

1 **Niche differentiation, reproductive interference, and range expansion**

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43 **Abstract**

44 Understanding species distributions and predicting future range shifts requires considering
45 all relevant abiotic factors and biotic interactions. Resource competition has received the
46 most attention, but reproductive interference is another widespread biotic interaction that
47 could influence species ranges. Rubyspot damselflies (*Hetaerina* spp.) exhibit a
48 biogeographic pattern consistent with the hypothesis that reproductive interference has
49 limited range expansion. Here we use ecological niche models to evaluate whether this
50 pattern could have instead been caused by niche differentiation. We found evidence for
51 climatic niche differentiation, but the species that encounters the least reproductive
52 interference has one of the narrowest and most peripheral niches. These findings
53 strengthen the case that reproductive interference has limited range expansion, and also
54 provide a counterexample to the idea that release from negative species interactions
55 triggers niche expansion. We propose that release from reproductive interference enables
56 species to expand in range while specializing on the habitats most suitable for breeding.

57

58 **Introduction**

59 Understanding how species ranges have been shaped by physiological tolerances and
60 biotic interactions is crucial for predicting how species ranges will shift in the future
61 (Lancaster 2022; Sexton *et al.* 2009). Physiological tolerances can be measured
62 experimentally or inferred statistically from the relationships between environmental
63 variables and occurrence data. While “biotic interactions” is a large and diverse category,
64 the vast majority of theoretical and empirical research has focused on trophic interactions
65 and exploitative resource competition, which are closely related phenomena. Exploitative
66 competition can exclude species from portions of their fundamental niche by reducing
67 resources below the level required for the population to increase when rare (Godsoe *et al.*
68 2017). Other negative species interactions have often been assumed to have effects similar
69 to exploitative competition, but theoretical and empirical research has shown that not to be
70 the case. Direct interactions, such as aggressive and reproductive interference, can shape
71 species distributions in ways that would not be predicted by niche theory (Grether *et al.*
72 2017; Grether & Okamoto 2022; Kishi & Nakazawa 2013; Patterson & Drury 2023).
73 Interspecific interference does not consistently favor species with higher intrinsic growth
74 rates and can cause Allee effects that prevent species from persisting at low densities in
75 environments that could support much higher densities (Case *et al.* 2005). If interference is
76 asymmetrical, the species harmed least could prevail even if it has a lower intrinsic growth
77 rate in the absence of interference (Grether & Okamoto 2022; Kishi & Nakazawa 2013;
78 Ribeiro & Spielman 1986). If interference is symmetrical, the first species established in
79 an area could become a barrier to the range expansion of other species (Case *et al.* 2005).

80 Although progress has been made, empirical research on the real-world consequences of
81 interspecific interference lags far behind theory and laboratory experiments.

82 Reproductive interference is a widespread form of interspecific interference that
83 occurs in all kinds of sexually reproducing organisms (Gröning & Hochkirch 2008;
84 Takahashi *et al.* 2016; Ting & Cutter 2018). In animals, reproductive interference is
85 typically caused by persistent mate recognition errors, which can be costly for both species
86 and yet impossible for selection to eliminate (Drury *et al.* 2019). Theoretical and
87 experimental work has shown that reproductive interference can cause Allee effects and
88 impede range expansion, particularly in taxa with limited dispersal capabilities
89 (Bargielowski & Lounibos 2016; Case *et al.* 2005; Kishi *et al.* 2009; Kishi & Nakazawa
90 2013; Kuno 1992; Kyogoku 2015; Kyogoku & Nishida 2012; Kyogoku & Wheatcroft
91 2020; Noriyuki & Osawa 2016). However, the best evidence that reproductive interference
92 has shaped species ranges comes from studies of invasive species actively displacing other
93 species (Butler and Stein 1985; Söderbäck 1995; Liu *et al.* 2007; Bargielowski *et al.* 2013;
94 Sun *et al.* 2014; Bargielowski and Lounibos 2016). Determining whether reproductive
95 interference has limited species ranges is more challenging.

96 Rubyspot damselflies (*Hetaerina* spp.; Garrison 1990) are a promising study system in
97 this regard because previous research has revealed a biogeographic pattern consistent with the
98 hypothesis that reproductive interference has limited range expansion in North America (Drury
99 *et al.* 2019, 2015; Grether *et al.* 2020). Briefly, the species that encounters the least reproductive
100 interference has the largest latitudinal range and overlaps with the most congeners, while the
101 species with the largest longitudinal range is allopatric across most of its range, and thus has also
102 largely escaped from reproductive interference. Field experiments and landscape genomic

103 analyses will be required to test this hypothesis directly. Here we use existing data to evaluate
104 whether niche differentiation could account for the same biogeographic pattern. If niche
105 differentiation cannot account for the large differences between species in range size and
106 overlap, the case for reproductive interference limiting species ranges would be strengthened.

107 A few conceptual and methodological distinctions are needed to clarify our approach to
108 testing predictions of the niche differentiation hypothesis. The various ecological niche concepts
109 all have in common the idea that a species' ecological niche pertains to environmental space
110 (Godsoe *et al.* 2017; Sillero *et al.* 2021; Vandermeer 1972). As such, niche axes are
111 environmental variables, not geographic or geo-environmental dimensions. Niche differentiation
112 refers to changes in the relationship between environmental variables and population growth
113 rates that occurred after two or more species diverged from a common ancestor. Species with
114 broader niches are capable of persisting in a broader range of environments and thus tend to have
115 larger geographic ranges (Brown 1984; Carscadden *et al.* 2020; Kambach *et al.* 2019; Slatyer *et*
116 *al.* 2013). In taxonomic groups in which niche differentiation has occurred and most species are
117 habitat specialists, a habitat generalist with a relatively broad niche (Figure 1a), or one near the
118 ancestral group centroid (Figure 1b), would be expected to have the largest geographic range and
119 the most geographic overlap with the other species (Suárez-Mota & Villaseñor 2020;
120 Tomašových *et al.* 2017).

