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Selective sweeps on novel and introgressed variation shape mimicry loci in a butterfly adaptive radiation

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Abstract

Natural selection leaves distinct signatures in the genome that can reveal the targets and history of adaptive evolution. By analysing high-coverage genome sequence data from 4 major colour pattern loci sampled from nearly 600 individuals in 53 populations, we show pervasive selection on wing patterns in the *Heliconius* adaptive radiation. The strongest signatures correspond to loci with the greatest phenotypic effects, consistent with visual selection by predators, and are found in colour patterns with geographically restricted distributions. These recent sweeps are similar between co-mimics and indicate colour pattern turnover events despite strong stabilising selection. Using simulations, we compare sweep signatures expected under classic hard sweeps with those resulting from adaptive introgression, an important aspect of mimicry evolution in *Heliconius* butterflies. Simulated recipient populations show a distinct ‘volcano’ pattern with peaks of increased genetic diversity around the selected target, characteristic of sweeps of introgressed variation and consistent with diversity patterns found in some populations. Our genomic data reveal a surprisingly dynamic history of colour pattern selection and co-evolution in this adaptive radiation.

Introduction

Identifying targets of selection and reconstructing their evolutionary history is central to understanding how populations adapt [1–3]. In particular, genome sequences contain a rich
The role of selective sweeps in the evolution of mimicry

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Abbreviations: CLR, composite likelihood ratio; CRE, cis-regulatory element; HIH12, integrated haplotype homozygosity score; JAK/STAT, Janus kinase/signal transducer and activator of transcription; Mya, million years ago; PIR, phase informative read; PSMC, Pairwise Sequentially Markovian Coalescent; SFS, site frequency spectrum; SF2, SweepFinder2.

source of information about past events in natural populations. The action of recent positive selection can leave a distinct signature known as a ‘selective sweep’, which provides information on the genomic location of targets of positive selection and the timing and strength of selection [4,5]. Although many classic examples of selective sweeps have been found in domesticated populations, such as maize [6], chicken [7], and cattle [8], or in humans [9], increasingly natural populations are also studied. Using genomic data, these latter studies can reveal the genetic architecture and evolutionary history of ecologically relevant traits [10–13] and provide insights into the action of natural selection by complementing field and experimental studies [14–16]. However, to date, few molecular studies of natural populations have used broad sampling in adaptive radiations with varying selection pressures and sources of adaptive variation for the same trait. Such studies will allow the investigation of both complexity and general mechanisms of natural selection in the wild at the genotypic level, especially where there is a priori information on the agents and targets of selection.

Positive selection can rapidly change allele frequencies leaving detectable signatures in a genome. These signals can be traced over ecological and evolutionary time scales, during which they are gradually eroded by new mutations and recombination [1]. However, the observed patterns will depend on the sources and frequency of genetic variation upon which selection acts [5]. For example, a classic ‘hard sweep’ due to selection on a single, novel beneficial mutation [4] or a very rare allele from standing variation [17], is distinct from a ‘soft sweep’ due to selection on standing variation already present at an appreciable frequency [17–20] or recurrent mutations [21,22]. Less well studied in the context of selective sweeps is the possibility that a new variant is introduced by gene flow from a related population or distinct species. Accumulating evidence suggests that this reuse of ancient variants is far more common than was previously envisioned [23–26]. However, the sweep signatures created by selection on introgressed and divergent haplotypes, and the effect of migration rate on these signatures, are largely unexplored (but see Setter and colleagues [27]).

Mimicry systems provide some of the best examples of natural selection and adaptation and thus exceptional opportunities to study selective sweeps. In the unpalatable Heliconius butterflies, mimicry of wing patterns is advantageous because resemblance to a common, well-protected pattern confers protection from predator attacks on individuals. The vast majority of pattern diversity seen in this group is controlled by a surprisingly simple genetic system, involving allelic variation at just 4 major effect loci, although additional regulators and modifiers of these mimicry patterns have also been mapped [26,28–34]. Although these regions comprise several genes with a putative function for colour patterning, current evidence suggests a major role for the transcription factors, optix [35] and aristalless, which comes in 2 tandem copies al1 and al2 [28], a signalling ligand, WntA [29], and a gene in a family of cell cycle regulators whose exact function remains unclear, cortex [30]. We therefore refer to these 4 regions by the name of the respective major colour pattern gene throughout the manuscript without excluding the potential involvement of additional genes within these regions. A complex series of regulatory variants at each of these loci is found in different combinations across populations and species, leading to great diversity of wing patterns. In many cases, candidate noncoding, cis-regulatory elements (CREs) are associated with specific wing patterns: CREs in the optix region are associated with the red forewing band, hindwing rays, and dennis patch [36–38]; in the cortex region with the yellow hindwing bar [30,38,39]; in the WntA region with various shape elements of the forewing band [33,38]; and in the aristalless region with white versus yellow colour variation [28].

Colour pattern novelty is generated by mutation, introgression, shuffling, and epistatic interaction of existing CREs that generate new pattern combinations [36,38–41]. In fact, adaptive sharing of mimicry colour patterns has been demonstrated across many species and
populations within the *H. melpomene* and *H. erato* clade [36,38,39,42–46]. The *H. melpomene* clade comprises the sister clades *H. melpomene* and *H. cydno/heurippa/timareta*, which split 1 to 1.5 million years ago (Mya) [47–49] and their outgroup silvaniform clade (4 Mya since divergence) [50]. Well-characterised cases of adaptive introgression in this clade include the exchange of red and yellow elements among *H. melpomene*, *H. timareta*, and the silvaniforms *H. elevatus* and *H. besckei* [36,44,45], as well as the sharing of elements controlling yellow hindwing colouration between *H. melpomene* and *H. cydno* [39]. Consequently, we can assess patterns of selection in well-defined genomic intervals with evidence for dated introgression events [36,39]. Likewise, hybridisation is also important within the *Heliconius erato* clade [46,51,52], but there is no evidence for gene flow between these 2 major clades that split around 12 Mya [50]. *Heliconius erato* comprises several colour pattern races that are co-mimics with *H. melpomene*, *H. timareta*, *H. besckei*, and *H. elevatus* and is often the more abundant co-mimic [53].

*Heliconius* colour patterns are known to be subject to remarkably strong natural selection in wild populations, which has been demonstrated through pattern manipulations [54], reciprocal transplants across a hybrid zone [55], reciprocal transfers between different co-mimic communities [56], and artificial models [57,58]. In all cases, estimates of selection strength were high with $s = 0.52–0.64$ (Table 1). Indirect estimates of selection strength from hybrid zones generated similarly high values with $s = 0.23$ for each of 3 colour pattern loci containing *optix*, *cortex*, and *WntA*, in *H. erato* and $s = 0.25$ for *optix* and *cortex* in *H. melpomene* [59–63] but also include cases of substantial variance in selection coefficients [64] (see Table 1 for details).

Although colour pattern loci in *Heliconius* are well studied, and their adaptive significance is apparent, the impact of selection at the molecular level has never been estimated in detail in natural *Heliconius* populations. Genetic studies have shown that populations often cluster by phenotype rather than geography at colour pattern loci [38,67,68], but these approaches may not detect recent adaptive changes. For example, closely related populations show peaks of high differentiation at colour pattern loci [34,69], but previous studies did not reveal strong sweep signatures [31,32,70], and more recent genomic analysis showed only weak evidence for reduced heterozygosity and enhanced linkage disequilibrium [68]. However, these studies have used either few amplicons or genomic data with small sample sizes and therefore potentially had little power to detect selective sweep signatures.

Here, we obtain a large genomic data set across the *H. melpomene* radiation, featuring both high coverage and large sample size, and combine simulations with population genomic analysis to investigate natural selection at 4 main colour pattern loci. We use forward-in-time simulations to compare the signal produced by classic and introgressed sweeps in genome scan data, to characterise expected patterns for introgressed sweeps under varying effective migration rate and strength of selection, patterns which have previously been little explored [27]. We parameterise our simulations with demographic estimates representative for *Heliconius* in order to inform inferences about the timing of sweeps detected in *Heliconius* populations. Our empirical data set covers almost the entire biogeographic range of an adaptive radiation and demonstrates clear signatures of selective sweeps across many populations. However, many widespread colour patterns show only modest signals of sweeps, with the strongest signals found in populations with geographically restricted patterns, suggesting recent and strong selection. For adaptive introgression, our simulations demonstrate that the signals have distinct shapes, are strongly affected by effective migration rates, and are more challenging to detect. Nevertheless, we identify sweep signatures among populations with known colour pattern introgression. Moreover, we identify new putative targets of selection around colour pattern genes in some populations. Finally, we also analyse genomic data from *H. erato*
populations, representing a distinct radiation of similar wing pattern forms, and find evidence for parallel evolution between co-mimetic butterfly species.

**Results**

**Phylogeography and demography of the *H. melpomene* clade**

We obtained approximately 5.2 Mb of sequence distributed across 8 chromosomes from 473 individuals and 39 populations representing 10 species from the *H. melpomene* clade (S1 and S2 Tables). Phylogenetic reconstructions confirmed that *H. cydno* populations, with the sole exception of *H. c. cordula* found east of the Andes and in the Magdalena Valley, and *H. timareta* populations from east of the Andes cluster as separate lineages from the *H. melpomene* clade (Fig 1B and 1D). Phylogenetic inferences including all sequenced regions agreed with previous multilocus phylogenies, in which *H. cydno* and *H. timareta* form a sister clade to *H. melpomene* (Figs 1D and S1) [44,50]. The tree built using only neutral background data (i.e., regions a priori not suspected to be under mimicry selection, see Materials and methods) largely clustered populations according to geography, i.e., *H. cydno* with western *H. melpomene* and *H. timareta* with eastern *H. melpomene* subspecies (Fig 1B and 1D). The neutral topology is consistent with ongoing gene flow between sympatric populations resulting in highly heterogeneous relatedness patterns along the genome [71,72]. Six out of nine individuals with

| Table 1. Direct and indirect estimates of selection on colour pattern loci. Combined estimates are integrating the effect of all loci involved in warning colouration. Regions/modules associated with *optix*: D, B; with cortex: Cr, Yb, N; with *WntA*: Sd, Ac; with *aristaless*: K. |
| Species | Colour pattern region under consideration | Estimated selection coefficient (s) | Method | Source |
|---------|-----------------------------------------|----------------------------------|--------|--------|
| *H. erato* | *optix* (red band) | s_D = 0.22 | Pattern manipulation, survival and bird attack rate | Benson [54] (s estimate calculated in Mallet and colleagues [65]) |
| *H. erato* | *optix/cortex/WntA* | combined s = 0.52 avg. per locus s = 0.17 | Reciprocal transplants, survival | Mallet and Barton [55] |
| *H. erato* | *optix/cortex/WntA* | s_D = 0.33 s_Cr = 0.15 s_Sd = 0.15 | Reciprocal transplants, survival | Mallet and colleagues [65] |
| *H. erato* | *H. melpomene* | *optix/cortex/WntA* | avg. per locus s = 0.23 avg. per locus s = 0.25 | Cline and LD analysis in a hybrid zone | Mallet and colleagues [61] |
| *H. erato* | cortex | s_Cr = 0.20–0.22 | Cline analysis in a hybrid zone | Blum [66] |
| *H. cydno* (polymorphic mimic) *H. sapho* (model) *H. eleuchia* (model) | *aristaless* | s = 0.64 | Reciprocal transplant of polymorphic *H. cydno* | Kapan [56] |
| *H. erato* | *optix/cortex/WntA* | avg. per locus s = 0.22 s_D = 0.38 s_Cr = 0.17 s_Sd = 0.15 | Cline and LD analysis in a hybrid zone | Rosser and colleagues [62] |
| *H. melpomene* | *optix/cortex* | avg. per locus s = 0.3 s_D = s_Yb = s_N = 0.31 s_B = 0.19/0.15 | Cline analysis in a hybrid zone | Salazar [63] |
| *H. erato* | *optix/WntA* | s_D = 0.15 s_Yb = 0.04 | Cline analysis in a hybrid zone | Thurman and colleagues [64] |
| *H. melpomene* | *optix/WntA* | s_D = 0.27 s_Ac = 0.04 | Cline analysis in a hybrid zone | Salazar [63] |
| *H. erato* | cortex | s_Cr = 0.05 | Cline analysis in a hybrid zone | Thurman and colleagues [64] |

**Abbreviations:** LD, linkage disequilibrium

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Fig 1. Distribution, phylogenetic relations, major colour pattern loci, and sequence capture targets of the *H. melpomene*, *H. cydno*, and *H. timareta* clade species. 

(A) Broad distributions of the *H. melpomene*, *H. cydno*, and *H. timareta* colour pattern races and species (based on all known sampling localities; for details, see S2 Fig). Distribution colours match the shadings around the phylogeny and butterfly images in panel B. The dashed line indicates the Andes. Note the distinct clusters formed by individuals sampled from the *H. m. vicina* population. The cluster grouping with *H. timareta* is referred to as *H. timareta* ssp. nov. (Colombia) 

(B) FastTree cladogram inferred using capture sequence from putatively neutral loci. Colours in the tree indicate the *H. melpomene* (pink), *H. cydno* (green), and *H. timareta* (blue) clades and match the boxes of the distribution maps in panel A. 

(C) Sequence information was obtained for 4 putatively neutral regions (green) and 4 regions to which functional
the dennis-ray pattern, sampled from the *H. melpomene vicina* population in the Colombian Amazon (Fig 1A and 1C), consistently clustered within *H. timareta*. This suggests the presence of a lowland population of *H. timareta* considerably further from the Andes than has been detected previously, hereafter referred to as *H. timareta* ssp. nov. (Colombia).

To assess demographic events, which may affect selection tests, we estimated effective population size across time for all populations with whole-genome data (S1 and S3 Tables). In line with previous studies [51,70], we found that bottlenecks were rare across those populations with the exception of a recent decline in population size in *H. heurippa* and older, moderate dips in *H. besckei* and *H. m. nanna* (S3 Fig).

Signatures and limits of detection of classic sweeps assessed by simulations

We used forward-in-time simulations to investigate differences in the signals produced by classic as compared to introgressed selective sweeps in genome scan data, which have been relatively unexplored [27]. Our simulation results are intended to demonstrate qualitative patterns, but we also parameterise the simulations according to the *Heliconius* populations. This allows us to assess the time period over which sweeps can be detected in real data and place bounds on the timing of selection in natural populations. In our analysis, we primarily use SweepFinder2 (SF2), which is appropriate for our genomic data because it is able to identify the sweep site. This method is also robust to demographic processes [73,74], because these are incorporated in the null model used by SF2 (for more details, see Materials and methods). However, to more qualitatively explore patterns of diversity at sites undergoing selection, we here also present results for Tajima’s *D*.

