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2 **Multiple paths towards repeated**

3 **phenotypic evolution in the spiny-leg**

4 **adaptive radiation (*Tetragnatha*;**

5 **Hawai i)**

6 **Running Title Repeated Phenotypic Evolution in spiny-legs**

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32 **Keywords** Parallel evolution, Convergent evolution, Araneae, Tetragnathidae, circadian

33 rhythms, neuronal, melanin, genomic divergence, hybridization, introgression

34 Abstract

35 The repeated evolution of phenotypes is ubiquitous in nature and offers some of the clearest
36 evidence of the role of natural selection in evolution. The genomic basis of repeated phenotypic
37 evolution is often complex and can arise from a combination of gene flow, shared ancestral
38 polymorphism and *de novo* mutation. Here, we investigate the genomic basis of repeated
39 ecomorph evolution in the adaptive radiation of the Hawaiian spiny-leg *Tetragnatha*. This
40 radiation comprises four ecomorphs that are microhabitat-specialists, and differ in body
41 pigmentation and size (Green, Large Brown, Maroon, and Small Brown). Using 76 newly
42 generated low-coverage, whole-genome resequencing samples, coupled with population
43 genomic and phylogenomic tools, we studied the evolutionary history of the radiation to
44 understand the evolution of the spiny-leg lineage and the genetic underpinnings of ecomorph
45 evolution. Congruent with previous works, we find that each ecomorph has evolved twice, with
46 the exception of the Small Brown ecomorph, which has evolved three times. The evolution of the

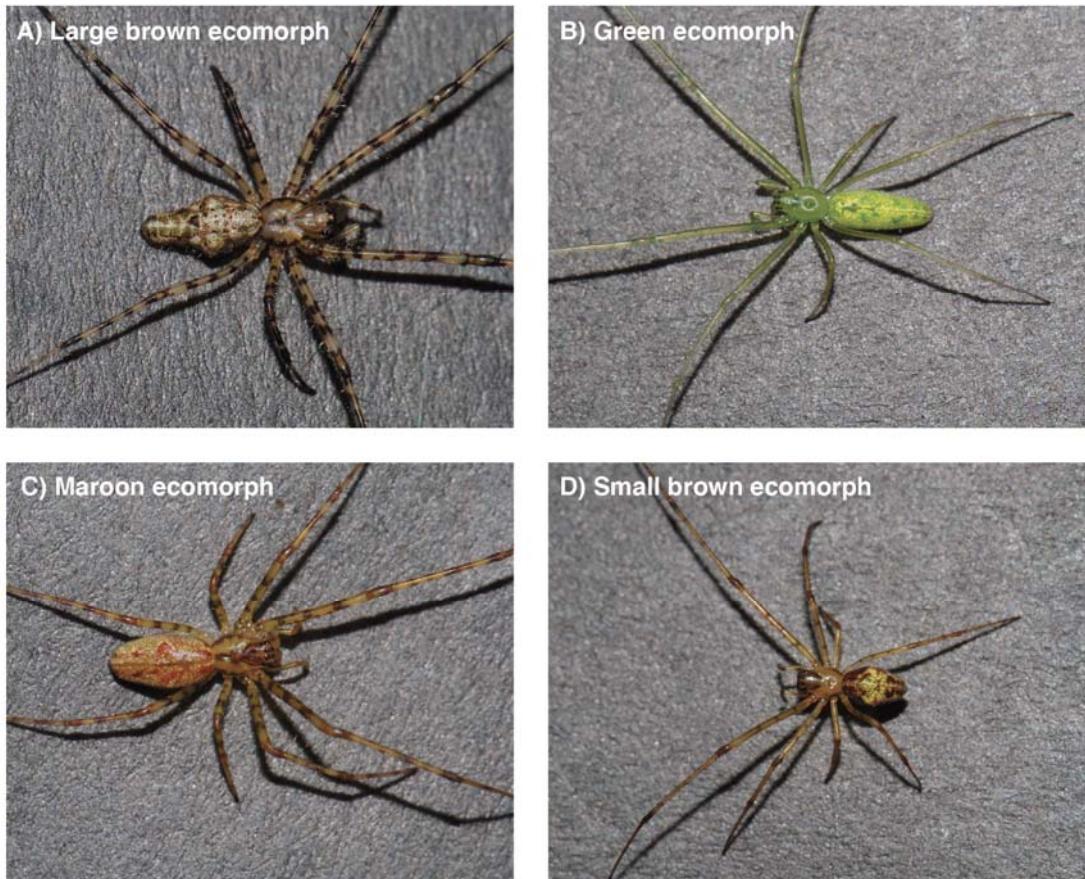
47 Maroon and the Small Brown ecomorphs likely involved ancestral hybridization events,
48 whereas the Green and the Large Brown ecomorphs likely evolved because of either standing
49 genetic variation or *de novo* mutation. Pairwise comparisons of ecomorphs based on the
50 fixation index (F_{ST}) show that divergent genomic regions include genes with functions
51 associated with pigmentation (melanization), learning, neuronal and synapse activity, and
52 circadian rhythms. These results show that the repeated evolution of ecomorphs in the
53 Hawaiian spiny-leg *Tetragnatha* is linked to multiple genomic regions and suggests a previously
54 unknown role of learning and circadian rhythms in ecomorph.

55 Introduction

56 Adaptive radiation, the evolutionary process where an ancestral lineage diversifies into
57 multiple phenotypically-distinct species which occupy different ecological niches, offers a
58 natural experiment to disentangle links between the phenotypic diversification and
59 environmental adaptation (Schluter 2000; Gillespie et al. 2020). Of particular interest in
60 adaptive radiation is the repeated phenotypic evolution (i.e. parallel or convergent phenotypic
61 evolution) (Cerca 2022), where equivalent phenotypes evolve in response to similar ecological
62 challenges (Losos and Ricklefs 2009; Losos 2010). Repeated evolution of similar phenotypes
63 has been found in multiple radiations and offers a powerful approach for disentangling
64 recurrent and potentially-deterministic phenotypic outcomes in response to similar
65 environmental conditions (Losos 2010, 2011; Gillespie et al. 2018, 2020; Malinsky et al. 2018;
66 Salzburger 2018; Masonick et al. 2022; Urban et al. 2022). For instance, the repeated evolution
67 of habitat-specialists, termed ecomorphs, has been reported in multiple adaptive radiations
68 including the Caribbean *Anolis* lizards (Losos and Ricklefs 2009) and Hawaiian *Tetragnatha*
69 (Gillespie 2004) and *Ariamnes* (Gillespie et al. 2018) spiders. In these cases, the repeated
70 evolution of ecomorphs has been explained based on the spatial segregation of environments, as
71 different islands offer similar environmental conditions (Losos and Ricklefs 2009; Losos 2010).
72 However, despite the deluge of genomic data, and new insights into phenomena of admixture,

73 epigenetics, and other phenomena that highlight the overall flexibility of the genome, we know
74 little about the genomic underpinnings involved in the evolution of discrete phenotypes that
75 appear to arise repeatedly in response to similar selective pressures. The current study set out
76 to examine the genomic basis of recurrent ecomorph evolution in a lineage of Hawaiian spiders.

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79 **Figure 1. *Tetragnatha* spiny-leg ecomorphs** including representatives of (A) the Large Brown
80 (*Tetragnatha quasimodo*), (B) the Green (*Tetragnatha brevignatha*), (C) the Maroon
81 (*Tetragnatha kamakou*), and (D) the Small Brown (*Tetragnatha anuenue*). Photographs by
82 Darko D. Cotoras.

83

84 The Hawaiian *Tetragnatha* spiny-leg species belong to a clade comprising ~17 species,
85 and are part of a large radiation endemic to the archipelago (Gillespie 2016; Kennedy et al.
86 2022). These species can be grouped into four ecomorphs (Figure 1 A-D), which are linked to
87 the substrate they inhabit (Gillespie 2004): the Large Brown ecomorph is found on tree bark
88 (Figure 1A), the Green ecomorph on leaves (Figure 1B), the Maroon ecomorph on mosses
89 (Figure 1C), and the Small Brown ecomorph on twigs (Figure 1D). Ancestral character-state
90 reconstructions suggest that the Green ecomorph is likely the ancestral form, having evolved
91 once, while the remaining ecomorphs each evolved twice (Gillespie 2004). Recent genomic
92 work has shown that co-occurring closely related species belonging to the Green ecomorph do
93 not hybridize, and it has been argued that there may be some overlap in their ecological niches
94 in the early stages of diversification, suggesting a possible avenue for the divergence of
95 ecomorphs through character displacement upon secondary contact (Schluter 2000; Cotoras et
96 al. 2018). The genomic bases of ecomorph evolution, which will allow for the ecological
97 differentiation of these closely related and ecologically equivalent species are still poorly
98 understood.