121 Ecological niche modeling is widely used to characterize and compare species' realized
122 niches (i.e., the portions of fundamental niches currently occupied) and to make predictions
123 about species distributions in regions that are incompletely surveyed or in the future based on
124 climate change projections (Ahmadi *et al.* 2018; Boys *et al.* 2021; Buermann *et al.* 2008; Guisan
125 *et al.* 2013; Inman *et al.* 2019; Melo-Merino *et al.* 2020; Novella-Fernandez *et al.* 2021; Pearson

126 & Dawson 2003; Peterson & Holt 2003). While the focus of such studies is usually on
127 geographic space, ecological niche models (ENMs) can also be used to make inferences about
128 niche differentiation (Ahmadi *et al.* 2018; Evans & Jacquemyn 2022; Warren *et al.* 2019, 2021).
129 To avoid ambiguity, we focus here on ENMs constructed using Maxent (Phillips *et al.* 2017), but
130 the same concepts apply to other correlative ENMs. Maxent is a machine-learning, presence-
131 background method that often outperforms presence-only and presence-absence methods (Elith
132 *et al.* 2011; Merow *et al.* 2013; Valavi *et al.* 2022; Wisz *et al.* 2008) and is widely used for
133 species comparisons (Jaime *et al.* 2015; Namyatova 2020; Wellenreuther *et al.* 2012). The basic
134 output is an equation for predicting the suitability of any combination of values of the
135 environmental variables in the training set. Suitability can be projected to the geographic extent
136 or to an n -dimensional environmental hypercube, where n is the number of environmental
137 variables (Figure 1c-e; Warren *et al.* 2019). The environmental hypercube is closely analogous to
138 Hutchinson's (1957) n -dimensional hypervolume, except that instead of being the discrete set of
139 environments where the species has a positive intrinsic growth rate, all environments in the
140 hypercube are weighted by their estimated suitability. Statistics commonly referred to as 'niche
141 breadth' and 'niche similarity' (or 'niche overlap') can be calculated using the hypercube or the
142 geographic projection (Figure 1d,e; Warren *et al.* 2021).

143 Importantly, niche metrics based on geographic projections are not suitable for studying
144 niche differentiation because they conflate geography and environment (Figure 1d,e). Consider
145 the question of whether species differences in niche breadth can explain differences in range
146 size. It would be circular to test whether range size correlates with niche breadth calculated using
147 geographic projections that are based on occurrence samples from across the species' ranges. For
148 this question, it only makes sense to use estimates of niche breadth in environmental space.

149 Likewise, estimates of niche similarity based on geographic projections confound niche overlap
150 with geographic overlap, and thus should not be used to evaluate whether niche similarity
151 predicts geographic overlap. While these points might seem obvious, they have not surfaced in
152 our reading of the niche modeling literature, and most papers with niche statistics based on
153 ENMs use the geographic versions (Evans & Jacquemyn 2022).

154 To evaluate whether niche differentiation can account for large differences in range size
155 and geographic overlap, we constructed ENMs for 17 of the 18 species of calopterygid
156 damselflies in North America and made several types of niche comparisons in environmental
157 space. Surprisingly, we found that the species with the largest latitudinal range and the most
158 geographic overlap with other species has one of the narrowest ecological niches and is no closer
159 in niche space to the other species than expected by chance. These results strongly suggest that
160 something other than niche differentiation accounts for this species' expansive range, and
161 therefore strengthen the case for reproductive interference limiting species ranges. This study
162 also provides a counterexample to the idea that release from negative species interactions
163 precipitates niche expansion (Herrmann *et al.* 2021; Lancaster 2022).

164

165 **Materials and methods**

166 *Study system*

167 The life cycle of rubyspot damselflies is centered around areas with suitable larval habitat in
168 perennial springs, streams, or rivers (Corbet 1999; Delgado 2002; Johnson 1973; Novelo-
169 Gutiérrez 2000). At the adult stage, females return repeatedly to the larval habitat to oviposit in
170 submerged vegetation, and males compete for perching sites above the water surface and attempt
171 to clasp arriving females (Córdoba-Aguilar & González-Tokman 2014). Most sympatric species

172 experience high levels of reproductive interference because the females are too similar
173 phenotypically for males to reliably distinguish between them, the only striking exception being
174 the smoky rubyspot damselfly, *Hetaerina titia* (Drury) (Drury *et al.* 2019, 2015; Grether *et al.*
175 2020). Female *H. titia* have darker wings than females of other species and consequently are
176 pursued and clasped at lower rates by heterospecific males (Drury *et al.* 2019, 2015). Release
177 from reproductive interference could explain why *H. titia* has the largest latitudinal range and
178 occurs in sympatry with the most congeners (Figure 2). The species with the largest longitudinal
179 range, *H. americana* (Fabricius), was recently split into two sister species based on genetic data
180 (Standring *et al.* 2022; Vega-Sánchez *et al.* 2019, 2020), but the occurrence records are
181 indistinguishable; we refer to this monophyletic subclade as “*H. americana* spp.”. The melanic
182 wing pigmentation that makes *H. titia* distinct from congeners increases its resemblance to
183 *Calopteryx* (Figure 3; Córdoba-Aguilar *et al.* 2007), a younger genus in the same family
184 (Calopterygidae; Standring *et al.* 2022). Our niche comparisons include 12 of the 13 species of
185 *Hetaerina* and all five species of *Calopteryx* that occur in North America.

186

187 *Occurrence data and range overlap*

188 Species occurrence data were obtained from the Global Biodiversity Information Facility in
189 December 2020 (<https://doi.org/10.15468/dl.wmxwvj>; <https://doi.org/10.15468/dl.cd58ag>) and
190 trimmed to a rectangular area (in degrees of longitude and latitude) slightly beyond the
191 geographic range of *H. titia* (-130, -60, 0, 55) to encompass the ranges of all species of
192 Calopterygidae with which *H. titia* overlaps. We refer to this area as the ‘full extent’, to
193 distinguish it from the rectangular areas of overlap between individual species pairs, which we
194 refer to as ‘shared extents’ (Figure 1c). To estimate species’ range sizes, we used R package

195 'alphahull' (Pateiro-López & Rodríguez-Casal 2010; The R Foundation for Statistical
196 Computing, R version 4.1.2).