The time over which we can expect to detect sweep signals is determined by the time to coalescence and is thus determined by *N*, the (effective) population size. We therefore here report time since the sweep in generations, scaled by 4*N* [75]. Sweep signals are expected to decay rapidly because of the joint effects of mutation, recombination, and drift. Indeed, SF2, which uses the predicted effect of a selective sweep on the local site frequency spectrum (SFS) to infer the probability and location of sweeps [73,74,76], has low power to detect even hard selective sweeps that occurred over 0.25 (scaled) generations ago and cannot localise sweeps older than 0.4 (scaled) generations [74]. Consequently, any detected sweep signals in *H. melpomene* are likely under 0.8 Mya, assuming an effective population size of 2 million [70,77] and a generation time of 3 months [78]. These estimates vary with *N*, so the time limit for sweep detection varies among species, from only 0.2 Mya for *H. besckei* (*N* approximately 0.5 million) to 1.4 Mya for *H. erato* (*N* approximately 3.5 million). We used simulations to further interpret the empirical signatures of selection and explore the limits of detection (Fig 2).

We initially simulated the case of a hard sweep, such that *s* = 0.5, which is appropriate to the very strong selection pressure experienced by the colour pattern loci in *Heliconius* (Table 1). We found that SF2 signals broke down rapidly after the sweep (Fig 2). The magnitude of the CLR peak decreased by an order of magnitude after just 0.1 scaled generations, corresponding to 0.2 Mya for *H. melpomene*, and was not distinguishable from background values after 0.2 scaled generations, i.e., 0.4 Mya in *H. melpomene* (Welch *t* test, *p* = 0.065). Similarly, the estimated strength of selection calculated with SF2 from our simulations declined rapidly with time. Although the magnitude of the SF2 peak is affected, we find that the time for which we
Fig 2. SFS signatures of selection for simulated classic hard sweeps (left) and introgressed sweeps (right). (A) CLR statistics (upper panel, [73,74]) and Tajima’s D (lower panel) across a simulated chromosome for different time points (0.01, 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, and 1 in units of scaled generations, i.e., 4N generations) after a classic hard (left) or introgressed (right) sweep (effective migration rate $M = 0.2$). The sweep occurs in the centre of the simulated chromosome. Different colours indicate time since sweep. Full, dashed, and dotted vertical black lines in the lower panel indicate positions at different distances from the sweep centre for which time series of CLR and Tajima’s D statistics are depicted in panel B in the same style. (B) CLR (upper panel) and Tajima’s D (lower panel) statistics over time at 3 positions relative to the sweep centre as shown in panel A. Also shown are neutral background values, BG, calculated over neutral simulations, either without migration (left hand panels, for classic sweeps) or with migration at $M = 0.2$ (right hand panels, for introgressed sweeps). Time is given in units of scaled generations. Data are available from https://github.com/markusmoest/SelectionHeliconius.git. BG, background; CLR, composite likelihood ratio.

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can detect selective sweeps does not change if we vary either the strength of selection (using alternative values of $s = 0.1$ and $s = 0.25$), or the mutation rate, which was scaled up such that levels of neutral diversity in our simulations are equivalent to those seen in our Heliconius populations (S4 Fig and S4 Table). Levels of linkage disequilibrium were in the range of the empirical data for all simulated scenarios (S4 and S15–S18 Tables).

Signatures and limits of detection of introgressed sweeps assessed by simulations

We extended our simulations to explore the expected SFS signature left by an allele undergoing adaptive introgression, by simulating a second population which exchanged migrants with the first, leading to an introgressed sweep in the second population. Adaptive introgression produces a highly distinctive SFS signature. At and very close to the selected site itself there was a reduction in diversity and an excess of rare alleles, similar to the pattern observed for a classic sweep. However, this reduction was narrow and flanked by broad genomic regions with high diversity and an excess of intermediate frequency variants. This is due to variants that have hitchhiked into the recipient population along with the beneficial variant and subsequently recombined before reaching fixation [20,27]. The overall SFS signature covered a considerably wider genomic area than that of a classic sweep (Fig 2).

The introgression signature we observe at the sweep site itself was very similar to that for a classical sweep, and we could detect it for a similar length of time. SF2 managed to detect introgressed sweeps, although it detected only the central region of lowered diversity, producing a high but very narrow CLR peak at the sweep site itself; this contrasts with the peaks for classic selective sweeps, which extended over a wider genomic area (Fig 2). The distribution of CLR values at the sweep site was significantly different from values calculated over neutral regions for up to 0.1 generations after the sweep ($p = 0.0041$). However, as for a classical sweep, the magnitude of the peak decreased rapidly.

In the simulations described above, we used an effective migration rate of $M = 0.2$. Estimates of $M$ between hybridising Heliconius species vary from 0.08 to 10 migrants per generation [47–49], and so we also explored a broad range of values of $M$, from 0.02 to 200, in order to cover the estimated range for Heliconius (S5 Fig). We find that the the reduction of diversity at the introgression site itself is strongly affected by migration rate. As $M$ increases, the central reduction in diversity becomes less pronounced, representing an increasingly 'soft' introgressed sweep (S5 Fig) [21,79]. Therefore, detecting introgressed sweeps from this central region will be difficult in populations in which $M$ is high. However, for values of $M$ below 2, varying $M$ had little effect on the regions of increased diversity and excess of intermediate frequency variants that flank the sweep locus (S5 Fig).

Strong signatures of selection across Heliconius colour pattern regions

In our empirical data, SF2 found strong support for positive selection acting across multiple populations and species for all 4 colour pattern loci (Fig 3). In contrast, our background regions, as well as regions flanking the colour pattern associated loci, showed little evidence of sweeps, apart from a few isolated examples (S6 Fig).

This is consistent with previous genome-wide selection scans in H. melpomene that detected only a few strong sweep signatures [70]. These results therefore lend support to the long-standing assertion that wing patterning loci are among the most strongly selected loci in the genome and have a distinctive evolutionary history [80], without excluding the potential presence of other local sweeps in the respective populations.
Broadly, signals of selection were stronger and more widespread in regions near cortex and optix and weaker near WntA and aristaless. For example, all 31 populations showed signal above threshold near cortex, 26 near optix, 24 near WntA, albeit less pronounced in most cases, and only 7 near aristaless (Fig 3 and S5–S8 Tables). A similar pattern was reflected in our estimates for strength of selection ($s$) calculated from $\alpha$ estimates (Table 2; see Materials and methods for a detailed description and formula for this calculation) with the highest selection strength at colour pattern loci being $s = 0.141$ for the cortex ($H. m. nanna$), $s = 0.036$ for the optix ($H. m. plesseni$), $s = 0.049$ for the WntA ($H. m. xenoclea$), and $s = 0.01$ ($H. t. florencia$) for the aristaless region ($H. t. florencia$). These patterns are broadly concordant with the expected phenotypic effects of these loci. For example, in $H. cydno$, which has primarily yellow and/or white patterns associated with the cortex region [39,81], significant peaks were mostly found at this locus, whereas in $H. melpomene$, which has red, yellow, and white patterns, strong signals were seen at both cortex and optix regions. Consistently, a lower strength of selection was found for the aristaless region ($s \leq 0.01$), which controls a modification of pale patterns from yellow to white that is putatively less salient to predators [82] and may contain fewer potential targets of selection.

There were also differences seen across the sampled populations. Widely distributed colour patterns (e.g., $H. m. melpomene$ and $H. m. malleti$) tended to show only modest evidence for selective sweeps (Figs 3 and S11). Comparisons with our simulated data nonetheless suggest selective events that occurred no more than 400,000 years ago. Although there was no significant general correlation between distributional ranges of populations and evidence for selection (S12 and S13 Figs), the strongest signatures of selection were found in geographically localised patterns and likely reflect sweeps within the last 100,000 years (Fig 3 and Table 2). For example, $H. m. plesseni$ is exclusively found in the upper Pastaza valley in Ecuador and shows a unique split red-white forewing band (Figs 1 and 4). This population showed strong selection at 3 colour pattern regions—optix, cortex, and WntA—suggesting recent selection acting on the entire pattern ($s_{\text{cortex}} = 0.074$, $s_{\text{WntA}} = 0.035$, and $s_{\text{optix}} = 0.035$), and patterns of both nucleotide diversity and Tajima’s $D$ are consistent with strong classic sweeps (Figs 3, 4 and S11 and S5 Table). $H. m. xenoclea$, also found on the eastern slopes of the Andes but further south in Peru, shows the same split forewing band associated with the WntA region and again a very strong selection signal at this locus ($s_{\text{WntA}} = 0.049$), as well as weaker signatures at cortex ($s_{\text{cortex}} = 0.04$) and optix ($s_{\text{optix}} = 0.022$; Figs 3 and S11 and S5 Table). The clear signatures of recent and strong selection pressure perhaps indicate that the split forewing band is a novel and highly salient signal. Additionally, $H. m. meriana$ from the Guiana shield revealed a striking signature of selection at optix ($s_{\text{optix}} = 0.023$). Its dennis-only pattern (Fig 4) has previously been shown to have arisen through recombination between adjacent dennis and ray regulatory modules at optix, and the signature of selection at this locus, which encompasses both of these regulatory modules, implies a recent sweep of this recombinant allele [36] (Figs 3, 4 and S11 and S5 Table).
In light of our simulations of introgressed sweeps, there were cases in our data in which previously well-documented adaptive introgression events showed signatures characteristic of introgressed sweeps. The hindwing yellow bar pattern was shown to have introgressed from *H. melpomene* into *H. c. weymeri* and then back again into the races *H. m. vulcanus* and *H. m. cythera* [39]. Accordingly, we found narrow SF2 peaks and an increase in Tajima’s *D* at surrounding sites at these modules in the cortex region in *H. m. cythera*, *H. m. vulcanus*, and *H. c. weymeri*, consistent with introgressed sweeps (Fig 3 and S11 Fig). *H. c. weymeri* also had a second, striking signature further upstream more typical of a classic sweep (Figs 3 and 4), at a region associated with the yellow forewing band in *H. melpomene* and *H. timareta* [30]. This is consistent with evidence for a role of cortex in controlling the white forewing band in *H. cydno* [81] and the presence of this band in the *weymeri* morph, which could therefore represent a recent evolutionary innovation. Other loci previously implicated as having introgressed include the optix region in *H. heurippa* and *H. elevatus*, which both showed signals coinciding with regions previously associated with the respective phenotypes [36,37]. In contrast, there was a lack of clear introgressed sweep signals in dennis-ray *H. timareta*, which is one of the best documented examples of introgression. This could be explained by the age of the sweeps and/or high rates of migration, which our simulations show can reduce the sweep signal in the recipient population (S5 Fig). We also performed scans with VolcanoFinder, a new method designed to detect SFS signatures created by introgressed sweeps [27]. Similar to SF2, VolcanoFinder detected strong signatures of selection in colour pattern regions in the respective populations but not in the neutral background regions (S14–S16 and S19 Figs). However, the estimated divergence values (*D*) did not allow for a clear distinction of introgressed from classic sweeps in our data.

### Table 2. Position, CLR statistics, and estimates for strength of selection (α, 2N_e s, and s) for populations and sweeps discussed in detail.

Annotated colour pattern genes and CREs that overlap with peaks are given. Positions are given in Hmel2 scaffold coordinates (see S5 and S7 Tables).

| Population           | Colour pattern region | Position | CLR   | α     | 2N_e s | s     | Annotated colour pattern gene or CRE |
|----------------------|-----------------------|----------|-------|-------|--------|-------|-------------------------------------|
| *H. m. plesseni*     | WntA                  | 1829355  | 1098  | 6.3   | 95215  | 0.035 | WntA gene, 1. exon*                  |
| *H. m. xenoclea*     | WntA                  | 1811430  | 971   | 4.54  | 118013 | 0.049 | WntA exon, 1. exon*                  |
| *H. c. weymeri f. weymeri* | cortex             | 1337975  | 2411  | 5.3   | 115568 | 0.065 | next to UTR4 of cortex gene^b        |
|                      | cortex                | 1218021  | 367   | 20.74 | 29538  | 0.017 | cortex gene, ventral Yb^c            |
| *H. m. meriana*      | optix                 | 801534   | 1250  | 9.45  | 35360  | 0.023 | dennis CRE^d                         |
| *H. m. plesseni*     | optix                 | 643924   | 2174  | 6.07  | 48223  | 0.035 | upstream of optix                    |
|                      | optix                 | 732278   | 1638  | 6.21  | 47109  | 0.034 | band CRE1^e                         |
|                      | optix                 | 783431   | 2371  | 6.97  | 41978  | 0.03  | band CRE2^e                         |
| *H. m. xenoclea*     | optix                 | 727532   | 1182  | 9.74  | 37910  | 0.022 | band CRE1^e                         |
| *H. e. notabilis*    | WntA                  | 4648024  | 909   | 14.09 | 66925  | 0.011 | Sd region^f                         |
| *H. e. notabilis*    | cortex                | 2497650  | 1387  | 15.2  | 93112  | 0.015 | WAS homologue 1^b                   |
|                      | cortex                | 1963287  | 472   | 49.76 | 28438  | 0.005 | Cr1^f                               |
| *H. e. demophoon*    | cortex                | 2277009  | 1050  | 13.99 | 103964 | 0.016 | Cr2^f                               |
| *H. e. notabilis*    | optix                 | 1294528  | 4690  | 3.03  | 370210 | 0.059 | optix gene and CREs^f               |

^aMazo-Vargas and colleagues [83]

^bNadeau and colleagues [30]

^cEnciso-Romero and colleagues [39]

^dWallbank and colleagues [36]

^eHanly [37]

^fVan Belleghem and colleagues [38]

**Abbreviations:** CLR, composite likelihood ratio; CRE, cis-regulatory element

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Novel targets of selection in colour pattern regions

Many of the signals of selection we detected overlap with previously identified regulatory regions associated with colour pattern variation. However, our analysis also found additional nearby regions showing consistent signals of selection that may also be involved in colour pattern evolution (Figs 3 and S17). For example, in the first intron of the WntA gene, we found a consistent signal across several *H. melpomene*, *H. timareta*, and *H. cydno* populations (S17B Fig). Within this region (Hmel210004:1806000–1833000), phylogenetic clustering of the 2 split forewing band races *H. m. plesseni* and *H. m. xenoclea* indicates a common origin of the split band in these currently disjunct populations (S7 Fig). Additionally, 2 strong selection signatures are frequently found in a region approximately 200 kb upstream of WntA (S17B Fig; Fig 4. Selected examples of sweeps. The 3 examples show the split forewing band (WntA region) in *H. m. plesseni*, the yellow and white patterns (cortex region) in *H. cydno weyneri* *f. weyneri* and the red dennis patch (optix region) in *H. m. meriana* (left to right). The respective colour pattern elements are indicated with red and grey arrows. Colour patterns and gene annotations in the colour pattern regions are depicted in the top panel. Colour pattern genes are annotated in red. Nucleotide diversity \( \pi \), Tajima's D, and SweepFinder2's CLR statistics (peaks capped at 1,000) show the signatures of a selective sweep (bottom panels). Loess smoother lines are depicted in yellow. The colour gradient in the CLR panel indicates the estimated intensity of selection \( \alpha \) [73] (black = high \( \alpha \) values, weak selection; red = low \( \alpha \) values, strong selection). Grey shadings indicate annotated CREs, and red and grey arrows depict associations with the respective colour pattern elements in the *H. melpomene* clade. Data are available from https://github.com/markusmoest/SelectionHeliconius.git. CLR, composite likelihood ratio; CRE, cis-regulatory element. 

