez et al**
 282 **2015). Few polymorphic populations have been studied in *Heliconius*, besides *H. numata*, but**
 283 **partial assortative mating based on colour differences was found in *H. cydno alithea* (Chamberlain**
 284 **et al. 2009 Science).**

285 To test our hypothesis we ran forward simulations **under two models: one model** with
 286 disassortative **and one model with** assortative mating using slim v3.6 (Messer et al. 2013). **Both**
 287 **models were run using different strengths of mate choice, notably in order to compare the level of**
 288 **genetic diversity under random mating to the level observed under scenarios with strong mate**
 289 **choice. We attempted to choose both biologically and computationally realistic parameters for our**
 290 **models.**

291 We simulated a stepping stone model with 10 demes **in order to reflect the significant isolation by**
 292 **distance pattern observed here. Each deme was** composed of 1,000 diploid individuals and
 293 connected by a (symmetric) migration parameter (m). Each individual received neutral and
 294 deleterious (ratio 16:6) mutations at a rate $\mu = 1e-8$ $\mu/\text{bp}/\text{generation}$ (rescaled to $\mu = 1e-6$ for faster
 295 simulation of a larger population). We simulated an individual with a pair of 1Mb chromosomes,
 296 including a single locus with 5 **possible** alleles with perfect dominance (allele 1 > allele 2 > allele 3
 297 > allele 4 > allele 5) given 5 possible alternative phenotypes (referred hereafter as “morph”) **based**
 298 **on patterns of dominance observed elsewhere (Le Poul et al. 2014).** Each allele was fully linked (no
 299 recombination) with a given **set of** deleterious recessive mutations, generating overdominance at
 300 this locus so that polymorphism is always maintained. Local adaptation was introduced in the
 301 model through a single parameter defining randomly which morphs were favoured in each
 302 population. In each population, either 2 or 3 morphs benefited from a fitness advantage compared to

303 the others. In *H. numata* up to 7 distinct wing-pattern morphs can co-occur within a single locality
 304 (Chouteau et al. 2017). Therefore, we allowed multiple morphs to coexist within a deme. Fitness
 305 varied between 0 (= null fitness for non-locally adapted individuals) and 1 (no reduction of fitness).
 306 We tested 4 possible values for this parameter (0, 0.25, 0.5 and 1).

307 Finally, disassortative mating was controlled by a mate choice parameter defining whether a morph
 308 would reproduce with another morph. The strength of the parameter varied between 0 (= complete
 309 disassortative mating, meaning that a given individual mates only with a different morph) and 1 (=
 310 no mating weight, meaning random mating). We tested 4 possible values for this parameter (0, 0.25,
 311 0.5 and 1 (i.e. meaning random mating).

312 We run the model for 80,000 generations to reach demographic equilibrium and assess levels of
 313 synonymous diversity (π_s). We tested all combinations of the 4 values for levels of disassortative
 314 mating and local adaptation and ran 10 replicates per combination in order to estimate the variance
 315 around π_s .

316 Similarly, we run a model with strict assortative mating, controlled by a parameter defining whether
 317 similar morphs reproduced together. The strength of the parameter varied between 0 (complete
 318 assortative mating where a given individual mate only with an identical morph) and 1 (where
 319 individual mates randomly with regards to the morph). We tested 4 possible values for this
 320 parameter (0, 0.25, 0.5 and 1). As for disassortative mating, all combinations of assortative mating
 321 and local adaptation values were tested. For each model we tested 3 values for the migration rate, m
 322 = 1e-4, 1e-6 and 1e-8, resulting in a total of 54 comparisons.

323 For graphical display in Figure 4, the values of assortative/disassortative mating were rescaled on a
 324 scale between -1 (complete assortative mating) and 1 (complete disassortative mating) with 0
 325 indicating random mating. Values between -1 and -0.25 indicated cases of assortative mating, while
 326 values between 0.25 and 1 represent disassortative mating.

327

328 Results

329 Using cross validation error as a measure of the optimal number of clusters with Admixture, we
 330 found that $K=2$ was the optimal cluster number describing within-species genetic variation in *H.*
 331 *numata* (Fig 2A). One cluster corresponds to the Atlantic population, forming a well-differentiated
 332 genetic entity compared to all other *H. numata* populations. All Amazonian populations of *H.*
 333 *numata* showed remarkable uniformity. This pattern is consistent with the population structure
 334 inferred using microsatellite markers (Fig S1). Population structure revealed by PCA is in line with
 335 the admixture analysis (Fig 2B). Individuals from the Atlantic populations of *H. numata* clustered
 336 together to one side of the first PCA axis, whereas all other individuals from all other populations
 337 clustered to the other side. The second axis of the PCA separates individuals from French Guiana
 338 from the other samples of the upper Amazon. This separation was not found with Admixture (i.e.
 339 with $K=3$) from the complete dataset. To better investigate the existence of a hierarchical
 340 population structure, we excluded individuals from the Atlantic populations and compared
 341 individuals from French Guiana to a randomly sampled set of Peruvian individuals. In this case we
 342 found a clear separation in two groups corresponding to French Guiana and Peru (Fig S2A). The
 343 same pattern was observed when replacing Peru by Colombia or Ecuador (Fig S2B,C). In
 344 accordance, pairwise genome-wide estimates of differentiation (F_{ST}) between *H. numata*
 345 populations showed elevated values when comparing the Atlantic population to other populations,
 346 low values between French Guiana and other Amazonian populations, and were lower when
 347 comparing pairs of Amazonian populations outside of French Guiana (Fig 2C, Table S4). For
 348 instance, the population from La Merced in Peru shows an $F_{ST} = 0.032$ with the population from