197

198 *Environmental variables*

199 Variable selection is a crucial part of ecological niche modeling (Sillero *et al.* 2021; Warren *et al.*
200 *al.* 2014). Maxent requires environmental data across the full geographic extent at locations
201 where the species is known to occur and randomly selected background points (Merow *et al.*
202 2013; Phillips & Dudík 2008). The standard approach for most applications is to use a
203 combination of bioclimatic and land cover layers, but it is important to consider the natural
204 history of the species and the goals of the study (Phillips *et al.* 2006). It might be supposed that
205 aquatic variables would be required to construct ENMs for insects with aquatic larvae, but the
206 abiotic variables that govern aquatic insect assemblages (e.g., light levels, water chemistry, water
207 clarity, water temperature, disturbance regimes, organic matter, sedimentation) are strongly
208 influenced by and therefore covary with climate, elevation, and land cover (Burgherr & Ward
209 2001; Faria *et al.* 2021; Null *et al.* 2013; Yoshimura 2012). Bioclimatic layers based on air
210 temperature and precipitation have often been used along with elevation and land cover layers to
211 construct ENMs for aquatic plants (Alahuhta *et al.* 2011; Gillard *et al.* 2017; Lumbreras *et al.*
212 2013), freshwater fish (Comte & Grenouillet 2015), and aquatic insects (Kusch 2015; Megna *et al.*
213 *al.* 2021; Shah *et al.* 2015; Sundar *et al.* 2021) including Odonata (Bhowmik & Schäfer 2015;
214 Cancellario *et al.* 2022). Using more proximal environmental variables does not necessarily
215 result in better niche models; e.g., for modeling the distributions of aquatic invasive species in
216 lakes across the US, the WorldClim air temperature layers consistently outperformed a new
217 global dataset of lake surface water temperatures (Burner *et al.* 2023).

218 Scale is another important consideration (Elith & Leathwick 2009; Phillips *et al.* 2006).
219 Testing the hypothesis that niche differentiation accounts for differences between species' ranges
220 on a continental scale requires environmental variables that could differ between species' ranges
221 on a continental scale. An environmental layer for the presence/absence of flowing water would
222 be useful for generating high-resolution habitat suitability maps but not for studying niche
223 differentiation, because flowing water is a habitat requirement for all calopterygid damselflies in
224 North America (Westfall & May 1996). Syntopic species of rubyspot damselflies often differ in
225 microhabitat use (e.g., current speed, canopy cover, stream width (Anderson & Grether 2011;
226 McEachin *et al.* 2021), which suggests that microhabitat data would be useful for explaining the
227 distributions and relative densities of species within streams, as has been shown in other aquatic
228 insects (Goss *et al.* 2020), but microhabitat differences could not account for species range
229 differences at the continental scale.

230 Based on the above considerations, we followed the standard approach and constructed
231 ENMs using bioclimatic and land cover layers. Air temperature, precipitation, and elevation data
232 were obtained from WorldClim version 2.1 (Fick & Hijmans 2017) and averaged over the years
233 1970-2000. Percent tree cover and land cover type (Table S1) were obtained from MODIS Terra
234 data 2008 (Geospatial Information Authority of Japan). Normalized difference vegetation index
235 (NDVI) was obtained from NASA Earthdata using 'MODISstp' (Busetto & Ranghetti 2016).
236 Land cover type was aggregated from 15 to 30 arc-seconds (~ 1 km) to match the spatial
237 resolution of the other variables.

238 Prior to niche modeling, we cropped the environmental raster files to the full extent using
239 the 'crop' function in R package 'raster'. To avoid problems with highly correlated predictor
240 variables (Sillero *et al.* 2021), we removed variables with variance inflation factors (VIFs)

241 greater than 2.5 (Hair *et al.* 2019) using the step-wise procedure ‘vifstep’ in R package ‘usdm’
242 (Naimi *et al.* 2014). The final variable set included ten variables: NDVI, tree cover, elevation,
243 the categorical variable landcover type, and six bioclimatic variables (BIO2, BIO7, BIO8,
244 BIO15, BIO18 and BIO19) (Table S2).

245

246 *Principal component analysis*

247 To visualize differences between species in environmental space, we carried out a principal
248 component analysis (PCA) of variation among occurrence sites in the continuous environmental
249 variables with the ‘prcomp’ function in R.

250

251 *Niche models*

252 We constructed ENMs for species with 15 or more occurrence records, after thinning, with
253 Maxent 3.4.1 in R package ‘dismo’ (Phillips *et al.* 2017; Hijmans *et al.* 2021). This sample size
254 threshold enabled all ENMs to be based on the same environmental variable transformations, and
255 thus to be directly comparable (Morales *et al.* 2017), but excluded *H. pilula* (Calvert). We
256 thinned the data to one occurrence per km² per species using ‘spThin’ (Aiello-Lammens *et al.*
257 2015) and used the following Maxent settings: 10⁴ background points; regularization parameter
258 of 1; 500 iterations; linear, quadratic, product, and hinge features; complementary log-log
259 (cloglog) output; and 10 replicates per model with cross-validation and 30% random test
260 percentage (Fithian *et al.* 2015; Phillips *et al.* 2017; Phillips & Dudík 2008). Model performance
261 can be evaluated by comparing the area under the curve (AUC) to the null AUC (Phillips &
262 Dudík 2008). We computed the null AUC with R package ‘ENMTools’ v1.0.3 and subtracted it
263 from the model AUC to obtain a measure of model performance (Δ AUC).

264

265 *Niche breadth and similarity*

266 A measure of niche breadth in environmental space (B_{env}) based on Levin's normalized niche
267 breadth (B2) equation was calculated with the 'env.breadth' function in 'ENMTools' (Warren *et*
268 *al.* 2019). Two measures of niche similarity in environmental space (D_{env}, I_{env}) based on
269 Schoener's D and Hellinger's I were calculated using the 'identity.test' function (Warren *et al.*
270 2019). To evaluate whether these metrics are affected by sample size, we carried out a Monte
271 Carlo simulation. Paired samples ranging in size from 10 to 1000 were drawn without
272 replacement from the *H. titia* ($n = 1080$) and *H. americana* ($n = 3364$) occurrence records. With
273 each pair of samples, 'ENMTools' was used to generate empirical estimates and 100 null values
274 of the niche similarity metrics. The empirical and null estimates both increased asymptotically
275 with the number of occurrence records (Figure S1). We explored different ways of adjusting the
276 empirical estimates to reduce their dependence on sample size. Subtracting the null mean was the
277 most effective adjustment (Figure S1), and thus we used this method. The niche breadth metric
278 (B_{env}) was not affected by sample size (Figure S1).