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He, cortex to be more complex and differs more strongly between the 2 clades (Figs 5, S17 and S18). species at the eral co-mimetic races suggests a rather simple and concordant regulatory architecture in the 2 comparison of the location of selection peaks between tions and 3 species in the erato. Homologous genes control corresponding phenotypes [30,35,95,96], but there is no allele whether one species evolved towards diverse phenotypes of the other, i.e., advergence [67,92–94]. Parallel selective sweep signatures between mimetic species There has been considerable interest in whether the H. erato and H. melpomene co-mimics have co-diverged and simultaneously converged onto the same colour pattern [88–91] or whether one species evolved towards diverse phenotypes of the other, i.e., advergence [67,92–94]. Homologous genes control corresponding phenotypes [30,35,95,96], but there is no allele sharing between the melpomene and erato clade [67,68]. We used published genomic data for H. erato (Van Belleghem and colleagues, 2017; S10 Table) to obtain 8.9 Mb of sequence homologous to the regions studied in the H. melpomene clade for 103 individuals from 13 populations and 3 species in the H. erato radiation and scanned for selective sweeps. Generally, a comparison of the location of selection peaks between H. melpomene and H. erato across several co-mimetic races suggests a rather simple and concordant regulatory architecture in the 2 species at the WntA locus. However, in the cortex and optix regions, this architecture appears to be more complex and differs more strongly between the 2 clades (Figs 5, S17 and S18).

Similar to the melpomene clade radiation, we found strong signatures of selection across the optix, cortex, and WntA regions (Figs 5 and S20–S22 and Tables 2 and S11–S14). Most notably, H. e. notabilis from Ecuador showed strong signals of selection at 3 colour pattern loci ($s_{\text{optix}} = 0.06$, $s_{\text{cortex}} = 0.015$, $s_{\text{WntA}} = 0.015$) similar to its co-mimic H. m. plesseni (Table 2). In both cases, selection across the 3 major loci represented some of the strongest signals in both species. Additionally, H. e. amalfreda, co-mimic with the red dennis-only race H. m. meriana, showed one of the strongest selection signals at optix. This suggests that these phenotypes are recent innovations in both species, consistent with co-divergence. Other geographically localised variants controlled by WntA also showed strong signals of selection, indicating a recent origin. For example, H. e. etylus, like H. m. ecuadoriensis, has a restricted forewing band shape that corresponds to the more distal element of the notabilis forewing band ($s_{\text{WntA}} = 0.015$). Clear, narrow, and very similar selection signals were found near WntA in H. e. amalfreda and H. e. erato ($s_{\text{WntA}} = 0.006$ in each), both with a broken forewing band, as well as H. e. emma ($s_{\text{WntA}} = 0.003$) and H. e. lativitta ($s_{\text{WntA}} = 0.004$), both with a narrow forewing band (S11 Table).

More broadly across the H. erato populations, there was a clear difference between the Amazonian dennis-ray races (i.e., H. e. amalfreda, H. e. erato, H. e. emma, H. e. etylus, and H.
Fig 5. Signatures of selection in the co-mimic populations of H. melpomene (upper panels) and H. erato (lower panels). The regions containing WntA, cortex, and optix are shown (left to right). Co-mimics in H. melpomene and H. erato are depicted in the same order with phenotypes on the left. The y-axis indicates CLR statistics across each region (capped at 1,000). The colour gradient indicates the estimated intensity of selection (r2) [73] (black = high r2 values, weak selection; red = low r2 values, strong selection). Grey shadings indicate annotated colour pattern CREs [30,36,37,39] (S7–S10 Figs) and blue horizontal bars indicate regions with CLR statistics above threshold. The central panel shows an alignment of the respective regions in H. melpomene and H. erato and gene annotations with colour pattern genes in red. Top and bottom panel show colour pattern phenotypes, and symbols indicate distinct colour pattern elements and their presence in each population panel. Note that the yellow hindwing bar controlled by the cortex region can be expressed on the dorsal and ventral side (yellow/yellow square symbol) or on the ventral side only (black/yellow square symbol) [39]. Full, grey lines connect colour pattern elements with annotated CREs [30,36,37,39] (S7–S10 Figs) and blue horizontal shadings indicate annotated colour pattern CREs [30,36,37,39] (S7–S10 Figs) and blue horizontal shadings indicate annotated colour pattern CREs (S7–S10 Figs) and blue horizontal shadings indicate annotated colour pattern CREs (S7–S10 Figs). The sweep signal therefore shows little obvious association with any particular wing pattern phenotype but may still indicate a locus involved in the colour pattern pathway. In addition, we detected very distinct signals between H. e. favorinus (Cr1) and H. e. demophoon (Cr2) consistent with previous studies [30,38,98] that found evidence for independent evolution of the yellow hindwing bar on either side of the Andes. Although H. e. favorinus lacks any signature at Cr2 and shows a weak signal at Cr1, a clear peak was found for H. e. demophoon at Cr2 indicating that this allele may be more recent (Figs 5, S18 and S20).

Discussion

Elucidating the evolutionary history and spread of advantageous variants in natural populations lies at the heart of evolutionary research, ever since Wallace [99] and Darwin [100] established the theory of evolution by natural selection. However, detecting and quantifying selection has been a challenge, particularly in wild populations [3]. We have combined a large data set of high coverage genomic data with novel theoretical analyses to identify molecular signatures of recent selection at genes known to control adaptive wing patterning traits in Heliconius butterflies. We demonstrate that these strongly selected loci have been subject to recent bouts of natural selection even within the last 100,000 years, with geography and phenotype standing out as strong predictors of selection (Fig 6).

Many studies have used naive genome scans to identify selection in natural populations, but such an approach can lead to false positives [101]. More integrative approaches, which
combine selection scans with information on phenotypic selection in the wild and genetic trait mapping, can give a more complete picture of how selection shapes specific loci and phenotypes [10,12,14,16,102]. Such studies are increasingly common but with few exceptions focus on a single locus, or a limited set of populations or phenotypes, often because of the high sampling and sequencing effort required. We take advantage of 150 years of Heliconius research, including field selection experiments, hybrid zone studies, detailed dissection of the genetics of colour pattern elements, and introgression studies, to survey genomic signatures of selective sweeps across many populations and loci. With our study design, we reconcile large geographic sampling and high-coverage sequence data by targeting well-defined regions in the genome. This combination of ‘top-down’ and ‘bottom-up’ approaches, as defined by Linnen and Hoekstra [1], reveals pervasive evidence for the action of natural selection on mimicry loci in an adaptive radiation associated with a great diversity of phenotypes.

We have shown a pervasive pattern of strong selection acting on mimicry colour patterns, which contrasts strongly with the regions flanking the selected loci and neutral background genome regions. This supports the assertion of ‘contrasted modes of evolution in the genome’, first formulated by John R. G. Turner 40 years ago [80], who concluded that mimicry genes and neutral parts of the genome were subject to different modes of evolution. Of course, our data do not preclude the existence of other strongly selected loci not associated with mimicry in the genome. The frequency of evidence for selection is consistent with the large effective population sizes in Heliconius that preserve the signature of selective sweeps over a relatively long period of time. Our estimates of selection strength indicate strong selection acting on mimicry genotypes, which is in line with field and hybrid zone studies on the colour pattern phenotypes (Tables 1, S6 and S11) and strong selection on colour polymorphisms in other species [1,10,103]. Heliconius butterflies therefore join a small group of systems for which strong natural selection on ecologically important traits has been documented in detail at both the phenotypic and molecular level [1,2]. Other examples include Darwin’s finches, in which climate-driven changes in seed size and hardness imposed strong selection on beak size and body weight [15,104,105], industrial melanism in the peppered moth Biston betularia [103,106], the body armour locus Eda in sticklebacks [107], and crypsis in Peromyscus maniculatus deer mice controlled by the Agouti pigment locus [16].

However, both strength and direction of selection can vary substantially in time and space, and a snapshot of a single population may be misleading about the action of selection in the wild [105,107–109]. One way to account for this variation is by studying patterns of selection in geographically widespread adaptive radiations, comprising ecological replicates. This approach allows us to describe general patterns in the action of selection on a continental scale. For example, there is consistently stronger selection on the optix and cortex loci across the range of these species, consistent with the greater phenotypic effect of alleles at these loci. In addition, we also identify what seem to be more recent phenotypes showing a stronger signature of selection, such as the split band phenotype in the Andes and the dennis-only phenotype on the Guiana shield (Fig 6).

One of the defining characteristics of the Heliconius radiation has been the importance of adaptive introgression and recombination of pre-existing variants in generating novelty [36,39,44]. We used simulations to explore the expected patterns resulting from both new mutations and introgressed selective sweeps. These demonstrated a distinct signature of selection on introgressed variation, consistent with recent theory [27], and revealed that depending on the frequency of the acquired variant, introgressed sweeps show a range of characteristics reminiscent of classic sweeps. Consistently, we found that tests designed for detecting classic sweeps can also detect introgressed sweeps, but the signal becomes narrower, and the time window for detection decreases. In addition, the power to detect selection decreases with
increasing effective migration rate between hybridising species. These conclusions may explain the scarcity of selection signatures in the *H. timareta* populations that represent well-documented recipients of adaptive introgression but also show strong genome-wide admixture, suggesting relatively high migration rates with *H. melpomene* [36,44,72]. Nonetheless, we detected putative introgressed sweeps in *H. c. weyni*, *H. m. cythera*, *H. m. vulcanus*, and *H. heurippa*, for which acquisition of colour pattern phenotypes via adaptive introgression has been demonstrated and introgressed genomic intervals were identified [39,87,110]. We also attempted to implement a new method for detecting introgressed sweeps directly (VolcanoFinder), but although this method detected signatures of selection [S14–S16 and S19 Figs], it did not strongly differentiate classic and introgressed sweeps in our data [27]. The signatures were broader but largely congruent with the SF2 results. Although VolcanoFinder found strong signals for most *H. timareta* populations as well as the cortex region in *H. cydno* and *H. melpomene* populations West of the Andes, the estimated divergence values were inconclusive,
most likely a consequence of low divergence between donor and recipient, ongoing admixture, and a complex history of selective events in our particular system. Therefore, combining prior phylogenetic evidence for introgression with scans for selection is likely to remain a powerful means to study adaptive introgression [111,112].

Our results imply a complex history in which multiple bouts of selection have occurred at the same loci. Although recurrent sweeps can alter or even eradicate previous signatures [5], there is nonetheless evidence for sweeps, both at previously characterised genomic regions and in novel locations. Previously, regulatory loci have been identified based on association studies across divergent populations [36,39,38], and many of these regions indeed show strong signatures of selection providing further support for their functional roles. However, consistent signatures of selection are also found at nearby loci, suggesting additional targets of selection, some of which had not previously been identified using top-down approaches. Some caution is required, because the signatures of selective sweeps are notoriously stochastic and can be misleading in their precise localisation because of linkage. Nonetheless, there are consistent patterns across multiple populations suggesting additional targets of selection that may represent regulatory elements affecting already characterised genes [36,39], similar to multiple mutations under selection at the Agouti gene in deer mice (Peromyscus maniculatus) [10]. In addition, however, some of these signals may represent selection at linked genes, and the architecture of colour pattern in Heliconius may be comparable to the situation in Antirrhinum snapdragons in which loci encoding flower pattern differences, i.e., ROSEA and ELUTA, are in tight linkage.[12]. Further functional studies will be required to unravel the roles of these loci, but theory suggests that physical linkage between genes contributing to the same adaptive trait can be favoured [113,12]. Intriguingly, Heliconius butterflies show both unlinked colour pattern loci, as well as tightly linked CREs and genes within loci, putatively preserving locally adaptive allelic combinations. It is conceivable that this architecture provides a high degree of flexibility that has facilitated the radiation of colour patterns in Heliconius.

Müllerian mimics can exert mutual selection pressures, offering the rare opportunity to study replicated selection in a co-evolutionary context. The diversity of mimicry alleles between H. melpomene and H. erato evolved independently [67,68], but several co-mimics between the 2 radiations show signatures of selection in homologous colour pattern regions, demonstrating repeated action of natural selection between co-mimics over recent time. Our findings also contribute to long-standing arguments on the origin and spread of the colour patterns [67,88–94]. Signatures of selection at the optix locus, particularly in H. erato, are consistent with the hypothesis that the red forewing band is ancestral and the dennis-ray is a younger innovation that spread through the Amazon. However, in contrast to this ‘recent Amazon’ hypothesis, we find the strongest signatures of selection in some of the unique and geographically restricted phenotypes found in Andean populations suggesting novel colour patterns have experienced strong recent selection in both species, consistent with co-divergence and ongoing co-evolution (Fig 6). The most striking examples are H. e. notabilis and H. m. plesseni, which show imperfect mimicry (see Fig 5) and are possibly still evolving towards an adaptive optimum. In summary, our results provide evidence for co-divergence and the potential for co-evolution in the sense of mutual evolutionary convergence [93] but do not rule out advergence in other cases.