349 French Guiana at a distance of 3019km, but an $F_{ST} = 0.311$ (an order of magnitude higher) with the
 350 Atlantic population at a similar distance. The comparison between La Merced and Ecuador was
 351 even lower ($F_{ST} = 0.0159$). Isolation by distance among Amazonian populations of *H. numata*,
 352 estimated using the proxy $F_{ST}/1-F_{ST} \sim \log_{10}(\text{km})$ was significant ($R^2 = 0.41$, $p = 1.61e-06$, slope =
 353 0.02). Comparison among other species did not reveal any significant IBD ($R^2 = 0.01$, $p = 0.29$,
 354 slope = 0.12). An analysis of the slope revealed a lower rate of increase in F_{ST} with distance in *H.*
 355 *numata* compared to all other taxa (Fig 2C, Table S4, Supp Text S1). By contrast, IBD between
 356 Atlantic and Amazonian populations of *H. numata* is close to what is observed in other species, and
 357 not significantly different (see Supp. Text S1).

358

359 Analyses of genetic diversity show that all populations of *H. numata*, except those from the Atlantic
 360 Forest, have a similarly high genetic diversity (Fig 3A). By contrast, closely related *Heliconius* taxa
 361 show lower genetic diversity (Fig 3A). These patterns are similar to those obtained using G-PhoCS
 362 to analyse the demographic histories in a phylogenetic context, where Amazonian populations of *H.*
 363 *numata* show higher population sizes compared to Atlantic Forest populations (Fig 3B, Table S5).
 364 G-PhoCS analyses also show a demographic history in which gene flow plays a crucial role (Table
 365 S6). For instance, our analyses show significant gene flow right at the beginning of the divergence
 366 between *H. ismenius* and the other silvaniforms, as well as in the divergence between *H. pardalinus*
 367 and *H. elevatus*. The effective population sizes inferred from Atlantic genomes are one order of
 368 magnitude lower than those obtained using *H. numata* populations from other localities (Fig 3A and
 369 Table S5). In our cladogram, the increase in *H. numata* population size is restricted to the
 370 Amazonian branch, excluding Atlantic populations.

371

372 **Demographic reconstruction from $\partial a \partial i$**

373 The model selection procedure based on AIC gave higher support for a model of secondary contact
 374 (SC) in the pairwise comparison between *H. numata* from Peru and *H. numata* from Brazil. The
 375 pairwise comparison between *H. numata* and *H. pardalinus* supported a model of divergence with
 376 continuous gene-flow (IM) (Table S7, Figure S3). All models supported an expansion occurring in
 377 the ancestral population, followed by further growth of the *H. numata* carrying the inversion
 378 supergene to reach a size of several millions, which was by far the largest effective size compared
 379 to all other species. This stands in stark contrast with the results observed in the samples from
 380 Brazil (which do not harbor the inversion) (Table 2). Accordingly, *H. numata* populations from the
 381 Atlantic forests of Brazil appear to have been subject to a bottleneck at the start of their divergence
 382 from Amazonian populations, followed by exponential growth, suggesting a strong (and recent)
 383 founding event, leading to a comparatively smaller population size than that observed in the rest of
 384 *H. numata*. It is worth noting however that effective population size was hard to estimate in
 385 pairwise comparisons between *H. numata* from Peru and SE Brazil. Indeed, parameter uncertainty
 386 was large, and model residuals (Figure S3) were also large. Our results indicated that *H. pardalinus*
 387 displayed an initially large population size followed by a comparatively smaller size expansion than
 388 *H. numata* (Table 2). There was a large variance in estimates of effective population size in *H.*
 389 *numata* from Peru in the two inferences (with *H. pardalinus* or with Brazilian *H. numata*). Yet, in
 390 both inferences, the effective population sizes of the Peruvian *H. numata* were large (greater than 11
 391 million). Estimates of current effective population sizes are therefore qualitatively similar to those
 392 from G-phocs.

393

394 **Forward simulations**

395 Forward simulations under different levels of local adaptation (controlled by the strength of
 396 divergent selection), disassortative mating and migration are displayed in Figure 4B. The same
 397 results under a model of assortative mating involving different levels of selection and migration are
 398 displayed in Figure 4A. Overall, synonymous genome-wide nucleotide diversity (π_s) was higher in
 399 73 % of the models including disassortative mating (average $\pi_s = 0.0145$) when compared to their
 400 equivalent under assortative mating (average $\pi_s = 0.011$), a weak but significant difference (p
 401 < 0.01). In summary, modest differences were observed among models with different strength of
 402 divergent selection or disassortative mating, the most influential variable being the rate of migration
 403 (Figure 4).