279

280 *Phylogenetic linear mixed models*

281 To account for phylogenetic nonindependence in comparisons of niche metrics between species
282 pairs (Drury *et al.* 2018; Tobias *et al.* 2014), we fit phylogenetic linear mixed models (PLMMs)
283 with the Markov Chain Monte Carlo (MCMC) algorithm in 'MCMCglmm' 2.3 (Hadfield 2010).
284 We used the fossil-calibrated phylogeny of Standring *et al.* (2022), appending *C. diminiata*
285 (Burmeister) and *C. amata* (Hagen) based on the topology of Waller & Svensson (2017) (Figure
286 S2). *C. angustipennis* (Selys) could not be included because no phylogenetic data were available

287 for this species. Each PLMM included a species comparison index, patristic distance, and
288 random-effects terms for species identifiers and most recent common ancestors. We ran each
289 model four times and merged the MCMC chains after verifying convergence (Gelman & Rubin
290 1992).

291

292 *Shared-extent comparisons*

293 ENMs based on different geographic extents are not comparable (Merow *et al.* 2013), but using
294 large extents for species with small ranges can result in overfitting and AUC inflation (Fourcade
295 *et al.* 2014; Sillero *et al.* 2021). To make pairwise species comparisons, we constructed ENMs
296 using the shared extents of each species pair. This allowed species to be compared in niche
297 breadth (B_{env}) and two additional metrics: mean suitability and proportion of the shared extent
298 occupied. If a geographic area has higher mean suitability for species A than species B, it should
299 be easier for species A to become established in the area. If species A occurs in a larger
300 proportion of a geographic area than species B, it follows that species A has some advantage
301 relative to species B. Making such comparisons in the full extent would be uninformative
302 because mean suitability is affected by range size, and the proportion of the full extent occupied
303 by a species is directly proportional to its range size. In the shared extents, however, these
304 comparisons are valid and could be informative.

305 We made two types of shared-extent comparisons. First, we compared the niche metrics
306 of *H. titia* to those of the other 16 species using Wilcoxon paired tests. Second, we compared *H.*
307 *americana* and *H. titia* in their respective shared extents with 10 other species by subtracting the
308 other species' niche metrics from those of *H. americana* and *H. titia* and comparing the two sets
309 of differences with PLMMs.

310

311 **Results**

312 *Principal component space*

313 The first three principal components (PCs) accounted for 68.5% of the variance in the continuous
314 environmental variables (Table S3). The first two PCs separated the species into two clusters,
315 with species with northerly ranges in one cluster and species with southerly ranges in the other;
316 the third PC largely separated species by mean temperature, elevation and tree cover (Figure 4).
317 On the first two PCs, *H. titia* occupied a position between the two clusters but closer to the
318 northern cluster, yet on PC3, *H. titia* occupied a peripheral position, with a mean exceeding that
319 of all species except *H. pilula* (Figure 4). The centroid of *H. titia* was closest to species that
320 occur in the eastern US and farthest from species restricted to Central America (Table S4).

321

322 *Maxent models*

323 The full-extent ENMs significantly outperformed chance expectations, with Δ AUC in the 0.2 -
324 0.4 range (Table S4); most shared-extent models also had Δ AUC in the 0.2 - 0.4 range (Tables
325 S5 & S6).

326 Climatic variables made the largest contributions to the full-extent ENMs, with landcover
327 type perhaps controlling for the relatively recent effects of human activities on damselfly
328 distributions (Table S7). Annual temperature range (BIO7) made the largest contribution to most
329 *Hetaerina* spp. ENMs, while precipitation seasonality (BIO15) or elevation made the largest
330 contribution to *Calopteryx* spp. ENMs. Elevation and precipitation in the coldest quarter (BIO19)
331 made the largest contributions to the ENMs of *H. vulnerata* (Hagen) and *H. fuscoguttata* (Selys),
332 respectively. Landcover type was the second or third most important variable for four *Hetaerina*

333 species and caused the largest decrease in gain when omitted from the *H. miniata* (Selys) model.
334 The vegetation index (NDVI) made the third largest contribution to the *H. americana* spp. ENM.
335 The top three variables for *H. titia* were annual temperature range, elevation, and mean
336 temperature in the warmest quarter (BIO8).

337

338 *Niche metrics*

339 The hypothesis that *H. titia* overlaps geographically with multiple congeners because it occupies
340 a niche near the genus centroid can be rejected. The ENM of *H. titia* in the full geographic extent
341 was not more similar to that of other calopterygid species than the other species' ENMs were to
342 each other (Figure 5a; Table S4; PLMM, $p \geq 0.24$, 120 species pairs), and the same was true
343 when the analysis was restricted to *Hetaerina* spp. ($p \geq 0.47$, 66 species pairs). The ENM of *H.*
344 *titia* differed significantly from all other species except *H. sempronia* (Hagen) and *H.*
345 *fuscoguttata*, and was least similar to species with higher latitudinal or elevational ranges (Table
346 S4).

347 The hypothesis that *H. titia* has an unusually broad niche can also be rejected. *H. titia* had
348 the fourth smallest niche breadth in the full extent (Table S4) and a smaller niche breadth than
349 the other species in the shared extents (Wilcoxon paired test $V = 125$, $n = 16$, $p = 0.002$; Figure
350 6a). *H. titia* did not differ in mean suitability from the other species ($V = 91$, $n = 16$, $p = 0.25$)
351 and yet occupied a larger proportion of the shared extents ($V = 12$, $n = 16$, $p = 0.002$; Figure 6b).

352 The congener with the largest longitudinal range, *H. americana* spp., had the largest
353 niche breadth in the full extent (Table S4). In their respective shared extents with other species,
354 *H. americana* spp. and *H. titia* did not differ in the proportion of the shared extents occupied

355 (PLMM, $p = 0.48$, 20 species pairs) or mean suitability ($p = 0.49$), but *H. americana* spp. tended
356 to have greater relative niche breadth ($p = 0.06$).

357 Sympatric species (67 species pairs) had more similar niches than allopatric species (53
358 species pairs; PLMM, $p < 0.001$; Figure 5b). Niche similarity was greater between species in the
359 same genus (72 species pairs) than between species in different genera (48 species pairs; Figure
360 5c), but with only two genera the phylogeny fully accounted for these differences. Across all
361 species, sympatric species were closer in patristic distance than allopatric species (Wilcoxon test,
362 $W = 2926$, $p < 0.0001$, $n_{allo} = 53$, $n_{sym} = 67$), but with the analysis restricted to *Hetaerina* spp.,
363 sympatric and allopatric species did not differ in patristic distance ($W = 420$, $p = 0.21$, $n_{allo} = 13$,
364 $n_{sym} = 53$). Patristic distance was not a significant predictor of species differences in any niche
365 metric in the full extent (PLMMs, $p \geq 0.21$) or shared extents ($p \geq 0.63$).