99 The repeated evolution of a phenotype can arise as a result of three non-mutually
100 exclusive genomic processes (Stern 2013; Pease et al. 2016; Lee and Coop 2017, 2019): *de novo*
101 mutation (different mutations causing similar phenotypes), shared ancestral polymorphism
102 (standing variation, old genetic variation being re-recruited), or hybridization (where a
103 particular allele is recruited from one lineage to another). All three patterns have been observed
104 in the context of adaptive radiations (Meier et al. 2018; Choi et al. 2021; Sowersby et al. 2021;
105 De-Kayne et al. 2022). Because they leave distinct footprints along the genome, genomic-level
106 data is able to tease apart the contribution of each of these processes (Lee and Coop 2017).

107 Here, we whole-genome re-sequence 76 genomes across the *Tetragnatha* spiny-leg
108 radiation with the aim of understanding the genomic basis of repeated ecomorph evolution. We
109 hypothesise that standing genetic variation has been the source for phenotypic repetition in
110 *Tetragnatha* spiny-legs. We start by reconstructing the evolutionary history of the spiny-leg

111 lineage using phylogenetic tools. Then, we explore patterns of excess allele sharing to infer
112 potential hybridization. Finally, we investigate genomic divergence (F_{ST}) to understand
113 differentiation between ecomorphs. Our results uncover a complex evolutionary history,
114 showing that repeated evolution may have emerged through multiple different genomic
115 processes.

116 **Methods**

117 **Field collection**

118 Spiny-leg *Tetragnatha* live in the montane rain forests of all the major Hawaiian islands
119 (1,200-1,800 meters) (Roderick et al. 2012). Specimens were collected either by hand (at night,
120 when spiders are active) or using a beating sheet (both day and night). Specimens were
121 preserved in 95% ethanol and stored at -20°C. A list of specimens including their volcano and
122 island is found in Supplementary Table 01.

123 **Molecular data generation**

124 We sequenced genomes from a total of 76 individuals (Supplementary Table 01). For
125 each specimen, we extracted DNA from 2-4 legs. Legs were first ground using a tube pestle and
126 incubated overnight in a solution of lysis buffer (10mM Tris pH, 100mM NaCl, 10mM EDTA,
127 0.5% SDS) and proteinase K at 54° C, we then followed the instructions of the commercial
128 provider Qiagen to extract the DNA. We diluted the DNA in 50 μ l of elution buffer and assessed
129 the DNA concentration of each of the 76 extracts using a Qubit fluorometer (ThermoFisher).

130 Samples with more than 500 ng of DNA were submitted to the QB3-Berkeley, Vincent J.
131 Coates Genomics Sequencing Laboratory at UC Berkeley, where library preparation was
132 conducted. DNA was fragmented using a Bioruptor Pico (Diagenode) and libraries prepared
133 using the KAPA Hyper Prep kit for DNA (KK8504). This involved adding truncated universal
134 stub adapters for DNA-adapter ligation, and indexed primers for PCR amplification (to complete

135 the adapters). The quality of the samples was checked on an AATI (now Agilent) Fragment
136 Analyzer, and the molarity of the library was measured using quantitative PCR with the KAPA
137 Library Quantification Kit (Roche KK4824) on a BioRad CFX Connect thermal cycler. Libraries
138 were then pooled by molarity and sequenced on an Illumina NovaSeq 6000 S4 flowcell for 2 x
139 150 cycles, with the aim of obtaining 10 Gbp per sample (~9.5x sequencing depth of the
140 genome). Raw sequencing data was converted to FASTQ format and demultiplexed using the
141 Illumina bcl2fastq2 software while allowing for up to one mismatch in the index sequences.

142 Bioinformatics data processing

143 All of the steps below are reported in [GITHUB JOSE](#). We checked the quality of the
144 sequencing data using fastQC v0.11.8 (Andrews 2017), and identified adapters using
145 AdapterRemoval v2.3.1 (--identify-adapters) (Schubert et al. 2016). We then used Trimmomatic
146 v0.39 (Bolger et al. 2014) to remove adapters and poor quality reads by specifying: maximum 2
147 mismatches in the adapter sequence; cut three base pairs in the beginning and the end of the
148 read if the quality drops below 20 in these regions; a sliding window of 4 bp with a minimum
149 quality threshold of 20, and a minimum read length of 50. These high-quality reads were aligned
150 to the *Tetragnatha kauaiensis* reference genome v1 (~1.1 Gb) (Cerca et al. 2021a) using the
151 Burrows-Wheeler Aligner v0.7.17, mem algorithm (Li and Durbin 2009). Because Trimmomatic
152 separates reads as forward-paired, reverse-paired, forward-unpaired, and reverse-unpaired, we
153 aligned all four files to the reference genome, and merged and sorted the final file using
154 Samtools v1.10 (Li et al. 2009). We then estimated mapping quality of each alignment using
155 samtools flagstat, finding no mapping-bias (Supplementary Table 01), and marked duplicates
156 using GATK 4.1.4.0 (McKenna et al. 2010), by running the SortSam algorithm which sorted
157 reads based on genomic coordinates, followed by the MarkDuplicates function. We then built
158 indexes for each alignment using BuildBamIndex (McKenna et al. 2010), and filtered reads with
159 a low mapping quality (MAPQ) using samtools. Specifically, we discarded reads with mapping
160 quality below 30, and calculated the sequencing depth using samtools depth (option -a was used

161 to output all sequencing positions, even those with 0 depth; Supplementary Table 01). We then
162 processed the data using ANGSD v0.935 (Kousathanas et al. 2017), a pipeline designed to
163 handle and analyze low coverage sequencing data.

164

165 Phylogenetic analyses

166 We started by doing a phylogenetic reconstruction while accounting for low-coverage to
167 understand the ancestral relationships of ecomorphs. As part of this analysis, we included five
168 specimens from four species belonging to an outgroup lineage, the *Tetragnatha* web builder
169 radiation, which is also endemic to Hawai'i (*Tetragnatha maka*, *Tetragnatha acuta*, *Tetragnatha*
170 *filiciphilia*, and *Tetragnatha stelarobusta*). We started by calling variants in ANGSD using the
171 GATK genotype likelihood model to output a beagle file (-GL 2 -doGlf 2), specifying that a
172 variant should be present in at least 38 individuals (half of the dataset), minimum base quality
173 of 30 (-minQ), a *p*-value threshold of 1e-6 (-SNP_pval), to remove allele counts in less than 5%
174 of the dataset (-minMaf 0.05), and to infer major and minor alleles from the genotype
175 likelihoods (-doMajorMinor 1). The output beagle file was then processed using an in-house
176 script to remove variants in repeat regions by downloading the general feature format file (gff3)
177 from (Cerca et al. 2021a), including the genomic location of repeat elements, converting it to a
178 bed file format using bedtools v2.26.0 (Quinlan and Hall 2010), and generating a white-list of
179 the genome (i.e. non-repeated genomic areas) using an in-house script (See Github). This
180 dataset was used as input in NgsDist (Vieira et al. 2015), where we specified a block size of 20
181 SNPs, and 100 bootstrap replicates. We then ran the NJ software fastme v2 (Lefort et al. 2015),
182 merged all the 100 replicates into a final tree with bootstrap support using RAXML while
183 specifying an optimization of branch-length, and specifying GTRCAT (GTR + Optimization of
184 substitution rates + Optimization of site-specific) as the substitution model.