404

405 Discussion

406 Our results suggest that populations displaying inversion polymorphism in the *P* supergene in *H.*
 407 *numata* also display distinctive population demography and gene flow. Differences in demographic
 408 and differentiation regimes associated with structural variation at this locus are revealed when
 409 comparing polymorphic populations of *H. numata* to closely-related monomorphic taxa, such as (1)
 410 peripheral populations of *H. numata*, (2) sister taxa, and (3) inferred ancestral lineages. This
 411 suggests that the existence of a mimicry supergene controlling polymorphism in *H. numata* may be
 412 associated, in time and in space, with major differences in population biology. We hypothesize this
 413 to be due to a change in the balancing selection regime due to heterozygote advantage (Jay et al.
 414 2021) and in the associated evolution of disassortative mating (Chouteau et al. 2017) following the
 415 onset of inversion polymorphism, causing direct effects on ecological parameters such as gene flow,
 416 immigration success and effective population size. Testing this hypothesis through forward
 417 simulation yielded mixed evidence for a genome-wide effect of this disassortative mating,
 418 especially when compared to a simple model of random mating.

419

420 Our analyses show large-scale variation in genetic diversity among closely related taxa in this clade
 421 of *Heliconius* butterflies. Within *H. numata*, genetic diversity found in polymorphic Amazonian
 422 populations is ~ 4 times higher than in populations from the Atlantic Forest. Generally, Amazonian
 423 populations of *H. numata* harbour the highest genetic diversity in the entire *melpomene/silvaniform*
 424 clade, which contrasts with the low diversity found in the most closely related taxa such as *H.*
 425 *ismenius* or *H. besckei*. Inferring historical demography during the diversification of the *H. numata*
 426 lineage reveals that the large effective population size in that species is only associated with the
 427 branch representing polymorphic, Amazonian *H. numata* populations, while internal branches all
 428 show very low diversity estimates. This suggests that ancestral and putatively monomorphic
 429 populations of *H. numata* were similar in their diversity parameters to current sister species *H.*
 430 *ismenius* populations, or to current peripheral Atlantic *H. numata* populations. Although low-
 431 diversity lineages could have lost diversity due to recent events such as strong bottlenecks, as
 432 estimates of effective population size from $\partial a \partial i$ indicated for the Atlantic population. Nevertheless
 433 our $\partial a \partial i$ estimates do suggest that the Amazonian populations of *H. numata* underwent a dramatic
 434 increase in effective population size posterior to their split with Central American (*H. ismenius*), *H.*
 435 *pardalinus* and the Atlantic populations. Those findings are in agreement with G-Phocs analyses.
 436 The Amazonian branch of the *H. numata* radiation is characterized by the long-term maintenance of
 437 inversion polymorphism, triggered by the introgression of a chromosomal inversion about 2.2 Ma
 438 ago. Therefore, the major shift in demography between Amazonian and Atlantic populations indeed
 439 appears to coincide **in time**, at least in the broad sense, with the occurrence of inversion
 440 polymorphism, even though the lack of replication of this event impedes firmly establishing

441 causality.

442

443 Another striking result is the lack of genetic structure displayed by *H. numata* across the Amazon,
 444 with all Amazonian populations forming a single genetic cluster. Only Atlantic Forest populations
 445 stand out and display high differentiation with other *H. numata* from the rest of the range. French
 446 Guiana and Peruvian populations, separated by over 3000 km across the Amazon, are weakly
 447 genetically differentiated compared to pairs of populations at comparable distances in other species,
 448 and show only modestly stronger differentiation than pairs of *H. numata* populations taken at short
 449 distances. *Heliconius numata* populations from the Amazon show significantly lower isolation by
 450 distance than all other taxa, as measured by the change in F_{ST} across distance (F_{ST}/km) (Fig. 2C),
 451 with a very distinctive, flat slope of isolation by distance. The only exception is found when
 452 comparing Amazonian populations with Atlantic Forest populations of Brazil, displaying a level of
 453 differentiation in line with that of pairs of populations at similar distances within other taxa.