366

367 **Discussion**

368 Modest differences between closely related species in range size might not be predictive of
369 differences in niche breadth, in part simply because some environments are more common than
370 others (Brown 1984). Yet species with exceptionally large geographic ranges are expected to
371 have relatively broad niches (Brown 1984; Carscadden *et al.* 2020; Kambach *et al.* 2019; Slatyer
372 *et al.* 2013). Likewise, species that overlap geographically with large numbers of congeners can
373 be predicted to have niches that, if not unusually broad, are close to the genus centroid (Figure
374 1). We tested these straightforward predictions in calopterygid damselflies, and while we found
375 evidence for niche differentiation within and between genera, the species with the largest
376 latitudinal range and which overlaps with the most congeners (*H. titia*) has a relatively narrow
377 and peripheral climatic niche. By contrast, the species with the largest longitudinal range (*H.*

378 *americana* spp.) has the largest niche breadth in the genus. Both of these species have largely
379 escaped from reproductive interference, *H. titia* by diverging in female coloration (Drury *et al.*
380 2019, 2015; Grether *et al.* 2020) and *H. americana* spp. by being allopatric to all other congeners
381 across most of its range. Niche differentiation and reproductive interference are not mutually
382 exclusive alternatives – both could have contributed to species differences in range.
383 Nevertheless, finding that the species with the largest latitudinal range, and which encounters the
384 least reproductive interference, has a relatively narrow climatic niche strengthens the hypothesis
385 that reproductive interference has limited the range expansion of other species in this genus.

386 When a species expands beyond its previous range edge into marginally suitable habitats,
387 ecological and evolutionary responses can broaden the species' niche and facilitate further range
388 expansion (Alexander *et al.* 2022; Case *et al.* 2005; Lancaster 2022). Release from negative
389 species interactions can also lead to niche expansion within a species' existing range (Bolnick *et*
390 *al.* 2010; Herrmann *et al.* 2021; Sjödin *et al.* 2018). If release from reproductive interference
391 facilitated the range expansion of *H. titia*, why does this species have a relatively narrow climatic
392 niche? In some cases, gene flow from the core of a species' range might prevent adaptation at the
393 edges (Case & Taper 2000; Kirkpatrick & Barton 1997), but *Hetaerina* populations in different
394 river drainages are sufficiently isolated genetically to evolve independently (Drury *et al.* 2019).
395 A better explanation may be that current theory on the effects of species interactions on niche
396 evolution, which is largely based on exploitative resource competition (Alexander *et al.* 2022),
397 does not apply to reproductive interference.

398 We propose that, instead of triggering niche expansion, release from reproductive
399 interference enables species to expand in range while specializing on the environments most
400 suitable for breeding. The basic argument for why release from interspecific competition leads to

401 niche expansion is that the absence of the competitor frees up resources and converts unsuitable
402 habitat into suitable habitat (Alexander *et al.* 2022; Bolnick *et al.* 2010). Release from
403 reproductive interference would not free up resources, but it would facilitate breeding in areas
404 with high heterospecific densities by eliminating Allee effects. With no barrier to expansion into
405 high-suitability habitats, a species released from reproductive interference would be expected to
406 be concentrated where it reaches the highest population densities, that is, in the core of its
407 fundamental niche. Put simply, release from reproductive interference would enable a species to
408 expand into habitats that support high population densities, resulting in a distribution of
409 suitability values in environmental space that resembles niche contraction (Warren *et al.* 2019).
410 Selection might then strengthen the preference for high suitability habitats, resulting in a
411 narrowing of the species' fundamental niche. This hypothesis is similar to the hypothesis that
412 reductions in pollinator-mediated reproductive interference promote coexistence between plants
413 that specialize on the same pollinators (Katsuhara *et al.* 2021).

414 We found evidence for niche divergence in the vast majority of congeneric species
415 pairs (Figure 5c), which would seem to differ from what has been found in most other taxa
416 (Peterson 2011), but *Hetaerina* is an ancient genus with estimated species divergence
417 times of 10 to 36 million years (Standring *et al.* 2022). Our finding that sympatric species
418 have more similar climatic niches than allopatric species is not surprising from a statistical
419 standpoint, because the same result would be expected if species ranges overlapped
420 randomly (Warren *et al.* 2014). However, several ecological and evolutionary processes
421 can cause the niches of sympatric species to differ more than those of allopatric species
422 (e.g., competitive displacement, species sorting, character displacement; Pfennig &
423 Pfennig 2012), and negative relationships between niche similarity and geographic overlap

424 have been found in other studies (Novella-Fernandez *et al.* 2021). Perhaps the factors that
425 mediate coexistence in calopterygid damselflies operate on smaller spatial scales than our
426 niche models (Anderson & Grether 2011; McEachin *et al.* 2021). Research on other
427 damselfly clades has shown that predation-mediated trade-offs at the larval stage promote
428 local coexistence in some species assemblages, while in other assemblages the species
429 may be ecologically equivalent and subject to neutral dynamics (Grether *et al.* 2023;
430 Leibold & McPeck 2006; Siepielski *et al.* 2010).

431 Niche differentiation has often been reported to be decoupled from phylogenetic
432 differentiation (Losos 2011; Warren *et al.* 2014), and our study provides another example.
433 Niche similarity was greater within than between genera, but after controlling for tree
434 topology, there was no relationship between niche divergence and patristic distance. This
435 indicates that more closely related congeners are not more similar ecologically. Across all
436 species included in this study, sympatric species were closer in patristic distance than
437 allopatric species, but this appears to be an artifact of the geographic ranges of the genera
438 (*Calopteryx* spp. only occur north of Mexico while most *Hetaerina* spp. occur south of the
439 US-Mexico border; Figures 2 and 3). With the analysis restricted to *Hetaerina*, sympatric
440 and allopatric species did not differ in patristic distance. This might seem to contradict the
441 hypothesis that reproductive interference impedes range expansion, because reproductive
442 interference is generally expected to be stronger between more closely related species
443 (Grether *et al.* 2017; Gröning & Hochkirch 2008), but *Hetaerina* does not appear to
444 conform to that expectation (Drury *et al.* 2015; Grether *et al.* 2020). In any case, a much
445 larger sample of species would be required to robustly test for phylogenetic effects on
446 geographic overlap between species, especially considering that the null expectation is for

447 sympatric species to be less closely related than allopatric species because speciation
448 usually occurs in allopatry (Tobias *et al.* 2014; Warren *et al.* 2014; Weir & Price 2011).