185 Because different parts of the genome can have different topologies as a result of
186 different evolutionary mechanisms, we additionally performed a mitochondrial phylogenetic

187 reconstruction and a phylogenetic reconstruction based on Ultra Conserved Elements (UCE).
188 We retrieved mitochondrial genomes and did a tree reconstruction to analyse nuclear-
189 mitochondrial discordances; which are typically associated with hybridization events. We
190 started by extracting the mitochondrial genomes from the cleaned Illumina libraries using
191 Novoplasty v4.2 (Dierckxsens et al. 2017), a method which uses an iterative baiting-and-expand
192 approach to construct genomic regions. We obtained a collection of seeds encompassing
193 different mitochondrial genes (16S rDNA (hereafter 16S), Cytochrome Oxidase I (COI), and
194 Cytochrome B (CytB)) for species from the radiation (Kennedy et al. 2022). For each specimen,
195 we ran all seeds until we obtained a circularized mitochondrial genome. If no complete
196 mitochondrial genome was obtained for a given specimen, we took the longest contig obtained
197 by Novoplasty. In total, we obtained 46 complete fully-circularized mitochondrial genomes, and
198 19 partially complete mitochondrial genomes (>5,000 bp). We then concatenated and aligned
199 the 65 mitogenomes using Mafft v7.130b (Katoh and Standley 2013), trimmed the ends of the
200 alignments and ran a tree using IQ-Tree v2.0.3 specifying 1,000 ultrabootstrap replicates and
201 automatic determination of substitution models.

202 We obtained a UCE dataset to complement the ngsDist methods with a maximum
203 likelihood-based analysis. For each sample, we started by calling every position of the genome
204 using the Bcftools' v1.10.2 mpileup and call algorithms (Danecek et al. 2021), using the *T.
205 kauaiensis* genome v1 as reference. The output files were normalized for indels using bcftools
206 norm, and filtered for gaps bigger than 4 base pairs, base quality above 20, depth above 1 and
207 below 30. These files were then turned into consensus fasta sequences using bcftools consensus.
208 To extract UCEs from these consensus files, we used the phyluce pipeline (Faircloth 2016), with
209 the 'Arachnida-UCE-1.1K-v1' set as bait. Phyluce allows extraction of UCEs, but because of the
210 missing data due to low coverage and fragmented genomes, we were only able to retrieve 29
211 UCEs that were present in half of the dataset (16 of the UCEs were present in all individuals; and
212 23 UCEs present in ≥ 70 individuals).

213 Partitioning of genetic variation and hybridization

214 After reconstructing species and ecomorph relationships we analysed the partitioning of
215 genetic variation and whether hybridization has occurred in the radiation. We studied the
216 partitioning of genetic variation by running a NGSAdmix analysis. This analysis required
217 processing the data with ANGSD, similar to the one used for the phylogenetic reconstruction
218 with NgsDist, but with three exceptions. Specifically, we included a minimum base quality of 20
219 (-minQ) filter, removed variants with a minimum allele frequency of 0.05 (-minMaf 0.05), and
220 inferred major and minor alleles (-doMajorMinor 1). After filtering for variants in repeat
221 regions, we filtered the data for severe deviations from the Hardy-Weinberg equilibrium (HWE)
222 using PCAngsd, which estimated a likelihood ratio test (LTR) to identify sites where severe
223 deviations of the HWE occur (LTR >24, following recommendations from the contributors of
224 ANGSD; <https://github.com/GenisGE/grantsGazelleScripts>) (Garcia-Erill et al. 2021).
225 Additionally, given that admixture-like analyses assume that variants are independent, we
226 implemented a linkage disequilibrium (LD) filter. This was done by calculating LD in windows
227 of 100,000 bps using plink (Purcell et al. 2007) and using an in-house script explore LD-decay.
228 We LD-pruned the data by removing genomic regions where linkage, measured as R2, was
229 above 0.11 in windows of 25,000 bps (5,000 bp steps). We then ran NGSadmix analyses (Skotte
230 et al. 2013). We specified cluster values (K) between 1-20, and for each we performed 10
231 independent runs until each reached chain convergence. We estimated the best K to be 2 using
232 the Evanno method (Evanno et al. 2005) as implemented as part of CLUMPAK (Kopelman et al.
233 2015). We present the results for $K=2$, 3, and 15 as ADMIXTURE plots help to clarify how
234 genetic variation is partitioned at different levels (Meirmans 2015).

235 Considering the nuclear-mitochondrial discordance, we calculated excess of shared
236 alleles in the dataset. To do so, we re-ran ANGSD, with the parameter specification as adone for
237 the NGSadmix analyses, further specifying a genotype likelihood depth filter between 7 and 30
238 (-geno_minDepth 7 -geno_maxDepth 30). The outputted genotype file was processed to remove
239 variants in repeat regions and the excess allele sharing was calculated using Patterson's D

240 (ABBA-BABA) and F_4 ratios (Patterson et al. 2012) using Dsuite (Malinsky et al. 2021). Both
241 these analyses benefited from a tree-backbone and we specified the tree obtained in ngsDist.

242 Scans of selection

243 Whole-genome level data allows understanding whether certain genes or genomic
244 regions are under selection as this process leads to shifts in allelic frequencies (Ravinet et al.
245 2017). In order to calculate population differentiation (F_{ST}), we obtained the Site Frequency
246 Spectrum (SFS) using ANGSD. As our goal is to understand the evolution of ecomorphs, we
247 analysed the data for each monophyletic ecomorph group, based on the nuclear phylogenetic
248 reconstruction obtained with NGSdist. Furthermore, because SFS and F_{ST} estimations require
249 allelic frequencies from multiple individuals, we selected only ecomorphs with 5 or more
250 specimens (Supplementary Table 01). To obtain SFS estimates we ran ANGSD separately for:
251 *Tetragnatha pilosa* (Large Brown group A), *T. quasimodo* (Large Brown group B), *T. kauaiensis*
252 (Green group A), *T. tantalus*, *T. polychromata*, *T. brevignatha*, *T. waikamoi* (Green group B), *T.*
253 *obscura*, *T. kukuiki*, *T. kikokiko*, *T. anuenue* (Small Brown ecomorph), and *T. kamakou* (Maroon
254 ecomorph). We ran ANGSD independently for each of these groups, specifying that loci had to be
255 in at least 5 individuals, keeping only reads with a single mapping (-uniqueOnly), discarding bad
256 reads (-remove_bads), and keeping variants with a minimum base quality of 20. This yielded a
257 1-dimensional SFS for each population, which was then processed using realSFS (using -fold 1)
258 to obtain 2-dimensional SFSs for the following pairs: the two Large Browns (Large Brown A vs
259 Large Brown B), the two Greens (Green A vs Green B), Green vs Maroon (Green B vs Maroon),
260 and Small Brown vs Large Brown (Small Brown vs Large Brown B). For each of the 2-
261 dimensional SFS, we ran the realSFS F_{ST} index algorithm, followed by realSFS F_{ST} stat to obtain
262 the overall F_{ST} . Because we were interested in identifying genomic areas of divergence and loci
263 within, we ran the realSFS F_{ST} stat2 algorithm to obtain F_{ST} in 5,000 bp non-overlapping windows.
264 We identified regions of differentiation by Z-transferring F_{ST} and using a Z cut-off of >3. After
265 identifying significantly differentiated regions, we retrieved genes for these windows from the

266 gene annotations of the *T. kauaiensis* genome (Cerca et al. 2021a). We then called orthologs
267 between *T. kauaiensis* and the *Drosophila melanogaster* genome annotations to obtain evidence
268 on gene function, benefiting from the decades of functional genetic research on the latter, using
269 *OrthoFinder2* (Emms and Kelly 2015). After identifying the closest *D. melanogaster* ortholog for
270 a particular gene on a region with high differentiation, we reviewed and summarized the
271 literature on that gene in Supplementary Table 2. When possible, we read at least three papers
272 for each gene.

273 Hybridization along the genome

274 We found evidence for hybridization, and because this process can leave tracks along
275 the genome, we performed an analyses to understand the breadth and impact of hybridization
276 on the genomes. Specifically, the results showed a strong signal of admixture between *T.*
277 *kamakou* (Maroon ecomorph), *T. perreirai* (Maroon ecomorph), and *T. restricta* (light Brown
278 ecomorph) and to understand whether hybridization occurred we did a TWISST analysis on the
279 three biggest scaffolds and D-suite investigate for the scaffolds with melanization genes. For the
280 TWISST analysis, we selected *T. anuenue*, *T. perreirai*, *T. restricta*, and *T. mohihi*. We started by
281 running ANGSD for non-overlapping regions of 5kb segments of the genome, specifying a
282 minimum map quality of 30, minimum base quality of 20, outputting the frequency of the
283 different bases. We then did a NJ-tree for each of these regions using the R package ape (Paradis
284 et al. 2008). We then ran TWISST specifying *T. anuenue*. For the windowed Patterson's D, we
285 used Dsuite investigate on the scaffolds with melanization genes.