454

455 Effective population size is affected by census size, mating system, and the force and type of
 456 selection acting on traits (Charlesworth 2009). Selection is often viewed as a force only affecting
 457 the genetic variation around specific, functional loci in the genome, but it may also affect whole
 458 genome diversity, for instance when its action is sufficient to modify local demography or mating
 459 patterns. In *H. numata*, morphs and therefore inversion genotypes show disassortative mate
 460 preferences, i.e., they preferentially mate with individuals carrying different chromosome types
 461 (Chouteau et al. 2017). Disassortative mating enhances heterozygosity and the mating success of
 462 individuals expressing rare alleles (negative frequency dependence) (Knoppien 1985; Hedrick et al.
 463 2018). Consequently, immigrants expressing rare, recessive alleles have a mating advantage in *H.*
 464 *numata*. Disassortative mating associated with the supergene should therefore bring an advantage to
 465 immigrant genomes in LD with recessive supergene alleles, possibly enhancing genome-wide gene
 466 flow. Supergenes are also characterised by single-locus Mendelian inheritance, by which mimicry
 467 phenotypes are maintained in the face of recombination, even after immigration. Effective
 468 migration regime in populations harbouring a mimicry supergene is therefore likely to be quite
 469 different to that observed in other mimetic taxa such as *Heliconius melpomene* or *H. erato*, in which
 470 mimicry variation is controlled by multiple loci with diverse dominance patterns. In those taxa,
 471 hybrid offspring display recombinant patterns breaking down mimicry, even after multiple
 472 generations of backcrossing, and pure forms mate assortatively with respect to wing pattern
 473 (McMillan et al. 1997, Mallet et al. 1998, Jiggins et al. 2001); both processes select against mimetic
 474 variants migrating from adjacent areas with distinct warning patterns. The expectation is that
 475 immigrant genomes should be consistently associated with mimicry breakdown in the case of
 476 multilocus architectures, which should translate into a reduction of the effective migration genome-
 477 wide, compared to situations with polymorphic mimicry supergenes. In *H. numata*, the evolution of
 478 a polymorphic mimicry supergene and disassortative mate preferences could therefore explain the
 479 relative lack, compared to other *Heliconius* taxa, of differentiation among polymorphic populations,
 480 even across large distances. Furthermore, enhanced gene flow could also cause an increase in
 481 effective population size estimates (Slatkin 1987), putatively explaining why polymorphic
 482 populations of *H. numata* harbour the highest genetic diversity, and display the highest N_e estimates
 483 in the entire *melpomene*-silvaniform clade of *Heliconius*. These hypotheses are also supported by
 484 our forward simulation which revealed an effect of the mating system. Yet, most of the difference in
 485 genetic diversity was due to a assortative mating reducing genetic diversity rather than a
 486 diversifying effect of disassortative mating. In addition, our results also suggest that a simple model

487 of random mating may explain well the data, thus purely demographic expansions may also
 488 generate high genetic diversity and high effective population size, as observed from our ∂adi
 489 demographic modelling. Overall, the change in genetic diversity due to mating system variation is
 490 limited compared to the effect of migration. Indeed, our simulation results indicated that migration
 491 between demes has the stronger effect on synonymous diversity. In addition, the magnitude of
 492 difference in the simulation is weak relative to the observed differences of genetic diversity between
 493 *H. numata* from Brazil and *H. numata* from Peru. Whether these differences between our
 494 simulations and those empirical observations can be directly compared is not straightforward
 495 however given the simplified assumption required by our modelling approaches. In all cases, our
 496 demographic simulations with ∂adi , do suggest a strong effect of founder events/bottlenecks in
 497 reducing the genetic diversity of the Brazilian population. Another limit of our model is that we
 498 tested the effect on genetic diversity on a 1 Mb segment of a single chromosome, not on unlinked
 499 chromosomes. Although testing the effect on unlinked data would be relevant, it would
 500 substantially increase the compute time, so we choose to focus on this simpler model and leave the
 501 question of large scale effect for future investigations.

502

503 Alternative processes may also contribute to the observed patterns. Amazonian and Atlantic
 504 populations may differ in other aspects that could also result in differences in genetic diversity.
 505 Habitat availability and structure may be different, possibly entailing differences in the maintenance
 506 of diversity. The Atlantic Forest is vast in area, but may represent a smaller biome compared to the
 507 Amazon, and is isolated from the bulk of the range of *H. numata*, which could result in populations
 508 displaying characteristics of peripheral populations with smaller effective population sizes (Eckert
 509 et al. 2008). Reduced effective population size is supported by our data. One major caveat
 510 associated to our inference remains the small number of individuals ($n = 12$) from the Atlantic
 511 Forest. Genetic diversity might be underestimated, notably if populations have a history of
 512 fragmentation in this area. The other *Heliconius* species in the clade have much in common with *H.*
 513 *numata* in terms of habitat and general ecology, yet their niche and life-history specificities and
 514 their phylogenetic histories may result in consistent differences with the polymorphic *H. numata*
 515 populations. All those specificities may contribute to the observed pattern in which polymorphic
 516 Amazonian populations of *H. numata* display high effective population size and a weak geographic
 517 structure in genome-wide genetic variation. Yet this pattern of variation correlates parsimoniously
 518 with the evolution of a supergene causing disassortative mating and single-locus control of mimicry
 519 variation in Amazonian *H. numata* populations, which provides an elegant mechanism explaining
 520 their differences with extant and ancestral closely-related lineages. However, we cannot rule out a
 521 role for conjectural differences in ecology and geography with all other taxa.