449 In principle, joint species distribution models (JSDMs) could be used to statistically
450 distinguish species interactions from niche differentiation (Poggiato *et al.* 2021; Wilkinson *et al.*
451 2019). We opted not to use this approach because species interactions cannot be distinguished
452 from correlated responses to unmeasured environmental variables (Ovaskainen *et al.* 2016;
453 Poggiato *et al.* 2021). With our study system and others, data are not available for all
454 environmental variables that are likely to affect species ranges in similar ways (e.g., dispersal
455 barriers, microhabitat, disturbance). The omission of such variables from a JSDM could result in
456 spurious positive correlations between species that mask negative species interactions
457 (Ovaskainen *et al.* 2016; Poggiato *et al.* 2021). Single-species ENMs (and SDMs) are also
458 limited by the data used to construct them, and niche differentiation could be missed if variables
459 that affect species ranges differently are omitted (Peterson & Nakazawa 2007). We judge this to
460 be a less severe problem than omitting variables that affect species ranges in similar ways from
461 JSDMs, but it is important to consider whether unexplained differences between species ranges
462 could be due to unmeasured variables. We are not aware of any unmeasured variables that might
463 explain continental-scale differences between the species ranges of calopterygid damselflies in
464 North America.

465 Several other recent studies have drawn inferences about species interactions using
466 ENMs. By comparing the known ranges of Palearctic bat species to those predicted by
467 ENMs, Novella-Fernandez *et al.* (2021) showed that two pairs of cryptic species with very
468 similar niches exhibit a pattern of geographic avoidance consistent with competitive
469 exclusion. Cavalcante *et al.* (2022) combined the outputs of ENMs with those of geospatial

470 abundance models and found evidence that the strength of interspecific competition
471 between two endangered Amazonian primates (Atelidae) varies with habitat suitability.
472 Likewise, Braz et al. (2020) found that abundance of a neotropical opossum (*Marmosops*
473 *incanus*) covaried with the number of potential competitor species at high suitability sites
474 but not at low suitability sites.

475 In closing we note that niche theory need not be limited to abiotic factors, trophic
476 interactions and resource competition just because Hutchinson (1957) framed it that way.
477 Embracing the full suite of biotic interactions that shape the realized niche would enhance
478 niche theory and its utility for understanding and predicting species distributions. On a
479 methodological note, we encourage niche modelers to pay closer attention to the
480 distinction between geographic and environmental space!

481

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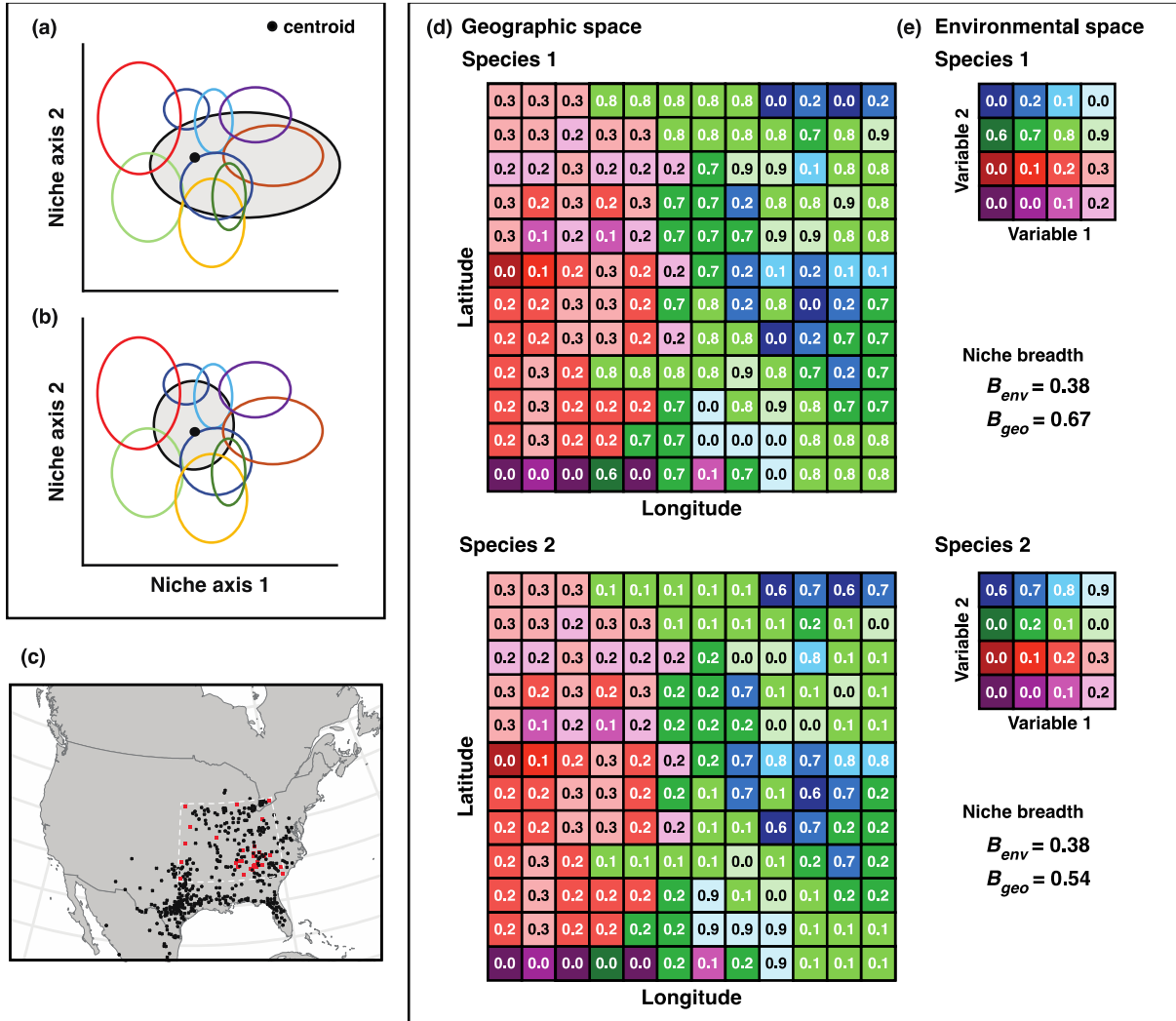
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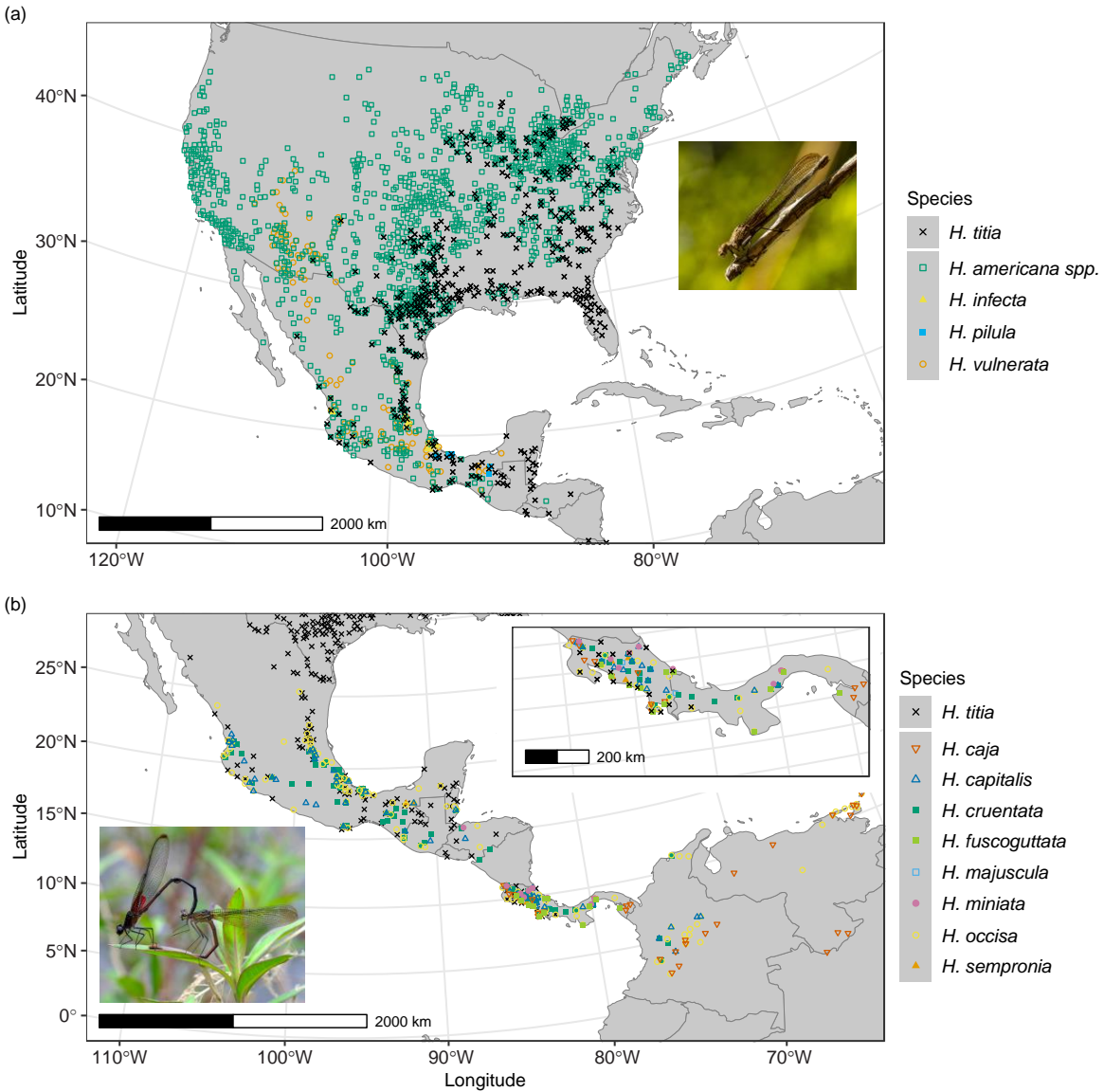