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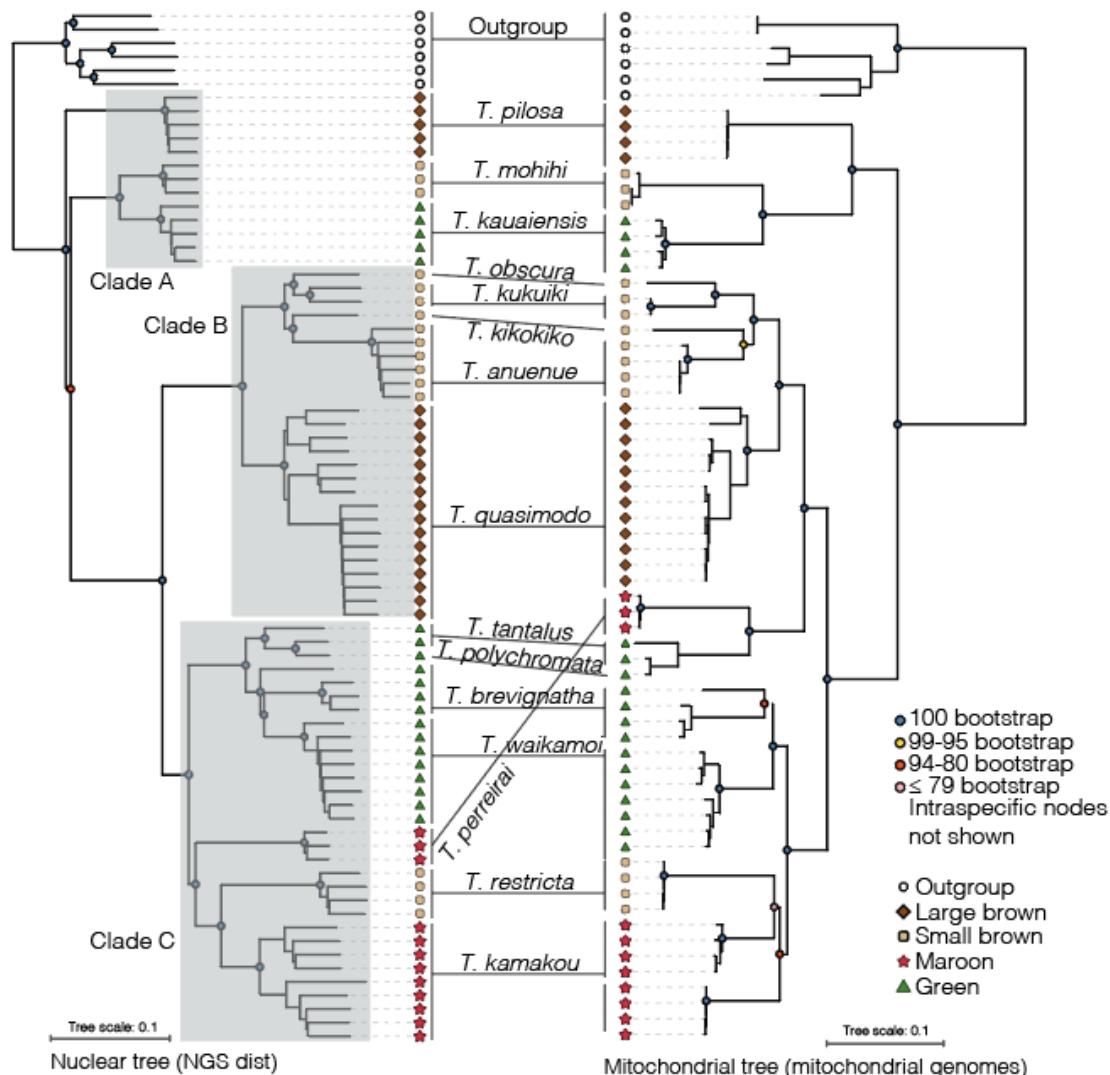
287 Results

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Tree reconstruction

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Figure 2. Nuclear and mitochondrial tree reconstructions. The nuclear tree (left) obtained by NgsDist is based on 1,848,915 variable sites. The mitochondrial tree (right) obtained by NovoPlasty and IQ-Tree is based on complete and partial mitochondrial genomes. Bootstrap is provided for each inter-specific node, and the tips of the phylogeny include ecomorph and outgroup labelling as coloured symbols. Clades A-C are plotted in the nuclear tree.

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297 In the nuclear tree, *T. pilosa* is sister to the remaining spiny-leg species. The node
298 including all other spiny-legs, however, has a bootstrap support below 95, while all other
299 interspecific nodes have a bootstrap support of 100. Furthermore, the node including all spiny-
300 legs but *T. pilosa* is preceded by a very short branch, and separates two clades: one clade
301 including *T. mohihi* and *T. kauaiensis*, and another clade with all remaining species. The
302 remaining species can be separated in two major clades, the first (clade B) including the Large
303 Brown *T. quasimodo* and a group of Small Brown species including *T. obscura*, *T. kukuiki*, *T.*
304 *kikokiko*, *T. anuenue*; and the second (clade C) including a group of species representing the
305 Green ecomorph (*T. tantalus*, *T. polychromata*, *T. waikamoi*, *T. brevignatha*), which is sister to a
306 group comprising two Maroon species (*T. perreirai* and *T. kamakou*) and a Small Brown species
307 (*T. restricta*). The maximum-likelihood UCE tree is topologically concordant with the ngsDist
308 tree at the ecomorph level (Supplementary Figure 01). The only topological discordance is the
309 species' placement within the Small Brown group as part of clade B (*T. obscura*, *T. kukuiki*, *T.*
310 *kikokiko*, *T. anuenue*). Specifically, the ngsDist tree (Figure 2) shows *T. kukuiki* as sister to *T.*
311 *obscura*, and *T. kikokiko* as sister to *T. anuenue*, whereas on the UCE tree, *T. anuenue* is sister to
312 a clade including all the aforementioned species (Supplementary Figure 01).

313 There are mitochondrial-nuclear tree discordances based on topology and bootstrap
314 support (Figure 2). First, in the mitochondrial tree, *T. pilosa* groups together with *T. mohihi* and
315 *T. kauaiensis*, forming a clade while *T. pilosa* branches separately in the nuclear tree. Notably, all
316 basal interspecific nodes have a bootstrap support of 100. Second, *T. tantalus* and *T.*
317 *polychromata* (Green ecomorphs) are sister to *T. perreirai* (Maroon ecomorph) in the
318 mitochondrial tree, and the clade comprising these three species is sister to the clade including
319 the Large Brown (*T. quasimodo*) and the Small Brown ecomorphs (*T. obscura*, *T. kukuiki*, *T.*
320 *kikokiko*, *T. anuenue*). These also have 100 bootstrap support at the relevant interspecific
321 nodes. The mitochondrial tree shows that the evolution of Green ecomorphs has occurred three
322 times (1 - *T. kauaiensis*; 2 - *T. tantalus* and *T. polychromata*; 3 - *T. waikamoi* and *T. brevignatha*),
323 as opposed to two clades in the nuclear tree (1 - *T. kauaiensis*; 2 - *T. tantalus*, *T. polychromata*,

324 *T. waikamoi* and *T. brevignatha*). Third, the mitochondrial dataset shows *T. kamakou* as
325 paraphyletic, separating the populations from Moloka'i and Maui (East Maui volcano).
326 However, this node is weakly supported as it has a bootstrap value below 80.