To conclude, understanding the adaptive process that creates biodiversity requires knowledge of the phenotypes under selection, of their underlying genetic basis, and estimates of phenotypic and genotypic strength and timing of selection [1]. Although decades of Heliconius research have resulted in a detailed understanding of most of these levels, our study fills a gap by providing estimates of the distribution and strength of genotypic selection across 2 radiations and dozens of populations. However, our results not only highlight the complexity of
mimicry selection across the *Heliconius* radiation but also reveal a surprisingly dynamic turnover in colour pattern evolution, in particular in geographically peripheral patterns (Fig 6). This is in stark contrast to the predicted evolutionary inertia of mimicry patterns due to strong stabilizing selection pressure exerted by mimicry selection [53]. We provide evidence that colour patterns are actively evolving under both classic and introgressed sweeps. Many of the detected sweep signatures are considerably younger than estimates of the age of colour pattern alleles based on phylogenetic patterns [36,39], suggesting ongoing improvement, innovation, and local switching between combinations of pattern elements. This is also consistent with observations of phenotypically distinct colour patterns restricted to the 5,000-year-old islands Ilha de Marajó in the South of Brazil and a few documented cases of rapid, local colour pattern turnover [114]. Therefore, our study offers a new perspective to the long-standing discussion of the paradox: How and why do new colour patterns arise? More generally, we here demonstrate that by considering selection across populations and species of an entire radiation, comparative information can capture spatial and temporal variability of genotypic selection and help to gain a more comprehensive understanding of the dynamics of adaptation in the wild.

**Methods**

**Ethics statement**

Panamanian specimens were collected under permit SE/AP-14-18 issued by the Ministerio de Ambiente de Panamá. Samples from Ecuador were collected with permission of the Ministerio del Ambiente under permits number 006-2012-IC-FAU-DPL-MA, 002-16-IC_FLO_FAU_DNB/MA, 033-10-IC_FAU/FLO_DPN/MA, and 0007-IC-FAU/FLO-DPPZ/MA. Colombian specimens were collected under the permit IDB0199/No16 and permit 530 granted to Universidad del Rosario by the Autoridad Nacional de Licencias Ambientales (ANLA-Colombia). Samples from Peru were collected under permit N°0148-2011-AG-DGFFS-DGEFFS and N°0236-2012-AG-DGFFS-DGEFFS from the Ministerio de la Agricultura, Peru. Samples from Suriname were collected and exported under a permit (No. 10865) from the Nature Conservation Division of the Suriname Forest Service. Field collections in Brazil were made under IBAMA/ICMBio license number 2024629 granted to GRPM. Recommendations of Animal Care and Use Committee (CEUA) of the Federal University of Rio Grande do Sul (UFRGS) were followed during laboratory procedures, including DNA extractions.

**Sampling and DNA extraction**

Our sampling covers most of the distribution and colour pattern variation of the *Heliconius* radiation in South and Central America. Specimens were sampled or provided by collaborators with the respective sampling permissions and stored in salt saturated DMSO or ethanol at −20°C until further processing. For DNA extractions, thorax muscle tissue was dissected, disrupted and digested, and DNA was extracted using a TissueLyser II bead mill together with the DNeasy Blood and Tissue Kit (Qiagen, Hilden, Germany) following supplier recommendations.

**Targeted capture and sequencing**

For hybridisation-based target enrichment a NimbleGen SeqCap EZ Library SR capture probes library was designed and synthesized by the provider (Roche NimbleGen Inc, United States). The templates for designing probes for 4 colour pattern regions (approximately 3.2 Mb) and 4 genomic background regions (approximately 2 Mb) were assembled and curated using the *H. melpomene* genome assembly Hmel1 [44], available BAC walks [31,115], fosmid
data [69], and alignments from Wallbank and colleagues [36]. The neutral background regions were chosen to represent the average genome. We therefore excluded regions with extended stretches of extreme values for diversity and/or divergence, and we only considered regions located on a single, well-assembled scaffold.

Sample DNA was sheared with an ultrasonicator (Covaris Inc, Massachusetts, United States) and adapter-ligated libraries with insert sizes of 200 to 250 bp were generated using the Custom NEXTflex-96 Pre-Capture Combo Kit (Bioo Scientific Corporation, United States). For sequence capture, 24 libraries each were pooled into a capture library, hybridised with blocking oligos and the biotinylated capture library probes, and subsequently captured with streptavidin-coated magnetic capture beads using the NimbleGen SeqCap EZ Kits (Roche NimbleGen Inc, Wisconsin, United States). After capture and clean-up, 3 capture library pools were combined each. For the resulting sequencing pools of 72 samples, Illumina 100 or 150 bp paired-end short read data were generated on Illumina’s HiSeq 2000 (BGI, China) and HiSeq 4000 (Novogene Co. Ltd, China), respectively (S1 Table).

Whole-genome data

Whole-genome resequencing data available for the *melpomene* clade from previously published work were also included [30,39,42,44,45,51,70–72]. For a few additional samples, 100 to 150 bp paired-end whole-genome resequencing data were generated on an Illumina X Ten platform (Novogene Co. Ltd, China; S1 Table). In addition, we downloaded, processed, and analysed a publically available data set for *H. cydno galanthus* [49] with a more moderate depth of coverage (for results see S14 Fig). For the *erato* clade already published, whole-genome-resequencing data were used [38] (S10 Table).

The whole-genome data were mainly used for demographic reconstructions, whereas, for other analyses, the regions matching the capture regions were used.

Genotyping

For *melpomene* clade data, sequenced reads were aligned to the *H. melpomene* version 2 reference genome (Hmel2, [137]), using BWA-mem version 0.7 [116]. PCR duplicated reads were removed using Picard version 2.2.4 (http://picard.sourceforge.net), and reads were sorted using SAMtools version 1.3.1 [117]. Genotypes for variant and invariant sites were called using the Genome Analysis Tool Kit’s (GATK) Haplotypecaller version 3.5 [118]. Individual genomic VCF records (gVCF) were jointly genotyped per population using GATK’s genotypingVCFs version 3.5 [118]. Genotype calls were only considered in downstream analyses if they had a minimum depth (DP) ≥ 10, and for variant calls, a minimum genotype quality (GQ) ≥ 30, and indels were removed. Filtering was done with bcftools version 1.4 [117], and for downstream calculations of summary statistics and creating SF2 input, VCF files were parsed into tab delimited genotype files (scripts available at https://github.com/simonhmartin). For the *erato* clade, read data were mapped to the *H. erato demophoon* version 1 genome reference [38] and further processed as described above.

Phasing

SHAPEIT2 [119] was used to phase haplotypes using both population information and paired read information. First, monomorphic and biallelic sites were filtered with GQ ≥ 30 and DP ≥ 10, and sites with less than 20% of sample genotypes were removed.

Next, phase informative reads (PIRs) with a minimum base-quality and read quality of 20 were extracted from individual BAM files using the extractPIRs tool. These BAM files were
obtained from BWA-mem [116] mappings to the *H. melpomene* version 2 genome, with duplicates removed.

Finally, SHAPEIT2 was run with PIR information and default parameters on each scaffold using samples from single populations, which resulted in a haplotype file that was transformed into VCF format. Sites with no genotype information were imputed.

**Phylogenetic reconstruction**

FastTree2 [120] was run using default parameters to infer approximate maximum likelihood phylogenies. Separate phylogenies for a concatenated SNP data set comprising neutral background regions only and for the full data set including the color pattern regions to account for the effect of including regions putatively under strong selection were produced.

**Population historical demography**

Changes in the historical population size were inferred from individual consensus whole-genome sequences (S3 Table) using Pairwise Sequentially Markovian Coalescent (PSMC') analyses as implemented in MSMC [121]. This method fits a model of changing population size by estimating the distribution of times to the most recent common ancestor along diploid genomes. When used on single diploid genomes, this method is similar to PSMC analyses [122]. Genotypes were inferred from BWA version 0.7 [116] mapped reads separately from previous genotyping analysis using SAMtools version 0.1.19 [117]. This involved a minimum mapping (-q) and base (-Q) quality of 20 and adjustment of mapping quality (-C) 50. A mask file was generated for regions of the genome with a minimum coverage depth of 30x and was provided together with heterozygosity calls to the MSMC tool. MSMC was run on heterozygosity calls from all contiguous scaffolds longer than 500 kb, excluding scaffolds on the Z chromosome. We scaled the PSMC' estimates using a generation time of 0.25 years and a mutation rate of 2×10^{-9} estimated for *H. melpomene* [47,77].

**SLiM simulations**

Simulations were conducted to compare the genomic signatures of classical selective sweeps and sweeps that occur via adaptive introgression using SLiM (version 2) forward-in-time population simulation software [123,124]. Because SLiM tracks mutations and individuals through time, we were able to track individual beneficial alleles going to fixation and post sweep; however, it is computationally intractable to simulate very large populations with SLiM, and so we instead simulated smaller populations and rescaled population genetic parameters, N and μ, such that our results are applicable to *Heliconius* (as is commonly done [124,125]). Two populations of N = 1,000 were simulated with a neutral mutation rate μ of 6×10^{-7} such that the expected level of neutral diversity in the population was 0.0024, which is within an order of magnitude of that observed in our *Heliconius* populations [38,70] (S15–S18 Tables). Each individual in our simulated populations was represented by a single diploid recombining chromosome (recombination rate was also scaled such that NR is within the values of those observed in *Heliconius*, 4x10^{-7}, or 40 cM/Mb) of length 750,000 bp. We also ran simulations on a shorter length of chromosome (50,000 bp) with a higher value of μ, raising levels of neutral diversity to those observed within *Heliconius*, to ensure our results are consistent for higher values of μ.

Our simulations were first allowed to equilibrate for a burn-in phase of 10N generations, after which we introduced a single strongly advantageous mutation of s = 0.5 in the centre of the chromosome in order to simulate a ‘classical’ hard selective sweep in the population (which we will refer to as p1). We also ran our simulations with 2 lower values of s (s = 0.1 and s = 0.25).
Only those simulations in which the mutation went to fixation were kept; if the beneficial mutation was lost during the course of a simulation, the simulation was reset to a point just after the burn-in phase and the mutation was reintroduced. The simulations were then allowed to run for a further 5\(N\) generations. During this time, \(p1\) does not experience any migration or population size change. In order to simulate an introgressed sweep, we simulated an additional neutrally evolving population, \(p2\), which exchanges migrants with population \(p1\) at a constant rate of 0.0001 migrants per generation, which allowed the beneficial mutation fixed in \(p1\) to introgress into \(p2\). The simulations were then allowed to run for a further 10\(N\) generations with a constant migration rate. For each set of parameters, we ran our simulations 100 times.

For both populations, a complete sample of the segregating neutral mutations was taken every 100 generations after the burn-in phase and prior to the introduction of the beneficial mutation, and every 50 generations after the introduction of the beneficial mutation. We also tracked the change in frequency over time of the beneficial mutation during the simulations. From these results we calculated 2 summary statistics, Tajima’s \(D\) and \(\pi\), in windows of 10,000 bp across our simulated chromosomes for a range of time points. Time points are as follows, in 4\(N\) generations post sweep: 0.01, 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 1, and 2 background rates: one post burn-in, during which populations are not experiencing any migration, and one post sweep, during which the populations are exchanging migrants. Values were then averaged across simulations. Additionally, to model the effect of changing effective migration rates on the introgression sweep signal, we ran simulations with different levels of migration, using the following 4 values of \(M\): 200, 2, 0.2, and 0.02, with recombination rate = 4cM/Mb and \(s = 0.1\). The simulations were otherwise set up as before, with 30 simulation runs generated for each set of parameters.

We used these results to generate SF2 [76] input files, after first subsampling the number of mutations down, such that our simulated SF2 files for each population represent a sample of 500 simulated individuals. This step is necessary because SweepFinder has an upper limit on the number of sequences that can be included per sample [126]. We then ran SF2 using model-lg 100 for each simulation for each of the time points, using 1 of 2 precomputed site frequency spectra as appropriate: one calculated across multiple neutral simulations without migration and one calculated across multiple neutral simulations with migration (these neutral simulations correspond to the 2 background rates described above). Further details of SF2 and its various run modes are included in the ‘SF2’ section.

Phylogenetic weighting

A phylogenetic weighting approach was used to evaluate the support for alternative phylogenetic hypotheses across colour pattern loci using Twisst [127]. Given a tree and a set of predefined groups, in this case Heliconius populations sharing specific colour pattern elements, Twisst determines a weighting for each possible topology describing the relationship of the groups. The weightings thus represent to what extent loci cluster according to phenotype, rather than geographic relatedness of populations. Topology weightings are determined by sampling a single member of each group and identifying the topology matched by the resulting subtree. This process is iterated over a large number of subtrees, and weightings are calculated as the frequency of occurrence of each topology. Weightings were estimated from 1,000 sampling iterations over trees produced by RAxML version 8.0.2681 [128] for 50 SNP windows with a stepping size of 20 SNPs. For phylogenetic weighting along the WntA interval, weightings of topologies that grouped populations with the split forewing band phenotype or, alternatively, the hourglass shape were assessed (S7 Fig). For the region containing the aristaless genes, we focussed on topologies that clustered populations with white or yellow colour
phenotypes (S8 Fig). For the cortex region, we focussed on topologies grouping populations showing the ventral and dorsal yellow hindwing bar, respectively (S9 Fig). Finally, for the optix interval, we assessed topologies grouping populations according to the absence or presence of the red dennis patch, the red hindwing rays, or the red forewing band and repeated the analysis for different geographic settings (S10 Fig). To obtain weightings for hypothesized phylogenetic groupings of specific colour pattern forms, we summed the counts of all topologies that were consistent with the hypothesized grouping.

**Inference of selection and summary statistics in sliding windows**

Summary statistics informative on diversity and selection patterns were calculated. From the unphased data, nucleotide diversity, Kelly’s $Z_{ns}$, Tajima’s $D$, and number of sites genotyped for each population were calculated in 1 kb nonoverlapping sliding windows with at least 100 sites genotyped for at least 75% of all individuals within that population using custom python scripts and the EggLib library version 3[129]. Scans for selection using signals of extended haplotype homozygosity and calculation of the pooled integrated haplotype homozygosity score (iHH12) [11,130] were performed using the program selscan1.2 [131] and our phased data set.