522

523 In conclusion, our results show a remarkable contrast in the demography and differentiation of
 524 populations within the Amazonian range of *H. numata* compared to closely related taxa and
 525 ancestral lineages, as well as with other taxa in the *melpomene/silvaniform* clade. The
 526 distinctiveness of this widely polymorphic species in the clade is consistent with the hypothesis that
 527 the evolution of a supergene maintained by balancing selection represents a major transition in this
 528 lineage, triggering changes in genome-wide patterns of diversity and population ecology over the
 529 last 2 million years since its formation. If this hypothesis is correct, the evolution of a locus under
 530 balancing selection may therefore feed-back on population ecology and diversification, and
 531 consequently on speciation.

532

533 Eco-evolutionary feedbacks between changes in genomic architecture and the ecological parameters
 534 of populations are still not well understood and few cases have been studied. The evolution of self-
 535 incompatibility in plants, affecting the rules of mating and feeding back on population ecology,
 536 connectivity, and demography, may be one example, but effects of the evolution of trait genetic
 537 architecture on population ecology may be more common than previously thought. In our study,
 538 more work on the determinants of variation in effective population sizes in the genus *Heliconius* is
 539 needed to determine the precise impact of the supergene on demography in *H. numata*. We believe
 540 that our results emphasise a potential link between genomic architecture, selection and demography,
 541 and should inspire future theoretical and modelling studies. Overall, our result suggests that
 542 balancing selection maintains structural polymorphisms affecting life-history traits may have a
 543 profound influence on species ecology.

544

545 **Contributions:**

546 MARdC, PJ, QR and MJ designed the study and wrote the manuscript. BH, AVLF, TTT, RRR,
 547 KLSB provided the Atlantic samples. CS provided the Colombian samples. MARdC, PJ and QR
 548 performed genomic analyses and simulations with input from AW. MARdC, PJ, MJ, FPP and MC
 549 collected the Peruvian and Ecuadorian samples. MC performed microsatellite analyses and
 550 organized fieldworks and butterfly rearing. All authors contributed to editing the manuscript.

551

552 **Acknowledgements:**

553 This work was funded by grants Hybevol (ANR-12-JSV7-0005) and Supergene (ANR-18-CE02-
 554 0019-01) from the Agence Nationale de la Recherche and European Research Council Grant
 555 MimEvol (StG-243179). We acknowledge the Genotoul and the Montpellier Bioinformatics
 556 Biodiversity (MBB) platforms for providing us with calculation time. We thank Dr. Vitor Becker, at
 557 the Serra Bonita Reserve (Bahia), Alexandre Soares, at the MN/UFRJ (Rio de Janeiro) and Dr.
 558 Marcelo Duarte at the MZ/USP (Sao Paulo) for their contribution to the collection of butterflies in
 559 Brazil. Field collections in Colombia were conducted under permit no. 530 issued by the Autoridad
 560 Nacional de Licencias Ambientales (ANLA). Field collections in Brazil were conducted under
 561 ICMBio permit SISBIO-10438. We are grateful to Marianne Elias and Violaine Llaurens for
 562 comments and discussions. AVLF acknowledges support from Fundação de Amparo à Pesquisa do
 563 Estado de São Paulo – (FAPESP) (Biota-Fapesp grants 2011/50225-3, 2013/50297-0 and
 564 2021/03868-8) and Conselho Nacional de Desenvolvimento Científico e Tecnológico (CNPq)
 565 (421248/2017-3 and 304291/2020-0). KLSB acknowledges the financial support of FAPESP
 566 Process # 2012/16266-7. Brazilian specimens are registered under SISGEN (A701768).