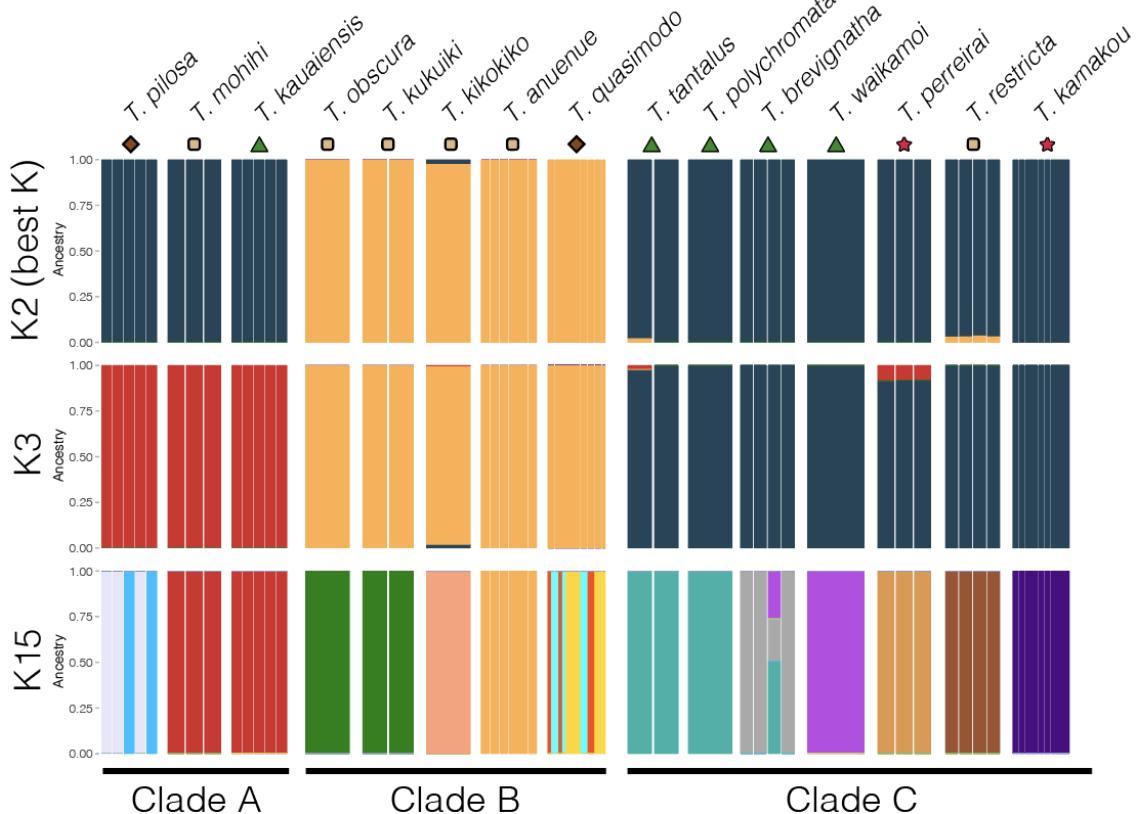
327 For simplicity, and based on the mitochondrial phylogenetic reconstruction (and
328 ngsAdmix below), we refer to clades A (*T. kauaiensis*, *T. pilosa*, *T. mohihi*), B (*T. anuenue*, *T.*
329 *obscura*, *T. kukuiki*, *T. kikokiko*, *T. quasimodo*) and C (*T. kamakou*, *T. restricta*, *T. perreirai*, *T.*
330 *tantalus*, *T. polychromata*, *T. waikamoi* and *T. brevignatha*) hereafter.

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333 Structuring of genetic diversity

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336 **Figure 3. Structure of genetic diversity** NgsAdmix analysis with cluster (K) values of 2 (best K), 3
337 (major clades), and 15 (number of species).

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339 The best K obtained by the Evanno method was $K= 2$ (Figure 3). This K separated clade
340 B from the remaining clades (A and C), while $K= 3$ separated all three clades obtained in the
341 phylogeny. There is evidence of a shared history for both K values, some species having minor
342 components of other clades. For instance, *T. kikokiko* (Clade B) has a minor component of
343 ancestry from clades A-C on $K= 2$, and a minor component from clade C on $K= 3$. We also
344 display $K= 15$ because that is the number of species in the dataset (Figure 3). For $K = 15$, some
345 species are assigned to the same genetic cluster: *T. kauaiensis* and *T. mohihi* (red), *T. obscura*
346 and *T. kukuiki* (green), and *T. polychromata* and *T. tantalus* (aqua-blue). Two species,
347 interestingly both Large Browns, have intra-specific population structure (*T. pilosa* has two
348 colours, while *T. quasimodo* has three). Only one *T. brevignatha* sample has mixed ancestries, all
349 in common with species belonging to the Green ecomorph.

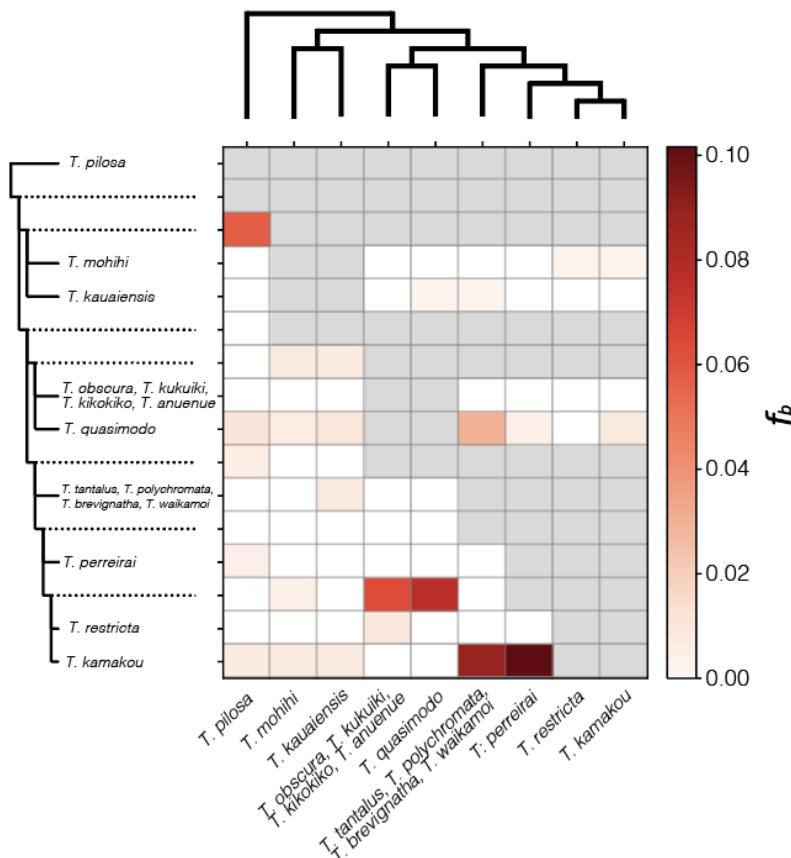
350

351

352

353

Excess allele sharing



354

355 **Figure 4. Excess allele sharing** F4-branch statistic plotted as a heatmap. The tree topology
356 obtained by NgsDist and UCEs is plotted above and on the left every branch of the tree is
357 displayed (including internal branches). Because the goal is to understand hybridization
358 between ecomorphs, clades of closely related species belonging to the same ecomorph (i.e.
359 monophyletic ecomorphs) are displayed.