**SF2**

SF2 was then run in 2 modes for each data set: with flag -s, calculating the likelihoods from the SFS of the respective region and with flag -l, using a SFS precalculated either from the background regions only or from background regions and colour pattern regions combined. These precalculated SFSs are used by SF2 as null models that incorporate the underlying demography of the populations of interest, making SF2 sensitive to selective sweeps even in populations that are not at equilibrium [132]. For the melpomene clade, recombination rate information from a fine scale recombination map was included (flag -r) [133]. To create a recombination file, recombination map coordinates were transferred to Hmel2 coordinates, and between sites recombination rates were calculated.

SF2 test runs for different grid spaces (flag–g; tested values: $-g1$, $-g5$, $-g50$, $-g100$, $-g1000$) were performed to find a setting allowing for reasonable runtimes without loss of accuracy and based on these test CLR and $\alpha$ were calculated for every 50th site ($-g50$) across all populations and regions.
Generally, the results were largely consistent among the different runs and data sets. As expected, power to detect sweeps was higher when including substitutions [74], and the minimum number of outgroup samples had only marginal effects. We therefore focussed on the results for data sets with outgroup minimum 1 (−OM1) and background SFS calculated from background regions and background regions and colour pattern regions combined, respectively. Including the colour pattern regions inflates the estimated background SFS with regions affected by selective sweeps which results in slightly lower CLR and higher $\alpha$ estimates. Because selective sweeps across the genome have been found to be rare in *H. melpomene* [70], these estimates represent a lower bound, and the estimates derived with background SFS from the background regions only are most likely a better approximation. Only CLR peaks exceeding a threshold defined as the 99.9th percentile of the distribution of CLR values across all background regions were considered as evidence for selection.

To obtain estimates for strength of selection ($s$) we calculated $s = r \times \ln(2N_e)/\alpha$ [132,134] with region- and population-specific estimates of effective population size ($N_e$) estimated from the data using the mutation rate given in Keightley and colleagues [77] and per chromosome recombination rate estimates ($r$) from Davey and colleagues [133] and Van Belleghem and colleagues [38].

VolcanoFinder

We also tested the new software VolcanoFinder on our data, described in a recent preprint, which is specifically designed to detect introgression sweeps but can also detect classic sweeps [27]. As for the SF2 runs, we used data sets with outgroup minimum 1 (−OM1) and background SFS calculated from background regions to generate the allele frequency files and the required unnormalized site frequency spectrum. We then ran VolcanoFinder with the following specifications: Model 1 and $P = 0$.

Supporting information

S1 Fig. Phylogenetic reconstruction of the *H. melpomene* clade. Phylogenetic reconstruction for *H. melpomene* clade samples used in this study including all sequenced region, i.e., colour pattern regions and neutral background regions. *H. cydno* (green) and *H. timareta* (blue) cluster together and form a sister clade to *H. melpomene* (red). The ‘silvaniforms’ outgroup is shown in orange. A high-resolution version can be found here: https://github.com/markusmoest/SelectionHeliconius.git.

S2 Fig. Distributional ranges as obtained from Rosser and colleagues [136] and samples localities of this study. Colour coding representing populations corresponds to colour coding in Fig 1A in the main text.

S3 Fig. Demographic history of *H. melpomene* clade populations. Demographic histories for populations in the *H. melpomene* clade for which whole-genome data were available reconstructed with PSMC’ [121]. Additional demographic histories for *Heliconius* species considered in this study are already published [38]. PSMC’, Pairwise Sequentially Markovian Coalescent.

S4 Fig. CLR statistic (SF2 [74,76]), over time at 3 positions relative to the sweep centre. Plotted is the CLR statistic over time at 3 chromosome positions relative to the sweep centre, which correspond to the sweep site itself (dark blue), 0.02 Mb from the sweep (mid blue), and
0.04 Mb from the sweep (light blue), for 4 different simulation parameters. Selection coefficient, \( s = 0.25 \), neutral mutation rate, \( \mu = 6e-07 \) corresponds to Fig 2, with average SF2 values calculated over 100 simulation runs, along with their standard errors. We also explored changes in \( s \) and \( \mu \) in our simulations. Averages over 20 simulation runs are shown, along with their standard errors. Time is given in units of scaled generations. CLR, composite likelihood ratio; SF2, SweepFinder2

S5 Fig. Effect of effective migration rate on introgressed sweep signatures. SFS signatures of simulated introgressed sweeps across a chromosome for different time points summarised as Tajima’s \( D \) statistics. The sweep occurs in the centre of the simulated chromosome. Different colours indicate patterns at different time points since sweep (0.01, 0.1, 0.5, 0.8, and 1 scaled generations, i.e., \( 4N \) generations). Simulated data for 4 different effective migration rates are shown (\( M = 200, 2, 0.2, \) and 0.002). SFS, site frequency spectrum

S6 Fig. Signatures of selection across neutral background regions in the \( H. melpomene \) clade. Genes are annotated in the top gene annotation panel. On the \( y \)-axis SF2's \( [74,76] \) CLR statistics is shown (peaks are capped at CLR = 1,000). The colour gradient indicates estimated intensity of selection (black = high \( \alpha \) values, weak selection; red = low \( \alpha \) values, strong selection). Blue horizontal bars indicate regions with CLR values above threshold. CLR, composite likelihood ratio; SF2, SweepFinder2

S7 Fig. Tree weighting (Twiss [127]) analysis of the WntA gene region. Topology weightings for topologies clustering the split forewing band phenotype (magenta) and the hourglass shape phenotype (blue) are shown. (ama = \( H. m. amaryllis \), ecu = \( H. m. ecuadoriensis \), ple = \( H. m. plesseni \), xen = \( H. m. xenoclea \), cyd = \( H. cydnides \), wey = \( H. c. weymeri \, f. weymeri \), gus = \( H. c. weymeri \, f. gustavi \), zel = \( H. c. zelinde \)).

S8 Fig. Tree weighting (Twiss [127]) analysis of the aristaless genes region. Topology weightings for topologies clustering the white (chi = \( H. c. chioneus \), zel = \( H. c. zelinde \)) and yellow (ecu = \( H. m. ecuadoriensis \), ple = \( H. m. plesseni \), heu = \( H. heurippa \), flo = \( H. t. florencia \), cyd = \( H. cydnides \), pac = \( H. pachinus \)) colour phenotypes (magenta) are shown.

S9 Fig. Tree weighting (Twiss [127]) analysis of the cortex gene regions. Topology weightings for topologies clustering the dorsal yellow hindwing bar (magenta) and ventral yellow hindwing bar (blue) phenotypes are shown (crt = \( H. m. cythera \), bur = \( H. m. burchelli \), nan = \( H. m. nanna \), ros = \( H. m. rosina \), vul = \( H. m. vulcanus \), chi = \( H. c. chioneus \), wey = \( H. c. weymeri \, f. weymeri \), gus = \( H. c. weymeri \, f. gustavi \), zel = \( H. c. zelinde \), pac = \( H. pachinus \)).

S10 Fig. Tree weighting (Twiss [127]) analysis of the optix gene regions. Topology weightings for topologies clustering the dennis (magenta), rays (blue), and band (brown) phenotypes. Including different red banded populations shows different phylogenetic clustering and thus potentially a different genetic basis underlying this trait among populations. (A) Tree weighting including the Peruvian red banded population \( H. t. thelsinoe \). (B) Tree weighting including red banded populations from East Brazil, \( H. m. burchelli \), \( H. m. nanna \) and \( H. besckei \).
florencia, lin = H. t. linaresi, the = H. t. thelexinoe, tim = H. t. timareta f. timareta, con = H. t. timareta f. contigua, ele = H. elevatus, bes = H. besckei, silvana = H. numata silvana).

**S11 Fig.** Summary and selection statistics across colour pattern regions for all populations analysed in the H. melpomene clade. For each population genotyping coverage (calculated as proportion of retained genotypes after quality filtering in 500 bp windows), nucleotide diversity, Kelly’s $Z_{ns}$, Tajima’s $D$, pooled integrated haplotype homozygosity score, and SweepFinder2’s [74,76] composite likelihood ratio statistics across each colour pattern region are shown (top to bottom). File names contain population and colour pattern region identifiers (Hmel201011 = aristaless scaffold, Hmel210004 = WntA scaffold, Hmel215006 = cortex scaffold, Hmel218003 = optix scaffold). The 120 single figures have been uploaded to GitHub: https://github.com/markusmoest/SelectionHeliconius/tree/master/S11_Fig_H_melpomene.

**S12 Fig.** Correlation between portion of genomic loci under selection and geographic range of co-mimicking H. melpomene (above) and H. erato (below) races. Portion of genomic loci under selection is summarised as percentage of CLR values across the colour pattern region which are above the CLR threshold [%CLR > th] scaled by the maximum value for WntA, cortex, and optix regions. Areas were calculated from distribution data obtained from [136] using an alpha hull polygon (code available at https://github.com/StevenVB12/Sample-distributions).

**S13 Fig.** Correlation between maximum intensity of selection $\text{max}(1/\alpha)$ and geographic range of co-mimicking H. melpomene (above) and H. erato (below) races. Areas were calculated from distribution data obtained from Rosser and colleagues [136] using an alpha hull polygon (code available at https://github.com/StevenVB12/Sample-distributions).

**S14 Fig.** Additional SweepFinder2 [74,76] and VolcanoFinder [27] analyses of publicly available data for H. c. galanthus [49]. The regions containing the tandem copies of aristaless, al1 and al2, WntA, cortex, and optix (left to right) are depicted. Colour pattern genes are annotated in red in the gene annotation panel. On the y-axis Sweepfinder2’s and VolcanoFinder’s CLR statistics are shown (peaks capped at 1,000). The colour gradient indicates the estimated intensity of selection $\alpha$ (black . . . high $\alpha$ values, weak selection; red . . . low $\alpha$ values, strong selection). Grey shadings indicate annotated colour pattern CREs [28,30,36,37,39] (S7–S10 Figs). Coloured horizontal bars indicate regions with CLR values above threshold and for VolcanoFinder results, the colour gradient indicates the estimated $D$ value. Top panel shows colour pattern phenotypes, and symbols indicate distinct colour pattern elements and their presence is annotated in population panels. Note that the yellow hindwing bar controlled by the cortex region can be expressed on the dorsal and ventral side (yellow/yellow square symbol) or on the ventral side only (black/yellow square symbol) [39]. Moreover, the actual shape of the forewing band can depend on the allelic state of WntA. Full, grey lines connect colour pattern elements with annotated CREs. The H. c. galanthus phenotype is depicted on the right. CLR, composite likelihood ratio; CRE, cis-regulatory element.

**S15 Fig.** VolcanoFinder [27] scans across colour pattern regions in the H. melpomene clade. The regions containing the tandem copies of aristaless, al1 and al2, WntA, cortex, and optix (left to right) are depicted. Colour pattern genes are annotated in red in the gene
annotation panel. On the y-axis VolcanoFinder’s CLR statistics is shown (peaks capped at 1,000). The colour gradient indicates the estimated intensity of selection $\alpha$ (black...high $\alpha$ values, weak selection; red...low $\alpha$ values, strong selection). Grey shadings indicate annotated colour pattern CREs [28,30,36,37,39] (S7–S10 Figs). Coloured horizontal bars indicate regions with CLR values above threshold and the colour gradient indicates the estimated $D$ value. Top panel shows colour pattern phenotypes and symbols indicate distinct colour pattern elements and their presence is annotated in population panels. Note that the yellow hindwing bar controlled by the cortex region can be expressed on the dorsal and ventral side (yellow/yellow square symbol) or on the ventral side only (black/yellow square symbol) [39]. Moreover, the actual shape of the forewing band can depend on the allelic state of WntA. Full, grey lines connect colour pattern elements with annotated CREs. CLR, composite likelihood ratio; CRE, cis-regulatory element.

S16 Fig. VolcanoFinder [27] scans across neutral background regions in the H. melpomene clade. Genes are annotated in the top gene annotation panel. On the y-axis VolcanoFinder’s CLR statistics is shown (peaks are capped at 1,000). The colour gradient indicates the estimated intensity of selection $\alpha$ (black...high $\alpha$ values, weak selection; red...low $\alpha$ values, strong selection). Coloured horizontal bars indicate regions with CLR values above threshold and the colour gradient indicates the estimated $D$ value. CLR, composite likelihood ratio.

S17 Fig. Superposition of SweepFinder2’s [74,76] composite likelihood ratio peaks of all H. melpomene clade populations for each of the 4 colour pattern regions. Superimposed, semi-transparent SweepFinder2 peaks are depicted in grey. Colour pattern genes (yellow), known CREs (red), and additional genes with evidence for a putative role in colour patterning (blue and green for genes discussed in the main text) are highlighted and assigned a number in the top row. The scale on the x-axes differs and the y-axis is capped at CLR = 1,500. (A) aristaless1 (yellow, 2), aristaless1 CRE (red, 3) [28], aristaless2 (blue, 1); (B) WntA (yellow, 4), CRE associated with split forewing band identified in this study (red, 5); (C) cortex (yellow, 10), CREs for dorsal (11) and ventral (12) hindwing topology [39], a region containing SNPs with strongest association with forewing band [30] (red), additional genes with evidence for wing patterning control [30] (blue: 7, 8, 9, 14, 15, 16, 18, 19, 21, 22, 23; green: 17 (LMTK1/HM00033), 20 (washout/WAS homologue 1/HM00036); also see S9 Table); (D) optix (yellow, 23), CREs for ‘band1’(24), ‘band2’(26), ‘rays’(25) and ‘dennis’(27) (red) [36,37], kinesin (green, 28) [86,87]. A genome viewer in which these regions and accession can be viewed in detail is available at http://lepbase.org/. CLR, composite likelihood ratio; CRE, cis-regulatory element; SNP, single-nucleotide polymorphism.

S18 Fig. Superposition of SweepFinder2 [74,76] composite likelihood ratio peaks of all H. erato clade populations for each of the 4 colour pattern regions. Superimposed, semi-transparent SweepFinder2 peaks are depicted in grey. Colour pattern genes (yellow), known CREs (red), and additional genes with evidence for a putative role in colour patterning (blue and green for genes discussed in the main text) are highlighted and assigned a number in the top row. The scale on the x-axes differs and the y-axis is capped at CLR = 1,500. (A) WntA (yellow,1), CREs associated with ‘Sd1’(2), ‘Sd2’(3), ‘St’(4), ‘Ly1’(5) and ‘Ly2’(6) elements (red); (B) cortex (yellow, 8), ‘Cr1’(7) and ‘Cr2’(9) regions (red) [38], and additional genes with evidence for wing patterning control [30] (blue: 10,12; green: 11 (washout/WAS homologue 1/HERA000061), 13 (lethal (2)/HERA000062); also see S9 Table; (C) optix (yellow,14), CREs for
'rays' (15), 'band' Y1 (16)/ Y2 (18), and 'dennis' D1 (17)/ D2 (19) elements (red) [38]. A genome viewer in which these regions and accession can be viewed in detail is available at http://lepbase.org/. CLR, composite likelihood ratio; CRE, colour pattern regulatory element.