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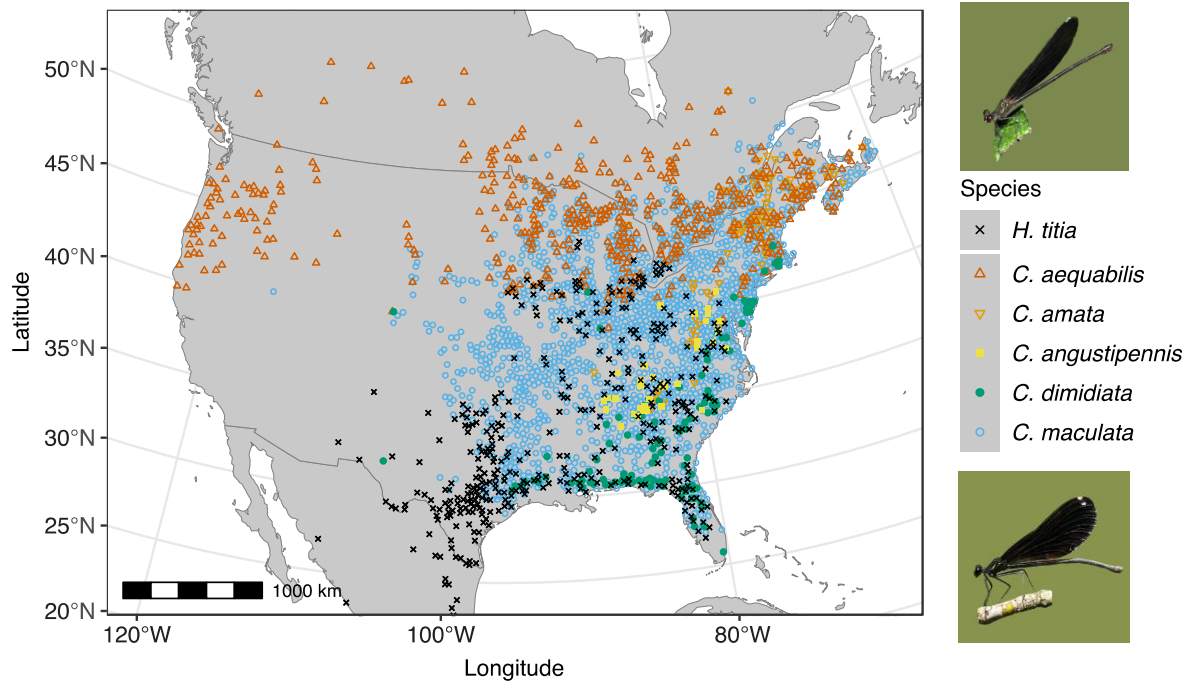
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796 **Figure 1.** Conceptual diagrams. Upper left: illustration showing that one species' niche (gray
 797 ellipse) could overlap the niches of all other species in the group by being (a) especially broad or
 798 (b) close to the group centroid (multivariate average). Lower left (c): map depicting the shared
 799 geographic extent of two species (within the dashed lines), with red and black symbols
 800 representing occurrence records. Right: hypothetical examples of ENM suitability values
 801 projected to (d) geographic space and (e) environmental space for two species. For illustration
 802 purposes, the environmental space consists of two variables and 16 environments (colors) and