360

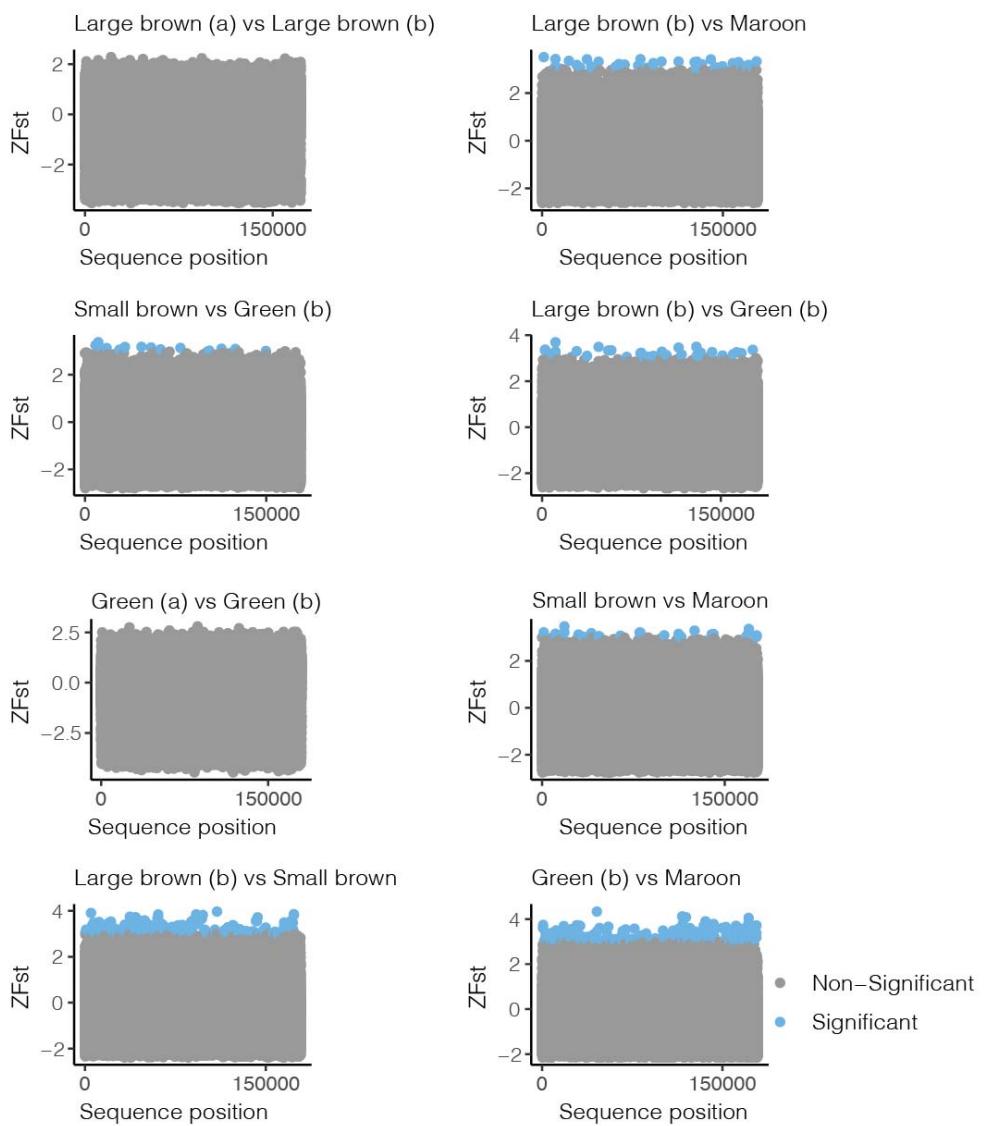
361 Based on Patterson's D statistics (Supplementary Figure 02) and F4-branch statistics,
362 we found excess allele sharing between ecomorphs and within ecomorphs (Figure 4). We
363 detected the largest excess in allele sharing between the Maroon ecomorph species (*T. perreirai*
364 and *T. kamakou*; the trio P1 = *T. restricta*; P2 = *T. kamakou*; P3 = *T. perreirai* had a D-statistic

365 of 0.15, Z-score 30.42 over >150,000 sites). In the Small Brown ecomorph, we also detected
366 excess allele sharing, specifically between *T. restricta* and the remaining species (Figure 4; the
367 trio P1 = *T. perreirai*; P2 = *T. restricta*; P3 = *T. obcura-kukuiki-kikokiko-anuenue* trio had a D-
368 statistic of 0.13, Z-score of 8.21; >150,000 sites; Supplementary Table 03). No evidence for
369 excess allele sharing was found between Green and Large Brown ecomorphs (Figure 4). When
370 comparing ecomorphs, we find evidence of excess allele sharing between the lineages of clade A,
371 namely between *T. pilosa* (Large Brown ecomorph), and *T. kauaiensis* and *T. mohihi* (Small
372 Brown ecomorph); there is also excess of allele sharing between the Green ecomorph group (*T.*
373 *tantalus*, *T. polychromata*, *T. brevignatha*, *T. waikamoi*) and one of the Maroon ecomorph
374 species (*T. kamakou*). The Large Brown species *T. quasimodo* had excess allele sharing with the
375 clade consisting of *T. restricta* (Small Brown ecomorph) and *T. kamakou* (Maroon ecomorph).

376 The windowed-analyses of D-statistics and TWISST showed no particular pattern of
377 introgression for *T. restricta*, *T. kamakou* and *T. perreirai*. Specifically, D-statistics of regions
378 where melanization genes were found (see F_{ST} scans below) were not elevated. The TWISST
379 analyses showed no specific tracks of hybridization in the largest three scaffolds
380 (Supplementary Figures 3-5).

381

F_{ST} scans



382

383 **Figure 5. F_{ST} Scans of genomic divergence** For each pairwise comparison, we provide the
384 sequence position and a Z-transformed F_{ST} value. Significant Z values are displayed in blue ($Z >$
385 3) and non-significant in grey. Large Brown (a) includes *T. pilosa* (5 individuals); Large Brown
386 (b) includes *T. quasimodo* (16 individuals); Green (a) includes *T. kauaiensis* (5 individuals);
387 Green (b) includes *T. talanus*, *T. polychromata*, *T. brevignatha*, *T. waikamoi* (15 individuals);
388 Small Brown includes *T. obscura*, *T. kukuiki*, *T. kikokiko*, *T. anuenue* (10 individuals); Maroon
389 includes *T. kamakou* (9 individuals).

390 We ran F_{ST} comparisons for lineages with five or more individuals (Figure 5). The two
391 within-ecomorph comparisons (i.e. Large Brown (a) *vs* Large Brown (b); Green (a) *vs* Green
392 (b)) yielded no significant F_{ST} outliers, and this is likely attributed to the overall high F_{ST}
393 between more distantly related species in the radiation (Large Brown (a) *vs* Large Brown (b)
394 had an global F_{ST} of 0.60; Green (a) *vs* Green (b) had an global F_{ST} of 0.61). The remaining
395 comparisons, displayed in Figure 5, all had overall F_{ST} between 0.31 (Green (b) *vs* Maroon) -
396 0.41 (Small Brown *vs* Maroon).

397 Several areas of divergence along the genome seem to be common in pairwise
398 comparisons. Specifically, while significant F_{ST} can be driven by a highly divergent region in a
399 single lineage, we found that multiple regions of the genome seem to be commonly
400 differentiated. For example, we found twelve genomic regions where significant F_{ST} values were
401 identified in different pairwise comparisons. As an example, the region between 10,000 - 15,000
402 bp on Scaffold-2,488 had F_{ST} outliers when comparing Large Brown (b) *vs* Green (b) and Small
403 Brown *vs* Maroon.

404 **Table 1** Summary of genes in areas with high genomic divergence (F_{ST}).

Comparison	Gene	Function	References
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Large Brown (b) vs Green (b)	NAT1	Circadian rhythm	(Bradley et al. 2012)
Large Brown (b) vs Green (b)	ALT	Neuron feeding source	(Volkenhoff et al. 2015)
Large Brown (b) vs Green (b)	Dscam	Immune-response associated	(Nazario-Toole et al. 2018)
Large Brown (b) and Small Brown	Mal-B1	Starch digestion	(Wu et al. 2016)
Large Brown (b) and Small Brown	vhaSFD	Longevity and life span	(Paaby and Schmidt 2009; Proshkina et al. 2015)
Large Brown (b) and Small Brown	Ank	Synaptic activity	(Koch et al. 2008)
Large Brown (b) and Small Brown	NAT1 and Gnpnat	Circadian rhythm	(Mattila and Hietakangas 2017)
Large Brown (b) and Small Brown	CG8483, and anon-WO0118547.80	Melanization-associated	(Bastide et al. 2016)

Large Brown (b) and Small Brown	CG8483	Posterior development genes	(Ibrahim et al. 2013)
Large Brown (b) and Maroon	ND-B17 and ALT	Synapsis	(Ghosh 2013)
Large Brown (b) and Maroon	NAT1, MFS10, and Period circadian protein	Circadian rhythm	(Curtin et al. 1995; Fernandez-Chiappe et al. 2021)
Large Brown (b) and Maroon	LIMK1	Learning and memory formation	(E. A. Nikitina, A. V. Medvedeva, Yu. F. Dolgaya, L. I. Korochkin, G. V. Pavlova & E. V. Savvateeva-Popova 2012)
Green (b) and Maroon	CG18473	Growth-regulation	(Hevia et al. 2017)
Green (b) and Maroon	Syp, CG8545, and Trpgamma	Synapses, neurojunctions, neurons and nervous-system related	(Guan et al. 2005; Liebl et al. 2006; Yang et al. 2017)
Green (b) and Maroon	LIMK1	Learning and memory formation	(E. A. Nikitina, A. V.