S19 Fig. Superposition of VolcanoFinder2’s [27] composite likelihood ratio peaks of all H. melpomene clade populations for each of the 4 colour pattern regions. Superimposed, semi-transparent VolcanoFinder2 peaks are depicted in grey. Colour pattern genes (yellow), known CREs (red), and additional genes with evidence for a putative role in colour patterning (blue and green for genes discussed in the main text) are highlighted and assigned a number in the top row. The scale on the x-axes differs and the y-axis is capped at CLR = 2,000. (A) aristaless1 (yellow, 2), aristaless1 CRE (red, 3) [28], aristaless2 (blue, 1); (B) WntA (yellow, 4), CRE associated with split forewing band identified in this study (red, 5); (C) cortex (yellow, 10), CREs for dorsal (11) and ventral (12) hindwing topology [39], a region containing SNPs with strongest association with forewing band [30] (13), additional genes with evidence for wing patterning control [30] (blue: 7, 8, 9, 14, 15, 16, 18, 19, 21, 22, 23; green: 17 (LMTK1 /HM00033), 20 (washout/WAS homologue 1/HM00036); also see S9 Table); (D) optix (yellow, 23), CREs for 'band1' (24), 'band2' (26), 'rays' (25) and 'dennis' (27) (red) [36, 37], kinesin (green, 28) [86, 87]. A genome viewer in which these regions and accession can be viewed in detail is available at http://lepbase.org/. CLR, composite likelihood ratio; CRE, cis-regulatory element.

S20 Fig. Signature of selection across colour pattern regions in the H. erato clade. The regions containing WntA, cortex, and optix (left to right) are depicted. Colour pattern genes are annotated in red in the gene annotation panel. On the y-axis Sweepfinder2’s [74,76] CLR statistics is shown (peaks are capped at CLR = 1,000). The colour gradient indicates the estimated intensity of selection (black = high $\alpha$ values, weak selection; red = low $\alpha$ values, strong selection). Blue horizontal bars indicate regions above the CLR threshold value. CLR, composite likelihood ratio

S21 Fig. Signature of selection across neutral background regions in the H. erato clade. Genes are annotated in the top gene annotation panel. On the y-axis Sweepfinder2’s [74,76] CLR statistics is shown (peaks are capped at 1,000). The colour gradient indicates the estimated intensity of selection (black = high $\alpha$ values, weak selection; red = low $\alpha$ values, strong selection). Blue horizontal bars indicate regions above the CLR threshold value. CLR, composite likelihood ratio

S22 Fig. Summary and selection statistics across colour pattern regions for all populations analysed in the Heliconius erato clade. For each population genotyping coverage (calculated as proportion of retained genotypes after quality filtering in 500 bp windows), nucleotide diversity, Kelly’s $Z_{ns}$, Tajima’s D, pooled integrated haplotype homozygosity score, and SweepFinder2’s [74,76] CLR statistics across each colour pattern region are shown (top to bottom). File names contain population and colour pattern region identifiers (Herato1001 = WntA scaffold, Herato1505 = cortex scaffold, Herato1801 = optix scaffold). The 18 single figures have been uploaded to GitHub: https://github.com/markusmoest/SelectionHeliconius/tree/master/S22_Fig_H_erato. CLR, composite likelihood ratio.

S1 Table. Sample information and genotyping statistics for all samples from the H. melpomene clade.
S2 Table. Per-population sample sizes for the *H. melpomene* clade and the *H. erato* clade used in the respective analyses.

(PDF)

S3 Table. Sample information for whole-genome sequence data used for PSMC’ analysis. PSMC’, Pairwise Sequentially Markovian Coalescent.

(PDF)

S4 Table. Average neutral equilibrium values of nucleotide site diversity (*π*), Tajima’s *D* and Kelly’s *Zns* for our simulated populations, both without migration (i.e., the simulated population at equilibrium prior to experiencing a classic sweep) and with migration (i.e., the simulated population at equilibrium after an introgressed sweep). Values are labelled by the parameter values of the simulations from which they were generated (*s* = selection coefficient; *μ* = mutation rate per base pair/generation).

(PDF)

S5 Table. Position, CLR statistics and strength of selection (*α*, 2*N*<sub>e</sub>*s*, and *s*) for the highest CLR and the smallest *α* value on each colour pattern scaffold (*α*<sub>min</sub>) for the *H. melpomene* clade. Additional relevant peaks on scaffolds are also given. Data are from SweepFinder2 [74,76] runs with background SFS estimated from background scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum.

(PDF)

S6 Table. Position, CLR statistics and strength of selection (*α*, 2*N*<sub>e</sub>*s*, and *s*) for the highest CLR and the smallest *α* value on each background scaffold (*α*<sub>min</sub>) for the *H. melpomene* clade. Data are from SweepFinder2 [74,76] runs with background SFS estimated from background scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum.

(PDF)

S7 Table. Position, CLR statistics and strength of selection (*α*, 2*N*<sub>e</sub>*s*, and *s*) for the highest CLR and the smallest *α* value on each colour pattern scaffold (*α*<sub>min</sub>) for the *H. melpomene* clade. Additional relevant peaks on scaffolds are also given. Data are from SweepFinder2 [74,76] runs with background SFS estimated from background and colour pattern scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum.

(PDF)

S8 Table. Position, CLR statistics and strength of selection (*α*, 2*N*<sub>e</sub>*s*, and *s*) for the highest CLR and the smallest *α* value on each background scaffold (*α*<sub>min</sub>) for *H. erato*. Data are from SweepFinder2 [74,76] runs with background SFS estimated from background and colour pattern scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum.

(PDF)

S9 Table. List of additional genes with significant colour pattern associations on the cortex scaffold from Nadeau and colleagues [30] that overlap with or are in proximity of selection signatures detected in this study.

(PDF)

S10 Table. Sample information and genotyping statistics for all samples from the *H. erato* clade from Van Belleghem and colleagues [38].

(PDF)

S11 Table. Position, CLR statistics and strength of selection (*α*, 2*N*<sub>e</sub>*s*, and *s*) for the highest CLR and the smallest *α* value on each colour pattern scaffold (*α*<sub>min</sub>) for *H. erato*. 

(PDF)
Additional relevant peaks on scaffolds are also given. Data are from SweepFinder2 [74,76] runs with background SFS estimated from background scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum.

(S12 Table. Position, CLR statistics and strength of selection (\(\alpha\), 2\(N_e s\), and \(s\)) for the highest CLR and the smallest \(\alpha\) value on each background scaffold (\(\alpha_{\text{min}}\)) for \(H.\ erato\). Data are from SweepFinder2 [74,76] runs with background SFS estimated from background scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum. (PDF)

(S13 Table. Position, CLR statistics and strength of selection (\(\alpha\), 2\(N_e s\), and \(s\)) for the highest CLR and the smallest \(\alpha\) value on each colour pattern scaffold (\(\alpha_{\text{min}}\)) for \(H.\ erato\). Additional relevant peaks on scaffolds are also given. Data are from SweepFinder2 [74,76] runs with background SFS estimated from background and colour pattern scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum. (PDF)

(S14 Table. Position, CLR statistics and strength of selection (\(\alpha\), 2\(N_e s\), and \(s\)) for the highest CLR and the smallest \(\alpha\) value on each background scaffold (\(\alpha_{\text{min}}\)) for \(H.\ erato\). Data are from SweepFinder2 [74,76] runs with background SFS estimated from background and colour pattern scaffolds. CLR, composite likelihood ratio; SFS, site frequency spectrum. (PDF)

(S15 Table. Per-population and per-scaffold summary statistics estimates and standard deviation for colour pattern scaffolds in the \(H.\ melpomene\) clade. (PDF)

(S16 Table. Per-population and per-scaffold summary statistics estimates and standard deviation for neutral background scaffolds in the \(H.\ melpomene\) clade. (PDF)

(S17 Table. Per-population and per-scaffold summary statistics estimates and standard deviation for colour pattern scaffolds in the \(H.\ erato\) clade. (PDF)

(S18 Table. Per-population and per-scaffold summary statistics estimates and standard deviation for neutral background scaffolds in the \(H.\ erato\) clade. (PDF)

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References

1. Linnen CR, Hoekstra HE. Measuring natural selection on genotypes and phenotypes in the wild. Cold Spring Harb Symp Quant Biol. 2009; 74: 155–168. https://doi.org/10.1101/csb.2009.74.045 PMID: 20413707

2. Kingsolver JG, Hoekstra HE, Hoekstra JM, Berrigan D, Vignieri SN, Hill CE, et al. The strength of phenotypic selection in natural populations. The American Naturalist. 2001; 157: 245–261. https://doi.org/10.1086/319193 PMID: 18707288

3. Endler JA. Natural Selection in the Wild. Princeton University Press, Princeton, New Jersey; 1986. Available from: https://press.princeton.edu/titles/2354.html. [cited 27 Jan 2020].

4. Smith JM, Haigh J. The hitch-hiking effect of a favourable gene. Genet Res. 1974; 23: 23–35. PMID: 4407212

5. Stephan W. Selective sweeps. Genetics. 2019; 211: 5–13. https://doi.org/10.1534/genetics.118.301319 PMID: 30626638

6. Clark RM, Linton E, Messing J, Doebley JF. Pattern of diversity in the genomic region near the maize domestication gene tb1. PNAS. 2004; 101: 700–707. https://doi.org/10.1073/pnas.2237049100 PMID: 14701910

7. Rubin C-J, Zody MC, Eriksson J, Meadows JRS, Shenwood E, Webster MT, et al. Whole-genome resequencing reveals loci under selection during chicken domestication. Nature. 2010; 464: 587–591. https://doi.org/10.1038/nature08832 PMID: 20220755

8. Qanbari S, Pausch H, Jansen S, Somel M, Strom TM, Fries R, et al. Classic selective sweeps revealed by massive sequencing in cattle. PLoS Genet. 2014; 10: e1004148. https://doi.org/10.1371/journal.pgen.1004148 PMID: 24546189

9. Williamson SH, Hubisz MJ, Clark AG, Payseur BA, Bustamante CD, Nielsen R. Localizing recent adaptive evolution in the human genome. PLoS Genet. 2007; 3: e90. https://doi.org/10.1371/journal.pgen.0030090 PMID: 17542651
10. Linnen CR, Poh Y-P, Peterson BK, Barrett RDH, Larson JG, Jensen JD, et al. Adaptive evolution of multiple traits through multiple mutations at a single gene. Science. 2013; 339: 1312–1316. https://doi.org/10.1126/science.1233213 PMID: 23493712

11. Garud NR, Messer PW, Buzbas EO, Petrov DA. Recent selective sweeps in North American Drosophila melanogaster show signatures of soft sweeps. PLoS Genet. 2015; 11: e1005004. https://doi.org/10.1371/journal.pgen.1005004 PMID: 25706129

12. Tavares H, Whibley A, Field DL, Bradley D, Couchman M, Copsey L, et al. Selection and gene flow shape genomic islands that control floral guides. PNAS. 2018; 115: 11006–11011. https://doi.org/10.1073/pnas.1801832115 PMID: 30297406

13. Marques DA, Taylor JS, Jones FC, Di Palma F, Kingsley DM, Reimchen TE. Convergent evolution of SWS2 opsin facilitates adaptive radiation of threespine stickleback into different light environments. PLoS Biol. 2017; 15. https://doi.org/10.1038/s41596-017-036947 PMID: 28399148

14. Marques DA, Jones FC, Palma FD, Kingsley DM, Reimchen TE. Experimental evidence for rapid genomic adaptation to a new niche in an adaptive radiation. Nature Ecology & Evolution. 2018; 2: 1128–1138. https://doi.org/10.1038/s41559-018-0581-8 PMID: 30297406

15. Lamichhaney S, Han F, Berglund J, Wang C, Alméén MS, Webster MT, et al. A beak size locus in Darwin’s finches facilitated character displacement during a drought. Science. 2016; 352: 470–474. https://doi.org/10.1126/science.aad8786 PMID: 27102486

16. Barrett RDH, Laurent S, Mallarino R, Pfeifer SP, Xu CCY, Foll M, et al. Linking a mutation to survival in wild mice. Science. 2019; 363: 499–504. https://doi.org/10.1126/science.aav3824 PMID: 30705184

17. Prezeworski M, Coop G, Wall JD. The signature of positive selection on standing genetic variation. Evolution. 2005; 59: 2312–2323. https://doi.org/10.1111/j.0014-3820.2005.tb00941.x PMID: 16396172

18. Barrett RDH, Schluter D. Adaptation from standing genetic variation. Trends in Ecology & Evolution. 2008; 23: 38–44. https://doi.org/10.1016/j.tree.2007.09.008 PMID: 18006185

19. Hermisson J, Pennings PS. Soft sweeps I—molecular population genetics of adaptation from standing genetic variation. Genetics. 2005; 169: 2335–2352. https://doi.org/10.1534/genetics.104.036947 PMID: 15716498

20. Pennings PS, Hermisson J. Soft sweeps II—molecular population genetics of adaptation from recurrent mutation or migration. Mol Biol Evol. 2006; 23: 1076–1084. https://doi.org/10.1093/molbev/msj117 PMID: 16520336

21. Pennings PS, Hermisson J. Soft sweeps III: the signature of positive selection from recurrent mutation. PLoS Genet. 2006; 2: e186. https://doi.org/10.1371/journal.pgen.0020186 PMID: 17173482

22. Jones FC, Grabherr MG, Chan YF, Russell P, Mauceli E, Johnson J, et al. The genomic basis of adaptive evolution in threespine sticklebacks. Nature. 2012; 484: 55–61. https://doi.org/10.1038/nature10944 PMID: 22481358

23. Meier JI, Marques DA, Mwaiko S, Wagner CE, Excoffier L, Seehausen O. Ancient hybridization fuels rapid cichlid fish adaptive radiations. Nature Communications. 2017; 8: 14363. https://doi.org/10.1038/ncomms14363 PMID: 2816104