567

568 **Data availability:**

569 The raw sequence data were deposited in NCBI SRA and accession numbers are indicated in
 570 Supplementary table 3.

571

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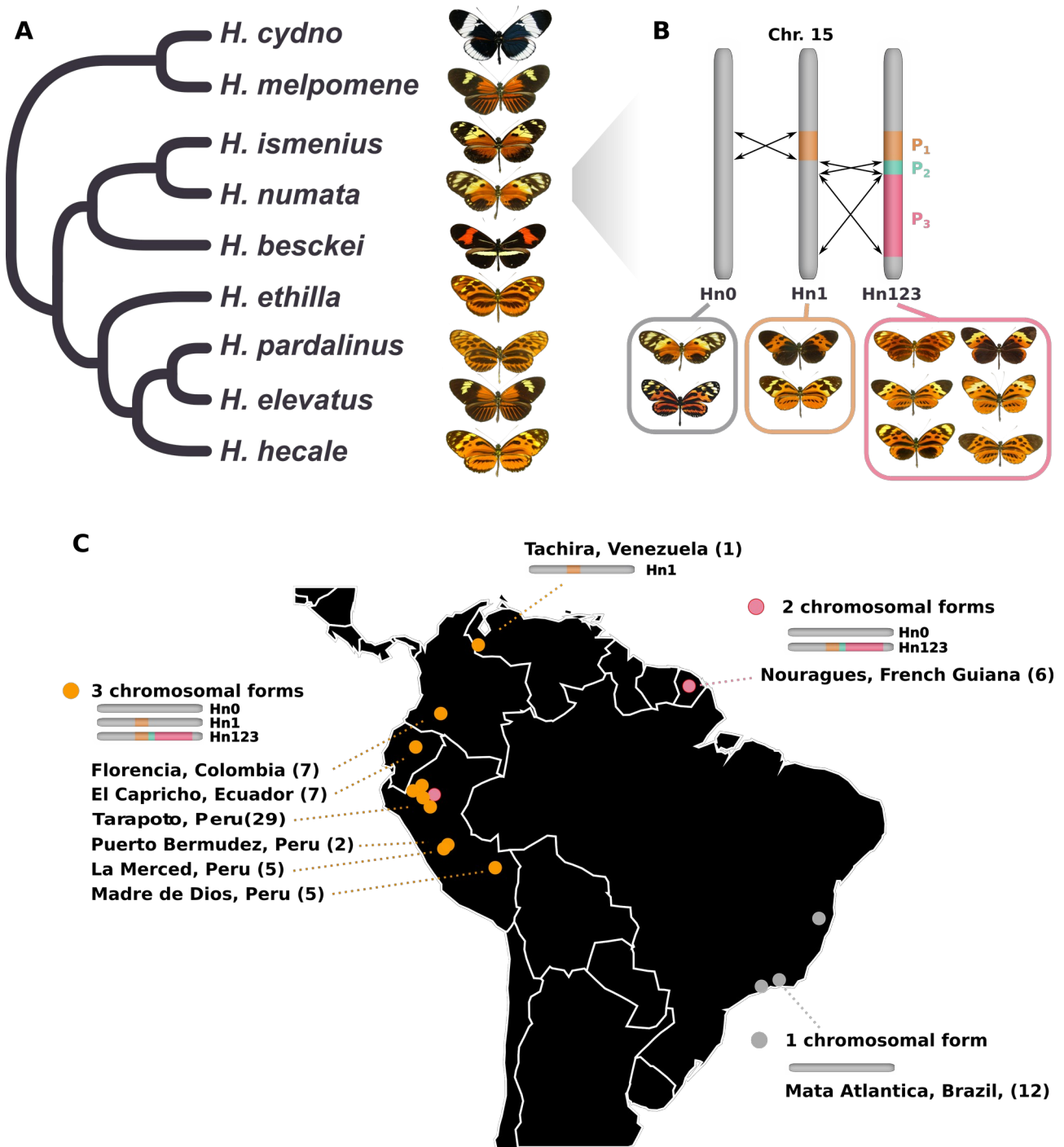
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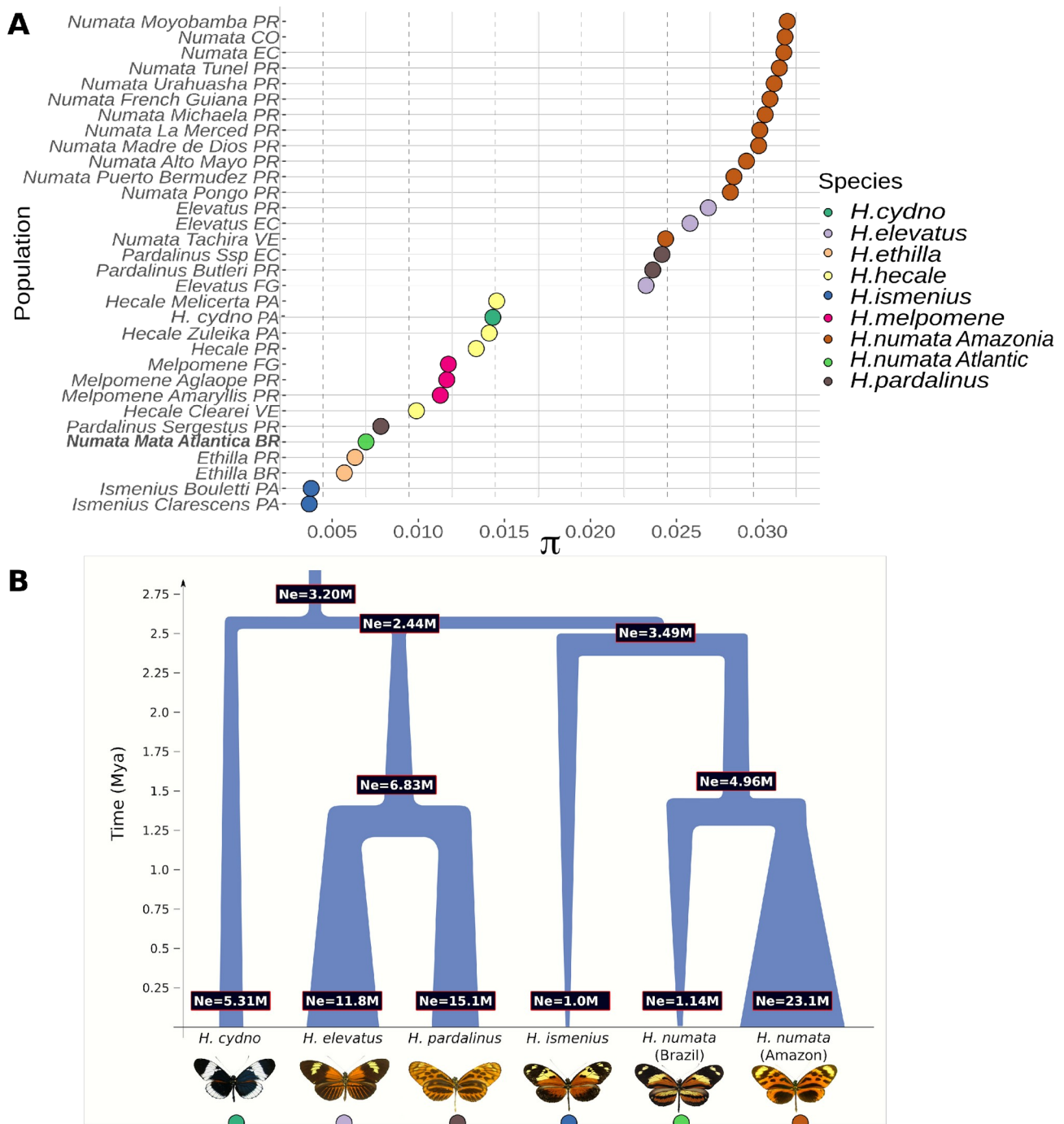
713 **Figure 1 | Genetic and population structure at the P supergene.**