			Medvedeva, Yu. F. Dolgaya, L. I. Korochkin, G. V. Pavlova & E. V. Savvateeva-Popova 2012)
Green (b) and Maroon	MSF10, Jhedup	Circadian behaviour and rythm	(Bloch et al. 2013; Fernandez-Chiappe et al. 2021)
Green (b) and Maroon	anon- WO0118547.8 0	Melanization- associated	(Bastide et al. 2016)
Green (b) and Maroon	SNX3	Development and morphogenesis	(Zhang et al. 2011)
Green (b) and Maroon	RpS8	Extention of lifespan	(Hoffmann et al. 2013)
Green (b) and Maroon	CG11471	Smell-sensory functions	(Anholt et al. 2001)
Small Brown and Maroon	anon- WO0118547.8 0	Melanization- associated	(Bastide et al. 2016)

Small Brown and Maroon	Per	Circadian rhythm	(Curtin et al. 1995)
Small Brown and Maroon	LIMK1	Learning and memory formation	(E. A. Nikitina, A. V. Medvedeva, Yu. F. Dolgaya, L. I. Korochkin, G. V. Pavlova & E. V. Savvateeva-Popova 2012)

405

406 The regions with significant genomic divergence yielded some interesting genes (Table
407 1; Supplementary Table 02). Notably, we found genes associated with circadian rhythms in all
408 five comparisons, and genes associated with neuronal functions in four comparisons, and
409 learning and memory formation in three comparisons (Table 1). Interestingly, we found one
410 gene associated with diet (Starch digestion) and another with smell. Two genes were associated
411 with longevity and life-span, and three genes were associated with growth or development
412 (Table 1). Three of the comparisons included melanization-associated genes.

413 Discussion

414 The three main findings of our work are: *(i)* The repeated evolution of *Tetragnatha*
415 spiny-leg ecomorphs involves more than a single genomic-basis; *(ii)* The genomic-basis of
416 ecomorph differentiation included genes associated with melanization, agreeing with previous
417 ecological and morphological evidence which reported differences in colouration (Gillespie

418 2004; Kennedy et al. 2022); and *(iii)* pairwise scans of divergence uncovered evidence for a new
419 axis of differentiation between ecomorphs involving neural and circadian rhythm changes.

420 **Evolutionary history of the *Tetragnatha* spiny-leg clade**

421 Evidence of repeated evolution of ecomorphs comes from the paraphyletic and
422 polyphyletic positioning of ecomorph species in the nuclear whole-genome phylogeny (Figure
423 2), and partitioning of genetic variation in admixture analyses (Figure 3). These analyses show
424 that the spiny-leg radiation can be divided into three large clades. The first clade comprising the
425 lineages in Kaua'i, the oldest island, and including three ecomorphs, namely Green (*T.*
426 *kauaiensis*), Small Brown (*T. mohihi*), and Large Brown (*T. pilosa*). This clade is monophyletic
427 and resolved with high confidence (bootstrap of 100) in the mitochondrial-genome tree (Figure
428 2), corroborating previous results from both nuclear and mitochondrial DNA (Pons and
429 Gillespie 2003; Gillespie 2004). However, the whole-genome data suggests that *T. pilosa* is sister
430 to the remaining radiation, grouping *T. kauaiensis* and *T. mohihi* as its own clade. The branches
431 for *T. kauaiensis*, *T. mohihi* and *T. pilosa* are short, and this could indicate a lower substitution
432 rate, higher missing data, incorrect attachment point of the outgroups, or root misplacement
433 due to artefacts such as low-branch attraction. Lower substitution rate seems the most plausible
434 of these scenarios as missing data can be discarded considering the filtering of the data
435 (Supplementary Table 01). Attachment of outgroups follows the taxonomic knowledge of the
436 group (Kennedy et al. 2022). Root misplacement due to NJ-algorithms, which can be prone to
437 artefacts such as long-branch attraction (Susko et al. 2004), can be indirectly excluded as the
438 UCE-based tree displays a similar topology for these branches and was done using ML. The
439 nuclear-mitochondrial discordance pattern may result from hybridization (Figure 4) and from
440 the fast diversification of the radiation, which can lead to an increase of rates of incomplete
441 lineage sorting (Suh et al. 2015, Cerca et al. 2021b). Evidence for a fast diversification comes
442 from the short branch and the low bootstrap support of the corresponding node (orange circle;
443 Figure 2).

444 The second clade includes the widely distributed Large Brown species *T. quasimodo* and
445 the Small Brown group including *T. obscura*, *T. kukuiki*, *T. kikokiko*, and *T. anuenue*, whose
446 constituent species are present on the islands of Hawai'i, O'ahu, Maui and Moloka'i. The
447 nuclear and the mitochondrial lineages show topological congruence (Figure 2B). The third
448 clade includes a group of Green ecomorph species (*T. tantalus*, *T. polychromata*, *T. waikamoi*,
449 and *T. brevignatha*) which are distributed on several islands (Hawai'i, O'ahu, Maui), two
450 separate lineages with Maroon ecomorphs (*T. kamakou* from Moloka'i and Maui; *T. perreirai*
451 from O'ahu), and a Small Brown ecomorph species (*T. restricta*). In this clade, we found
452 evidence for nuclear-mitochondrial disagreements (Figure 2 A-B) and likely hybridization
453 (Figure 4). Most prominently, part of the group comprised by Green ecomorphs, namely *T.*
454 *tantalus* and *T. polychromata*, and the Maroon *T. perreirai*, nests as sister to the second clade in
455 the mitochondrial tree, rendering this clade paraphyletic in the mitochondrial dataset.

456 The Green ecomorph species *Tetragnatha macracantha* (Haleakalā volcano in Maui, and
457 Lana'i) was not included due to the lack of specimens. However, based on previous studies
458 (Gillespie 2004; Cotoras et al. 2018) it clusters with the other Green ecomorph species present
459 in Maui (*T. waikamoi* and the Maui population of *T. brevignatha*). Therefore, the described
460 patterns most likely also include this species. Similarly, *Tetragnatha kukuhaa*, a Small Brown
461 from the Big island is not-included, but it has been previously suggested as a sister to *T. obscura*
462 (Gillespie 2004; Kennedy et al. 2022).

463

464 **Genomic basis of ecomorph evolution**

465 We find evidence that the evolution of ecomorphs has happened repeatedly, in
466 agreement with previous works (Gillespie 2004), and extend these findings by finding evidence
467 that multiple genomic sources underlie repeated phenotypic evolution. Specifically, we find
468 evidence for excess allele sharing, which usually results from hybridization. The strongest signal
469 for hybridization in the dataset, where as much as 10% of the whole-genome variants may be
470 introgressed, occurs from *T. perreirai* to *T. kamakou*, the two species belonging to the Maroon

471 ecomorph (Figure 4). This evidence is further corroborated by the nuclear-mitochondrial
472 disagreements (Figure 2). Noticeably, *T. perreirai* and *T. kamakou* are not sister species in any
473 of the phylogenetic analysis carried by us (Figure 1; Supplementary Figure 01) and do not
474 currently overlap geographically (*T. perreirai* occurs on the island of O‘ahu, *T. kamakou* on
475 Maui and Moloka‘i). Several scenarios that could have led to the lack of monophyly of this
476 group in the face of hybridization.

477 First, introgression may have occurred between *T. perreirai* and an ancestral lineage
478 that was present on Maui or Moloka‘i. This introgression event may have carried adaptive
479 alleles associated with the Maroon ecomorph, and opened the Maroon-niche to the admixed
480 lineage, which is now *T. kamakou*. This event of introgression may have been subtle as our
481 exploration of genomic windows for the largest scaffolds (TWISST) and for genomic windows
482 ($F\delta$ ratio) where we found genes associated with melanization (CG8483, anon-W00118547.80),
483 yielded no evidence for introgression. It is possible that we missed signals of introgression due
484 to the sparsity of low-coverage data along the genome, since missing data may blur genomic
485 patterns and signals (Cerca et al. 2021b). Additionally, we cannot exclude recurrent ancestral
486 hybridization, and this may blur the distinction between hybridization and standing genetic
487 polymorphism (Ferreira et al. 2021). In any case, the admixture analyses suggested no shared
488 ancestry tracks between these three species, even at $K= 15$, and this indicates that signatures of
489 introgression may have been diluted through time.

490 The second scenario is that the Maroon phenotype evolved first in *T. kamakou*, and was
491 introgressed to O‘ahu, leading to the evolution of *T. perreirai*. This alternative scenario implies
492 a back colonization of O‘ahu, which is older than Maui Nui. At this stage, we cannot confidently
493 ascertain any of these scenarios, and further data including more genomes, higher coverage, and
494 analyses such as demographic simulations are necessary. We speculate that the combination of
495 ancestral bouts of introgression, fast diversification creating incomplete lineage sorting,
496 changes in population size, population-extinctions as a result of island-cycles, as observed in

497 other adaptive radiations (Meier et al. 2018; De-Kayne et al. 2022), limits our capacity to
498 understand the evolution of this group.