24. Van Belleghem SM, Vangestel C, Wolf KD, Corte ZD, Most M, Rastas P, et al. Evolution at two time frames: polymorphisms from an ancient singular divergence event fuel contemporary parallel evolution. PLoS Genet. 2018; 14: e1007796. https://doi.org/10.1371/journal.pgen.1007796 PMID: 30422983

25. Nadeau NJ, Pardo-Diaz C, Whibley A, Supple MA, Saenko SV, Wallbank RWR, et al. The gene cortex controls mimicry and crypsis in butterflies and moths. Nature. 2016; 534: 106–110. https://doi.org/10.1038/nature17961 PMID: 27251285
31. Baxter SW, Nadeau NJ, Maroja LS, Wilkinson P, Counterman BA, Dawson A, et al. Genomic hotspots for adaptation: the population genetics of Müllerian mimicry in the Heliconius melpomene clade. PLoS Genet. 2010; 6: e1000794. https://doi.org/10.1371/journal.pgen.1000794 PMID: 20140188

32. Counterman BA, Araujo-Perez F, Hines HM, Baxter SW, Morrison CM, Lindstrom DP, et al. Genomic hotspots for adaptation: the population genetics of Müllerian mimicry in Heliconius erato. PLoS Genet. 2010; 6: e1000796. https://doi.org/10.1371/journal.pgen.1000796 PMID: 20140239

33. Morris J, Navarro N, Rastas P, Rawlins LD, Sammey J, Mallet J, et al. The genetic architecture of adaptation: convergence and pleiotropy in Heliconius wing pattern evolution. Heredity. 2019; 1. https://doi.org/10.1038/s41371-018-0180-0 PMID: 30670842

34. Nadeau NJ, Ruiz M, Salazar P, Counterman B, Medina JA, Ortiz-Zuazaga H, et al. Population genomics of parallel hybrid zones in the mimetic butterflies, H. melpomene and H. erato. Genome Res. 2014; 24: 1316–1333. https://doi.org/10.1101/gr.169292.113 PMID: 24823669

35. Reed RD, Papa R, Martin A, Hines HM, Counterman BA, Pardo-Díaz C, et al. optix drives the repeated convergent evolution of butterfly wing pattern mimicry. Science. 2011; 333: 1137–1141. https://doi.org/10.1126/science.1208227 PMID: 21778360

36. Wallbank RWR, Baxter SW, Pardo-Díaz C, Hanly JJ, Martin SH, Mallet J, et al. Evolutionary novelty in a butterfly wing pattern through enhancer shuffling. PLoS Biol. 2016; 14: e1002353. https://doi.org/10.1371/journal.pbio.1002353 PMID: 26771987

37. Hanly JJ. Developmental basis of wing pattern diversity in Heliconius. University of Cambridge. 2017.

38. Van Belleghem SM, Rastas P, Papanicolaou A, Martin SH, Arias CF, Supple MA, et al. Complex modular architecture around a simple toolkit of wing pattern genes. Nature Ecology & Evolution. 2017; 1: 0052. https://doi.org/10.1038/s41559-016-0052 PMID: 28523290

39. Enciso-Romero J, Pardo-Díaz C, Martin SH, Arias CF, Linares M, McMillan OW, et al. Evolution of novel mimicry rings facilitated by adaptive introgression in tropical butterflies. Molecular Ecology. 2017; 26: 5160–5172. https://doi.org/10.1111/mec.14277 PMID: 28777894

40. Lewis JJ, Geltman RC, Pollak PC, Rondem KE, Belleghem SMV, Hubisz MJ, et al. Parallel evolution of ancient, pleiotropic enhancers underlies butterfly wing pattern mimicry. PNAS. 2019 [cited 13 Nov 2019]. https://doi.org/10.1073/pnas.1907068116 PMID: 31712408

41. Concha C, Wallbank RWR, Hanly JJ, Fenner J, Livraghi L, Rivera ES, et al. Interplay between developmental flexibility and determinism in the evolution of mimetic Heliconius wing patterns. Current Biology. 2019; 0: 4096–4099. https://doi.org/10.1016/j.cub.2019.02.002 PMID: 31355676

42. Jay P, Whibley A, Frézal L, Rodríguez de Cara MÁ, Nowell RW, Mallet J, et al. Supergene evolution triggered by the introgression of a chromosomal inversion. Current Biology. 2018; 28: 1839–1845.e3. https://doi.org/10.1016/j.cub.2018.04.072 PMID: 29804810

43. Pardo-Díaz C, Salazar C, Baxter SW, Merot C, Figueiredo-Ready W, Joron M, et al. Adaptive introgression across species boundaries in Heliconius butterflies. PLoS Genet. 2012; 8: e1002752. https://doi.org/10.1371/journal.pgen.1002752 PMID: 22737081

44. The Heliconius Genome Consortium, Dasmahapatra KK, Walters JR, Briscoe AD, Davey JW, Whibley A, et al. Butterfly genome reveals promiscuous exchange of mimicry adaptations among species. Nature. 2012; 487: 94–98. https://doi.org/10.1038/nature11041 PMID: 22722851

45. Zhang W, Dasmahapatra KK, Mallet J, Moreira GRP, Kronforst MR. Genome-wide introgression among distantly related Heliconius butterfly species. Genome Biology. 2016; 17: 25. https://doi.org/10.1186/s13059-016-0889-0 PMID: 26921238

46. Edelman NB, Frandsen PB, Miyagi M, Clavijo B, Davey J, Dikow RB, et al. Genomic architecture and introgression shape a butterfly radiation. Science. 2019; 366: 594–599. https://doi.org/10.1126/science.aaw2090 PMID: 31672890

47. Martin SH, Eriksson A, Kozak KM, Manica A, Jiggins CD. Speciation in Heliconius butterflies: minimal contact followed by millions of generations of hybridisation. bioRxiv. 2015; 015800. https://doi.org/10.1101/015800

48. Lohse K, Chmelik M, Martin SH, Barton NH. Efficient strategies for calculating blockwise likelihoods under the coalescent. Genetics. 2016; 202: 775–786. https://doi.org/10.1534/genetics.115.183814 PMID: 26715666

49. Kronforst MR, Hansen MEB, Crawford NG, Gallant JR, Zhang W, Kulathinal RJ, et al. Hybridization reveals the evolving genomic architecture of speciation. Cell Reports. 2013; 5: 666–677. https://doi.org/10.1016/j.celrep.2013.09.042 PMID: 24183670

50. Kozak KM, Wahlaberg N, Neild AFE, Dasmahapatra KK, Mallet J, Jiggins CD. Multilocus species trees show the recent adaptive radiation of the mimetic Heliconius butterflies. Syst Biol. 2015; 64: 505–524. https://doi.org/10.1093/sysbio/syv007 PMID: 25634098
51. Van Belleghem SM, Baquero M, Papa R, Salazar C, McMillan WO, Counterman BA, et al. Patterns of Z chromosome divergence among Heliconius species highlight the importance of historical demography. Molecular Ecology. 2018;0. https://doi.org/10.1111/mec.14560 PMID: 29569384

52. Kozak KM, McMillan O, Joron M, Jiggins CD. Genome-wide admixture is common across the Heliconius radiation. bioRxiv. 2018; 414201. https://doi.org/10.1101/414201

53. Jiggins CD. The Ecology and Evolution of Heliconius Butterflies. Oxford, New York: Oxford University Press, Oxford, New York; 2016.

54. Benson WW. Natural selection for Mullerian mimicry in Heliconius erato in Costa Rica. Science. 1972; 176: 936–939. https://doi.org/10.1126/science.176.4037.936 PMID: 17829303

55. Mallet J, Barton NH. Strong natural selection in a warning-color hybrid zone. Evolution. 1989; 43: 421–431. https://doi.org/10.1111/j.1558-5646.1989.tb04237.x PMID: 28568556

56. Kapan DD. Three-butterfly system provides a field test of mullerian mimicry. Nature. 2001; 409: 338–340. https://doi.org/10.1038/35053066 PMID: 26858416

57. Chouteau M, Arias M, Joron M. Warning signals are under positive frequency-dependent selection in nature. PNAS. 2016; 113: 2164–2169. https://doi.org/10.1073/pnas.1519216113 PMID: 26858416

58. Merrill RM, Wallbank RWR, Bull V, Salazar PCA, Mallet J, Stevens M, et al. Disruptive ecological selection on a mating cue. Proc Biol Sci. 2012; 279: 4907–4913. https://doi.org/10.1098/rspb.2012.1968 PMID: 23075843

59. Barton NH, Hewitt GM. Analysis of hybrid zones. Annu Rev Ecol Syst. 1985; 16: 113–148. https://doi.org/10.1146/annurev.es.16.110185.000553

60. Mallet J, Barton N. Inference from clines stabilized by frequency-dependent selection. Genetics. 1989; 122: 921–936. PMID: 2759433

61. Mallet J, Barton N, Gerardo LM, Jose SC, Manuel MM, Eeley H. Estimates of selection and gene flow from measures of cline width and linkage disequilibrium in Heliconius hybrid zones. Genetics. 1990; 124: 921–936. PMID: 2323556

62. Rosser N, Dasmahapatra KK, Mallet J. Stable Heliconius butterfly hybrid zones are correlated with a local rainfall peak at the edge of the Amazon basin. Evolution. 2014; 68: 3470–3484. https://doi.org/10.1111/evol.12539 PMID: 25311415

63. Salazar PA. Hybridisation and the genetics of wing colour–pattern diversity in Heliconius butterflies. University of Cambridge. 2013.

64. Thurman TJ, Szegen-Sigal A, McMillan WO. Movement of a Heliconius hybrid zone over 30 years: A Bayesian approach. Journal of Evolutionary Biology. 2019;0. https://doi.org/10.1111/jeb.13499 PMID: 31216075

65. Mallet J, Jiggins CD, McMillan WO. Mimicry and warning colour at the boundary between races and species. In: Howard DJ, Berlocher SH, editors. Endless Forms: Species and Speciation. Oxford University Press, Oxford, New York; 1998. pp. 390–403.

66. Blum MJ. Rapid movement of a Heliconius hybrid zone: evidence for phase III of Wright’s shifting balance theory? Evolution. 2002; 56: 1992–1998. https://doi.org/10.1111/j.0014-3820.2002.tb00125.x PMID: 12449486

67. Hines HM, Counterman BA, Papa R, Moura PA de, Cardoso MZ, Linares M, et al. Wing patterning gene redefines the mimetic history of Heliconius butterflies. PNAS. 2011; 108: 19666–19671. https://doi.org/10.1073/pnas.1100961108 PMID: 22084094

68. Supple MA, Hines HM, Dasmahapatra KK, Lewis JJ, Nielsen DM, Lavoie C, et al. Genomic architecture of adaptive color pattern divergence and convergence in Heliconius butterflies. Genome Res. 2013; 23: 1248–1257. https://doi.org/10.1101/gr.150615.112 PMID: 23674305

69. Nadeau NJ, Annabel W, Jones Robert T., Davey,John W., Dasmahapatra Kancho n K., Baxter Simon W., et al. Genomic islands of divergence in hybridizing Heliconius butterflies identified by large-scale targeted sequencing. Philosophical Transactions of the Royal Society B: Biological Sciences. 2012; 367: 343–353. https://doi.org/10.1098/rstb.2011.0198 PMID: 22201164

70. Martin SH, Møst M, Palmer WJ, Salazar C, McMillan WO, Jiggins FM, et al. Natural selection and genetic diversity in the butterfly Heliconius melpomene. Genetcs. 2016; 203: 525–541. https://doi.org/10.1534/genetics.115.183288 PMID: 27017626

71. Martin SH, Dasmahapatra KK, Nadeau NJ, Salazar C, Walters JR, Simpson F, et al. Genome-wide evidence for speciation with gene flow in Heliconius butterflies. Genome Res. 2013; 23: 1817–1828. https://doi.org/10.1101/gr.159426.113 PMID: 24045163

72. Martin SH, Davey JW, Salazar C, Jiggins CD. Recombination rate variation shapes barriers to introgression across butterfly genomes. PLoS Biol. 2019; 17: e2006288. https://doi.org/10.1371/journal.pbio.2006288 PMID: 30730876
73. Nielsen R, Williamson S, Kim Y, Hubisz MJ, Clark AG, Bustamante C. Genomic scans for selective sweeps using SNP data. Genome Res. 2005; 15: 1566–1575. https://doi.org/10.1101/gr.4252305 PMID: 16251466

74. Huber CD, DeGiorgio M, Hellmann I, Nielsen R. Detecting recent selective sweeps while controlling for mutation rate and background selection. Molecular Ecology. 2016; 25: 142–156. https://doi.org/10.1111/mec.13351 PMID: 26293047

75. Przeworski M. The signature of positive selection at randomly chosen loci. Genetics. 2002; 160: 1179–1189. PMID: 11901132

76. DeGiorgio M, Huber CD, Hubisz MJ, Hellmann I, Nielsen R. SweepFinder 2: increased sensitivity, robustness and flexibility. Bioinformatics. 2016; 32: 1895–1897. https://doi.org/10.1093/bioinformatics/btw051 PMID: 27153702

77. Przeworski M. The signature of positive selection at randomly chosen loci. Genetics. 2002; 160: 1179–1189. PMID: 11901132

78. Keightley PD, Pinharanda A, Ness RW, Simpson F, Dasmahapatra KK, Mallet J, et al. Estimation of the spontaneous mutation rate in *Heliconius melpomene*. Mol Biol Evol. 2015; 32: 239–243. https://doi.org/10.1093/molbev/msu302 PMID: 25371432

79. Turner JR, Johnson MS, Eanes WF. Contrasted modes of evolution in the same genome: allozymes and adaptive change in *Heliconius*. PNAS. 1979; 76: 1924–1928. https://doi.org/10.1073/pnas.76.4.1924 PMID: 287032

80. Naisbit RE, Jiggins CD, Mallet J. Mimicry: developmental genes that contribute to speciation. Evolution & Development. 2003; 5: 269–280. https://doi.org/10.1046/j.1525-142X.2003.03034.x PMID: 12752766

81. Kazemi B, Gamberale-Stille G, Tulberg BS, Leima O. Stimulus salience as an explanation for imperfect mimicry. Current Biology. 2014; 24: 965–969. https://doi.org/10.1016/j.cub.2014.02.061 PMID: 24726157