714 **A.** Schematic phylogeny of the sampled species. It includes all members of the silvaniform clade
 715 and two outgroups, *H. melpomene* and *H. cydno*.

716 **B.** Schematic description of the genetic structure of the P supergene. Three chromosomal
 717 arrangements coexist in *H. numata* and are associated with different morphs.

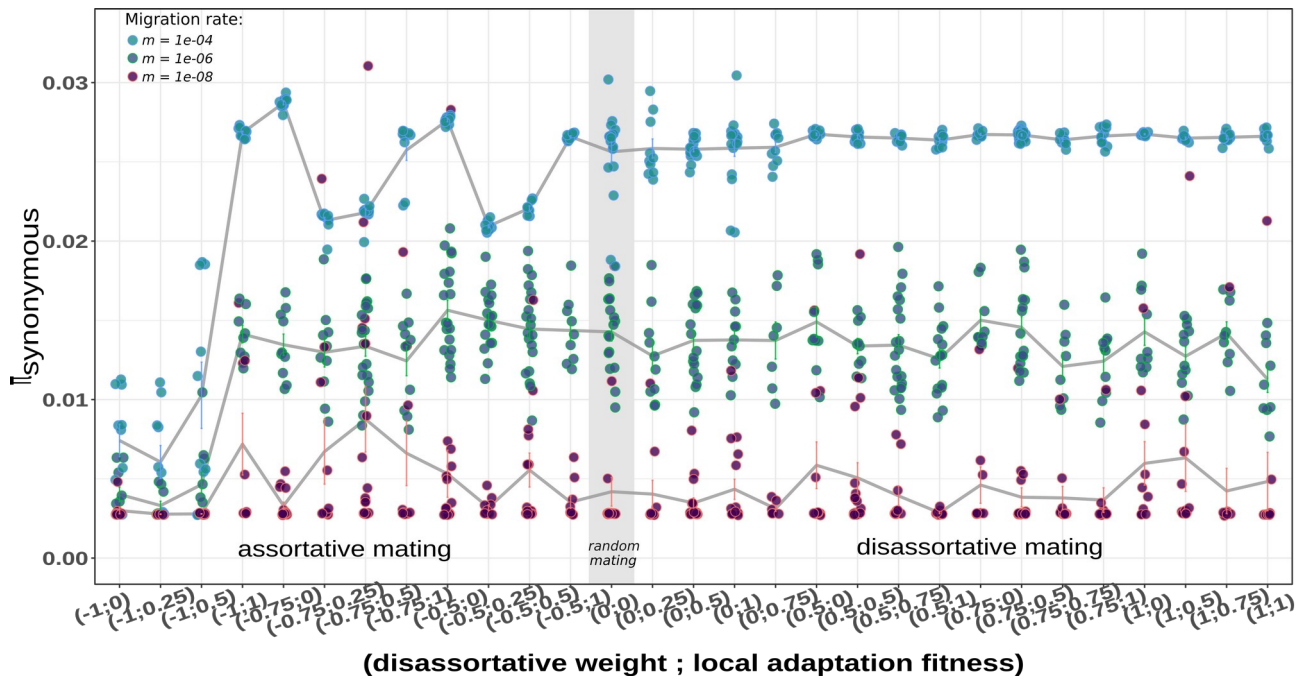
718 **C.** Origin of *H. numata* specimens used for analyses and distribution of chromosome arrangements
 719 across the neotropics. Numbers in brackets indicate sampled specimens in each locality (the
 720 Tarapoto population lumps several neighbouring subsamples on the map)

721



731 **Figure 3 | Variation in present and past effective population size in *Heliconius* species**

732 **A.** Variation in π in several *Heliconius* populations, showing higher genetic diversity in *H. numata*
 733 populations from the Amazon than other taxa. Population names indicates their origin as in Figure 2
 734 (e.g. PR=Peru), with the addition of PA=Panama. The *H. numata* population with a lowest diversity
 735 is the one from the Atlantic forest (Brazil). **B.** Schematic representation of Gphocs results
 736 (presented in table S5-6). Gene flow was modelled but not represented graphically for clarity.
 737 showing that Amazonian populations of *H. numata*, which have the P supergene, show a dramatic
 738 increase in population size posterior to their split with the Atlantic populations of Brazil, which lack
 739 the supergene.