499 We also find evidence of excess allele sharing between members of the Small Brown
500 ecomorph. Namely, there is a signal of hybridization between *T. restricta*, *T. mohihi*, and the
501 clade comprising 4 Small Brown ecomorph species (*T. anuenue*, *T. kukuiki*, *T. obscura*, *T.*
502 *kikokiko*). This signal is particularly clear from the Patterson's D analysis (Supplementary
503 Figure 02), but less clear from the F4 ratio test (Figure 4). This mismatch is likely explained by
504 the different topologies underlying both tests. Two scenarios can explain these results. First,
505 considering that we find evidence for excess lineage sorting between lineages on different
506 islands, it can be speculated that dispersal may have occurred at some time in the past, leading
507 to hybridization. Second, hybridization is not the only cause leading to excess allele sharing and
508 this can confound Patterson's D and F4 -ration analyses. For example, the occurrence of
509 speciation between lineages with different population sizes may cause asymmetries that lead to
510 elevated Patterson's D statistics (smaller populations will have less variants due to the effect of
511 drift). Similarly, if multiple alleles underlie the evolution of an ecomorph, there may be
512 imbalances in allelic variation along the genome, confounding patterns of allele excess. While
513 scenarios of population-size differences and multi-alleles are hard to rule out with low-coverage
514 data, they are unlikely as we also observe mitochondrial-nuclear discordances. Our results
515 contribute to the growing body of work showing that introgression can be a potent force in the
516 passage of adaptive alleles from one lineage to another (Meier et al. 2017; Marques et al. 2019;
517 Sowersby et al. 2021) and can thereby drive repeated phenotypic evolution in the context of an
518 adaptive radiation.

519 Despite the evidence of hybridization in the spiny-leg lineage, hybridization alone is not
520 sufficient to explain the repeated evolution of every ecomorph. Specifically, we do not find
521 evidence for excess allele sharing between the two Green ecomorph clades (*T. kauaiensis* and
522 the clade comprising the lineages *T. brevignatha*, *T. waikamoi*, *T. tantalus*, *T. polychromata*) and
523 a weak signal of hybridization between *T. pilosa* and *T. quasimodo*. The lack of hybridization

524 within the Green ecomorph is consistent with previous studies, which found no hybridization
525 between Green ecomorph species (Cotoras et al. 2018). The repeated evolution of some
526 ecomorphs may have occurred by either *de novo* mutation or ancestral genetic variation
527 (Barrett and Schluter 2008). Nonetheless, distinguishing between these scenarios benefits from
528 the study of sweeps and coalescence of variants (Barrett and Schluter 2008; Lee and Coop
529 2017), which would require higher coverage and wider population sampling obtained by us.
530 Despite these limitations, several pieces of evidence may point towards a role of repeated
531 phenotypic evolution of Green and the Large Brown ecomorphs through standing genetic
532 variation. First, the repeated recruitment of variation may be more likely than repeated *de novo*
533 mutations producing the same phenotype (Barrett and Schluter 2008). Second, as discussed
534 above, the phylogenetic reconstruction includes some short branches, which drives higher rates
535 of incomplete lineage sorting, translating into standing genetic variation passing down the
536 lineages. Third, we found 12 genomic regions that were repeatedly significantly-diverged in
537 pairwise F_{ST} comparisons. However, we are careful not to exclude the existence of *de novo*
538 mutation in the radiation and in specific genomic regions, which could be hotspots of mutation
539 with some adaptive value; e.g. (Xie et al. 2019).

540 Because adaptive introgression seems to play a role in the evolution of some ecomorphs
541 (Small Brown and Maroon), but not in others (Green and Large Brown), this opens the question
542 as to how easy it becomes to re-evolve certain traits without hybridization. In other words, is
543 the transfer of some key genes a strict requirement for the appearance of the Small Brown and
544 Maroon ecomorphs? If that were to be the case, this could explain why the Maroon ecomorph is
545 entirely absent from Lana'i and the Big Island (Gillespie 2004).

546 **Ecological and genomic drivers of ecomorph evolution**

547 Spiny-leg *Tetragnatha* spiders are largely confined to mid-elevation wet and mesic
548 forests on Hawaiian volcanoes (1,200-1,800 meters) and diversification has occurred largely
549 within this environment (Hiller et al. 2019). This suggests that, in addition to the physical
550 barrier between islands imposed by the ocean, the lowland area may act as a strong isolating

551 barrier, as has been shown for many taxa in more stable environments (Janzen 1967). The
552 spiders' forest habitats have mosaic distributions, and it is likely that the separation between
553 them has triggered a dynamic interplay between natural selection in response to micro-habitat
554 availability and allopatric divergence (Vandergast et al. 2004; Roderick et al. 2012; Cotoras et al.
555 2018). In particular, ecomorph divergence is likely due to allopatric establishment on different
556 islands and volcanoes, followed by secondary contact, competition between ecologically similar
557 species and hence the accumulation of divergence (Schluter 2000; Cotoras et al. 2018).
558 Dispersal and secondary contact of diverged populations may have led to a macroevolutive
559 ecological character displacement and the evolution of particular ecomorphologies. This
560 scenario is consistent with the evidence for hybridization at various points of the radiation,
561 which suggests that dispersal may have occurred multiple times across the phylogeny, leading
562 to the introgression of variants.

563 A major goal of adaptive radiation research is to disentangle the ecological drivers of
564 species formation (speciation) and adaptation to the environment (local adaptation). Previous
565 research has suggested that ecomorph-colouration has been associated with selective pressure
566 from predators (Gillespie 2004). In the current study, F_{ST} comparisons uncovered melanization
567 genes in regions of genomic divergence, which agree with the natural history observations of
568 the group and the colour-characterization of the ecomorphs. Interestingly, we also find evidence
569 for genes associated with circadian rhythms, neuronal and synaptic activity genes, life span,
570 learning and memory formation, which can be seen as an unexplored axis of physiological and
571 ecological differentiation. Specifically, these genes may indicate that ecomorph-evolution may
572 involve more phenotypic changes than shifts in coloration and micro-habitat use. For example,
573 circadian rhythm genes are found across the tree of life and are important for a wide array of
574 functions, from allowing organisms to synchronise with their immediate environment to
575 regulating reproductive activity, and it has been suggested that they play an important role in
576 environmental adaptation (Yerushalmi and Green 2009). Finally, the evidence for genomic
577 differentiation in developmental genes may suggest that shifts in developmental stages may

578 have an important function in the *Tetragnatha* spiny-leg adaptive radiation. This is consistent
579 with previous work which shows that the initial appearance of the Maroon ecomorph is
580 associated with developmental shifts in coloration on the oldest island of Kauai (Brewer et al.
581 2014), by identifying several genes under selection likely responsible for changes in coloration
582 and developmental processes – reticulon nog and apolipoprotein d.

583 Conclusion

584 The whole-genome sequencing of *Tetragnatha* spiny-leg ecomorphs showed that the
585 genomic basis of repeated ecomorph evolution is multifarious even in closely related species, as
586 some ecomorphs likely arose through hybridization (Small Brown, Maroon), while others likely
587 arose by shared standing genetic variation or *de novo* mutation (Green, large Brown). We also
588 found that ecomorph evolution in the spiny-legs may go beyond their colouration, as regions
589 with high genomic divergence consistently pointed to genes associated with other potential
590 ecological axes of differentiation, such as learning and memory formation, circadian rhythms,
591 and developmental shifts. However, it is likely that a complex evolutionary history involving
592 population extinctions, population structuring and ghost-lineages blurry the reconstruction and
593 interpretation of evolutionary patterns.

594

595 Author Contributions

596 JC, THS, RGG obtained funding. JC and RGG designed the study. RGG, DDC, SK, HK, AJW
597 provided samples. JC, LH generated the data. JC analysed the data with contributions on
598 scripting and interpreting the data from CGS, VCB, JM, DDC, JML, CFP and DD. JC drafted
599 a first manuscript, and all the authors contributed to the writing and clarity of the
600 manuscript.

601 Data Accessibility

602 Data is being made available through ENA XXX. Code is at github José.

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