82. Mazo-Vargas A, Concha C, Livraghi L, Massardo D, Wallbank RWR, Zhang L, et al. Macroevolutionary shifts of WntA function potentiate butterfly wing-pattern diversity. PNAS. 2017; 114: 10701–10706. https://doi.org/10.1073/pnas.1708149114 PMID: 28923954

83. Saenko SV, Roux C, Welch JJ, Bluegon C, Joron M, Laurens V. Unraveling the genes forming the wing pattern supergene in the polymorphic butterfly *Heliconius numata*. EvoDevo. 2019; 10: 16. https://doi.org/10.1186/s13227-019-0129-2 PMID: 31406559

84. Liu R, Abreu-Blanco MT, Barry KC, Linardopoulos EV, Osborn GE, Parkhurst SM. Wash functions downstream of Rho and links linear and branched actin nucleation factors. Development. 2009; 136: 2849–2860. https://doi.org/10.1242/dev.035246 PMID: 19633175

85. Pardo-Diaz C, Jiggins CD. Neighboring genes shaping a single adaptive mimetic trait. Evolution & Development. 2014; 16: 3–12. https://doi.org/10.1111/ede.12058 PMID: 24393463

86. Salazar C, Baxter SW, Pardo-Diaz C, Wu G, Surridge A, Linares M, et al. Genetic evidence for hybrid trait speciation in *Heliconius butterflies*. PLoS Genet. 2010; 6: e1000930. https://doi.org/10.1371/journal.pgen.1000930 PMID: 20442862

87. Sheppard MP, Turner JR, Brown K. S, Benson W. W., Singer M. C., Smith John Maynard. Genetics and the evolution of Muellerian mimicry in *Heliconius* butterflies. Philosophical Transactions of the Royal Society of London B, Biological Sciences. 1985; 308: 433–610. https://doi.org/10.1098/rstb.1985.0066

88. Hoyal-Cuthill JF, Charleston M. Phylogenetic codivergence supports coevolution of mimetic *Heliconius* butterflies. PLoS ONE. 2012; 7: e36464. https://doi.org/10.1371/journal.pone.0036464 PMID: 22586474

89. Hoyal-Cuthill JF, Charleston M. Phylogenetic codivergence supports coevolution of mimetic *Heliconius* butterflies. PLoS ONE. 2012; 7: e36464. https://doi.org/10.1371/journal.pone.0036464 PMID: 22586474

90. Hoyal-Cuthill JF, Charlsteon M. Wing patterning genes and coevolution of Müllerian mimicry in *Heliconius* butterflies: support from phylogeography, cophylogeny, and divergence times. Evolution. 2015; 69: 3082–3096. https://doi.org/10.1111/evol.12812 PMID: 26552726

91. Hoyal-Cuthill JF, Guttenberg N, Ledger S, Crowther R, Huertas B. Deep learning on butterfly phenotypes tests evolution’s oldest mathematical model. Science Advances. 2019; 5: eaaw4967. https://doi.org/10.1126/sciadv.aaw4967 PMID: 31453326

92. Eltringham H. IV. On specific and mimetic relationships in the genus *Heliconius*. L. Transactions of the Royal Entomological Society of London. 1916; 64: 101–148. https://doi.org/10.1111/j.1365-2311.1916.tb03123.x
93. Causes Mallet J. and consequences of a lack of coevolution in Müllerian mimicry. Evolutionary Ecology. 1999; 13: 777–806. https://doi.org/10.1023/A:1011060330515

94. Turner JRG. Mimicry as a model for coevolution. Arai R, Kato M, and Doi Y (eds) Biodiversity and Evolution. Tokyo: National Science Museum Foundation, Tokyo; 1995. pp. 131–152.

95. Baxter SW, Papa R, Chamberlain N, Humphray SJ, Joron M, Morrison C, et al. Convergent evolution in the genetic basis of Müllerian mimicry in Heliconius butterflies. Genetics. 2008; 180: 1567–1577. https://doi.org/10.1534/genetics.107.082982 PMID: 18791259

96. Joron M, Papa R, Beltrán M, Chamberlain N, Mavárez J, Baxter S, et al. A conserved supergene locus controls colour pattern diversity in Heliconius butterflies. PLoS Biol. 2006; 4: e303. https://doi.org/10.1371/journal.pbio.0040303 PMID: 17002517

97. Brown S, Hu N, Hombria JC-G. Identification of the first invertebrate interleukin JAK/STAT receptor, the Drosophila gene domeless. Current Biology. 2001; 11: 1700–1705. https://doi.org/10.1016/s0960-9822(01)00524-3 PMID: 11696329

98. Marroja LS, Alschuler R, McMillan WO, Jiggins CD. Partial complementarity of the mimetic yellow bar phenotype in Heliconius butterflies. PLoS ONE. 2012; 7: e48627. https://doi.org/10.1371/journal.pone.0048627 PMID: 23119074

99. Wallace AR. On the tendency of varieties to depart indefinitely from the original type. Proceedings of the Linnean Society of London. 1858. pp. 53–62.

100. Darwin C. On the Origin of Species, 1859. John Murray, London; 1859.

101. Pavlidis P, Jensen JD, Stephan W. Searching for footprints of positive selection in whole-genome SNP data from nonequilibrium populations. Genetics. 2010; 185: 907–922. https://doi.org/10.1534/genetics.110.116459 PMID: 20407129

102. Pfeifer SP, Laurent S, Sousa VC, Linnen CR, Foll M, Excoffier L, et al. The evolutionary history of Nebraska deer mice: local adaptation in the face of strong gene flow. Mol Biol Evol. 2018; 35: 792–806. https://doi.org/10.1093/molbev/msy004 PMID: 29346646

103. Saccheri IJ, Rousset F, Watts PC, Brakefield PM, Cook LM. Selection and gene flow on a diminishing cline of melanic peppered moths. PNAS. 2008; 105: 16212–16217. https://doi.org/10.1073/pnas.0803785105 PMID: 18854174

104. Price TD, Grant PR, Gibbs HL, Boag PT. Recurrent patterns of natural selection in a population of Darwin’s finches. Nature. 1984; 309: 787–789. https://doi.org/10.1038/309787a0 PMID: 6738692

105. Grant PR, Grant BR. Predicting microevolutionary responses to directional selection on heritable variation. Evolution. 1995; 49: 241–251. https://doi.org/10.1111/j.1558-5646.1995.tb02236.x PMID: 28565006

106. Van’t Hof AE, Campagne P, Rigden DJ, Yung CJ, Lingley J, Qual MA, et al. The industrial melanism mutation in British peppered moths is a transposable element. Nature. 2016; 534: 102–105. https://doi.org/10.1038/nature17951 PMID: 27251284

107. Barrett RDH, Rogers SM, Schluter D. Natural selection on a major armor gene in threespine stickleback. Science. 2008; 322: 255–257. https://doi.org/10.1126/science.1159978 PMID: 18755942

108. Reznick DN, Shaw FH, Rodd FH, Shaw RG. Evaluation of the rate of evolution in natural populations of guppies (Poecilia reticulata). Science. 1997; 275: 1934–1937. https://doi.org/10.1126/science.275.5308.1934 PMID: 9072971

109. Siepielski AM, DiBattista JD, Carlson SM. It’s about time: the temporal dynamics of phenotypic selection in the wild. Ecology Letters. 2009; 12: 1261–1276. https://doi.org/10.1111/j.1461-0248.2009.01381.x PMID: 19740111

110. Mavárez J, Salazar CA, Bermingham E, Salcedo C, Jiggins CD, Linares M. Speciation by hybridization in Heliconius butterflies. Nature. 2006; 441: 868–871. https://doi.org/10.1038/nature04738 PMID: 16778888

111. Arnold BJ, Lahner B, DaCosta JM, Weisman CM, Hollister JD, Salt DE, et al. Borrowed alleles and convergence in serpentine adaptation. PNAS. 2016; 113: 8320–8325. https://doi.org/10.1073/pnas.1600405113 PMID: 27357660

112. Ozizor EM, Reid NM, Yair S, Lee KM, VerPlieog SG, Bruns PC, et al. Adaptive introgression enables evolutionary rescue from extreme environmental pollution. Science. 2019; 364: 455–457. https://doi.org/10.1126/science.aav155 PMID: 31048485

113. Yeaman S, Aeschbacher S, Bürger R. The evolution of genomic islands by increased establishment probability of linked alleles. Molecular Ecology. 2016; 25: 2542–2558. https://doi.org/10.1111/mec.13611 PMID: 27206531

114. Mallet J. Speciation, raciation, and color pattern evolution in Heliconius butterflies: evidence from hybrid zones. In: Harrison RG, editor. Hybrid Zones and the Evolutionary Process. New York: Oxford University Press, New York; 1993. pp. 226–260.
115. Ferguson L, Lee SF, Chamberlain N, Nadeau N, Joron M, Baxter S, et al. Characterization of a hotspot for mimicry: assembly of a butterfly wing transcriptome to genomic sequence at the HmYb/Sb locus. Molecular Ecology. 2010; 19: 240–254. https://doi.org/10.1111/j.1365-294X.2009.04475.x PMID: 20331783

116. Li H. Aligning sequence reads, clone sequences and assembly contigs with BWA-MEM. arXiv:1303.3997 [q-bio]. 2013. Available from: http://arxiv.org/abs/1303.3997, [cited 17 Sep 2018].

117. Li H, Handsaker B, Wysoker A, Fennell T, Ruan J, Homer N, et al. The sequence alignment/map format and SAMtools. Bioinformatics. 2009; 25: 2078–2079. https://doi.org/10.1093/bioinformatics/btp352 PMID: 19505943

118. Van der Auwera GA, Carneiro MO, Hartl C, Poplin R, Angel G del, Levy-Moonshine A, et al. From fastQ data to high-confidence variant calls: the genome analysis toolkit best practices pipeline. Current Protocols in Bioinformatics. 2013; 43: 11.10.1–11.10.33. https://doi.org/10.1002/0471250953.bti110s43 PMID: 25431634

119. Delaneau O, Marchini J, Zagury J-F. A linear complexity phasing method for thousands of genomes. Nature Methods. 2012; 9: 179–181. https://doi.org/10.1038/nmeth.1785 PMID: 22138821

120. Price MN, Dehal PS, Arkin AP. FastTree 2—approximately maximum-likelihood trees for large alignments. PLOS ONE. 2010; 5: e9490. https://doi.org/10.1371/journal.pone.0009490 PMID: 20224823

121. Schifflers E, Durbin R. Inferring human population size and separation history from multiple genome sequences. Nature Genetics. 2014; 46: 919–925. https://doi.org/10.1038/ng.3015 PMID: 24952747

122. Li H, Durbin R. Inference of human population history from individual whole-genome sequences. Nature. 2011; 475: 493–496. https://doi.org/10.1038/nature10231 PMID: 21753753

123. Haller BC, Messer PW. SLiM 2: flexible, interactive forward genetic simulations. Mol Biol Evol. 2017; 34: 230–240. https://doi.org/10.1093/molbev/msw211 PMID: 27702775

124. Messer PW. SLiM: simulating evolution with selection and linkage. Genetics. 2013; 194: 1037–1039. https://doi.org/10.1534/genetics.113.152181 PMID: 23709637

125. Hernandez RD. A flexible forward simulator for populations subject to selection and demography. Bioinformatics. 2008; 24: 2786–2787. https://doi.org/10.1093/bioinformatics/btn522 PMID: 18843601

126. Pavlidis P, Žíkovčí D, Stamatakis A, Alachiotis N. SweeD: likelihood-based detection of selective sweeps in thousands of genomes. Mol Biol Evol. 2013; 30: 2224–2234. https://doi.org/10.1093/molbev/msr112 PMID: 23777627

127. Martin SH, Van Belleghem SM. Exploring evolutionary relationships across the genome using topology weighting. Genetics. 2017; 206: 429–438. https://doi.org/10.1534/genetics.116.194720 PMID: 28341652

128. Stamatakis A. RAxML version 8: a tool for phylogenetic analysis and post-analysis of large phylogenies. Bioinformatics. 2014; 30: 1312–1313. https://doi.org/10.1093/bioinformatics/btu033 PMID: 24451623

129. De Mita S, Siol M. EggLib: processing, analysis and simulation tools for population genetics and genomics. BMC Genetics. 2012; 13: 27. https://doi.org/10.1186/1471-2156-13-27 PMID: 22494792

130. Torres R, Szpiech ZA, Hernandez RD. Human demographic history has amplified the effects of background selection across the genome. PLoS Genet. 2018; 14: e1007387. https://doi.org/10.1016/j.tpb.2004.04.002 PMID: 15302222

131. Szpiech ZA, Hernandez RD. selScan: an efficient multithreaded program to perform EHH-based scans for positive selection. Mol Biol Evol. 2014; 31: 2824–2827. https://doi.org/10.1093/molbev/msu352 PMID: 25015648

132. Nielsen R, Williamson S, Kim Y, Hubisz MJ, Clark AG, Bustamante C. Genomic scans for selective sweeps using SNP data. Genome Research. 2005; 1566–1575. https://doi.org/10.1101/gr.4252305 PMID: 16251466

133. Davey JW, Barker SL, Rastas PM, Pinharanda A, Martin SH, Durbin R, et al. No evidence for maintenance of a sympatric Heliconius species barrier by chromosomal inversions. Evolution Letters. 2017; 1: 138–154. https://doi.org/10.1002/evl3.12 PMID: 30283645

134. Durrett R, Schweinsberg J. Approximating selective sweeps. Theoretical Population Biology. 2004; 66: 129–138. https://doi.org/10.1016/j.tpb.2004.04.002 PMID: 15932222

135. Challis RJ, Kumar S, Dasmahapatra KKK, Jiggins CD, Blaxter M. Lepbase: the lepidoptera n genome database. bioRxiv. 2016; 056994. https://doi.org/10.1101/056994

136. Rosser N, Kozak KM, Phillimore AB, Mallet J. Extensive range overlap between heliconine sister species: evidence for sympatric speciation in butterflies? BMC Evolutionary Biology. 2015; 15: 125. https://doi.org/10.1186/s12862-015-0420-3 PMID: 26123545

137. Davey JW, Chouteau M, Barker SL, Maroja L, Baxter SW, Simpson F, et al. Major improvements to the Heliconius melpomene genome assembly used to confirm 10 chromosome fusion events in 6 million years of butterfly evolution. G3: Genes, Genomes, Genetics. 2016;6: 695–708. https://doi.org/10.1534/g3.115.023655