741 **Figure 4 | Weak but significant differences in synonymous nucleotide diversity (π_s) emerged at**
 742 **a genome-wide scale under divergent selection and mating regime.** Results from forward
 743 simulations of 10 replicate per parameter combination. A set of 10 demes with varying levels of
 744 migration, local adaptation fitness and different mating strategy (from assortative mating to
 745 disassortative mating) are presented. Shown are levels of synonymous diversity obtained under each
 746 combination of parameters. Parameters on the left part of the brackets display the disassortative
 747 mating weight (from -1 to 1). Parameter on the right side of the brackets displays the fitness value
 748 for local adaptation. A left value of -1 in the bracket means complete assortative mating and 0
 749 means random mating. 1 = complete disassortative mating. A right value of 0 in the bracket means a
 750 fitness of 0 for non locally adapted individuals in a deme. A value of 0.5 means a reduced fitness of
 751 0.5 relative to the maximum value. A value of 1 means no loss of fitness.

752 **Table 1: Estimates of demographic parameters for each best model in each comparison.**
 753 Biological parameters assumed a mutation rates of 2.9e-09 μ /bp/generation. NeAnc, = Effective
 754 population sizes for the ancestral population. NeParda, NePeru and NeBrasil corresponds to
 755 effective population size of *H. pardalinus* *H. numata* in Peru and *H. numata* in Brazil respectively.
 756 NeB1Parda and NeB2Peru, NeB1Brasil provides effective population size for the population *H.*
 757 *pardalinus* *H. numata* in Peru and *H. numata* in Brazil respectively after their expansion. Hrf =
 758 Hill-Robertson factor reflecting the reduction in effective population size due to linked selection. Q
 759 = proportion of the genome with a reduced effective population size. m12 = migration from
 760 population 2 into population 1. m21 = migration from population 1 into population 2 (scaled
 761 according to $2N^*_{ref}m_{ij}$). T_{ANC} = Time of population size change in the ancestral population, T_{p1} =
 762 Time of size change in population 1 and T_{p2} = Time of size change in population 2. T_{ANC} , T_{SPLIT} and
 763 T_{sc} , T_{p1} , T_{p2} are provided assuming four generations per year.

<i>H. Numata</i> Perou - <i>H. pardalinus</i>		<i>H. numata</i> Perou- <i>H. numata</i> Brazil	
Best model	IMA2NG	Best. Model	SCA2NG
Log Lik.	-634.48	Log. Lik	-1430.36
Theta	16,757.39	Theta	3481.23
Nref	6,829,000	Nref	1,036,000
NeAnc	538,000 [0 – 1,281,000]	NeAnc	12,912,000[702,000 – 32,850,000]
NeParda	9,209,000 [3,163,000 – 15,254,000]	NeBrasil	83,000 [74,000 – 93,000] 60,373,000
NePeru	650,000 [625,000 – 676,000]	NePeru	[6,969,000 – 113,775,000]
NeB1Parda	25,233,000 [7,282,000 – 48,477,000] 11,746,000	NeB1Brasil	292,000 [147,000 – 465,000] 373,573,000
NeB2Peru	[6,511,000 – 17,368,000]	NeB2Peru	[13,310,000 – 1,190,748,000]
m12	1.99e-08 [0 – 8.59e-08]	m12	4.71E-08 [0 – 1.25e-07]
m21	1.37e-07 [9.84e-8 – 1.79e-7]	m21	3.42E-09 [0 – 2.19e-07]
T_{ANC}	3,237,000 [3,039,000 – 6,169,000]	T_{ANC}	501,000 [254,000 – 748,000]
T_{SPLIT}	1,892,000 [1,674,000 – 2,111,000] 108,000	T_{SPLIT}	110,000 [69,000 – 151,000]
T_{p1}	[662,000 – 1,500,000]	T_{p1}	11,000 [6,000 – 17,000]
T_{p2}	964,000 [779,000 – 1,150,000]	T_{p2}	55,000 [0 -157,000]
H_{RF}	0.10 [0.03 – 0.17]	T_{sc}	41,000 [7,000 – 75,000]
Q	0.37 [0 – 0.99]	H_{RF}	0.47 [0.43 – 0.50]
		Q	0.47 [0.39 – 0.56]

764

765 **List of Supplementary Materials:**

766 Table S1-6

767 Fig S1-2

768 Text S1

769