803 the geographic space consists of 144 locations. Each environment is represented by one cell in
804 environmental space and by multiple cells in geographic space. The numbers in the cells
805 represent suitability values. Green environments have higher suitability for species 1 and blue
806 environments have higher suitability for species 2, but the species have the same distribution of
807 suitability values and therefore the same niche breadth in environmental space (B_{env}). Green
808 environments are more common on the geographic landscape than blue environments, and thus
809 species 1 has greater niche breadth than species 2 in geographic space (B_{geo}). Niche breadth was
810 calculated with Levin's normalized niche breadth equation (B2).



811
 812 **Figure 2.** Occurrence maps for all species of rubyspot damselflies (*Hetaerina* spp.) within the
 813 range of *H. titia*, separated into (a) northern and (b) southern groups to facilitate viewing. The
 814 inset map in (b) is a close-up view of Costa Rica and Panama. Points for *H. titia* were plotted on
 815 top of the other species' points in (a) and the inset map. Occurrence records were thinned to a
 816 minimum separation distance of 25 km for plotting. Inset photos: (a) female *H. americana*
 817 hunting; (b) mating pair of *H. cruentata*. Mapped with Albers equal-area conic projection for
 818 North America using R package 'sf' version 1.0-13. Photo credits: G.F. Grether.



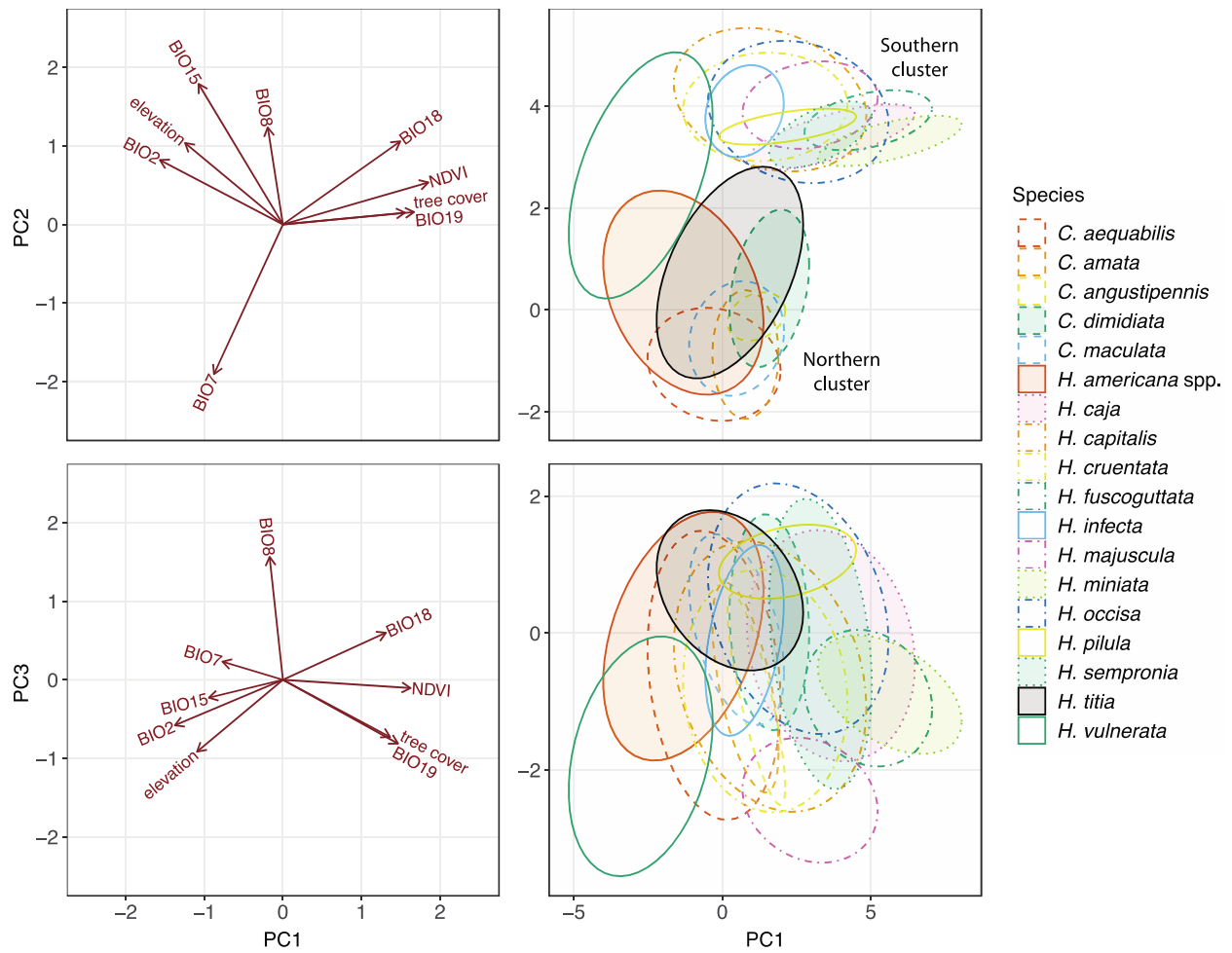
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820 **Figure 3.** Occurrence maps for all species of *Calopteryx* within the range of *H. titia*. Points for

821 *H. titia* were plotted on top of the other species' points. Upper photo: female *H. titia*. Lower

822 photo: female *C. maculata*. Photo credits: G.F. Grether. For mapping details, see Figure 1.

823



824

825 **Figure 4.** Principal component analysis of environmental variation among the occurrence
 826 records of 18 species of calopterygid damselflies within the geographic range of *H. titia*. Left

827 panels: arrows representing the principal component loadings of the environmental variables.

828 Right panels: probability ellipses encompassing 68% (1 standard deviation) of the occurrence

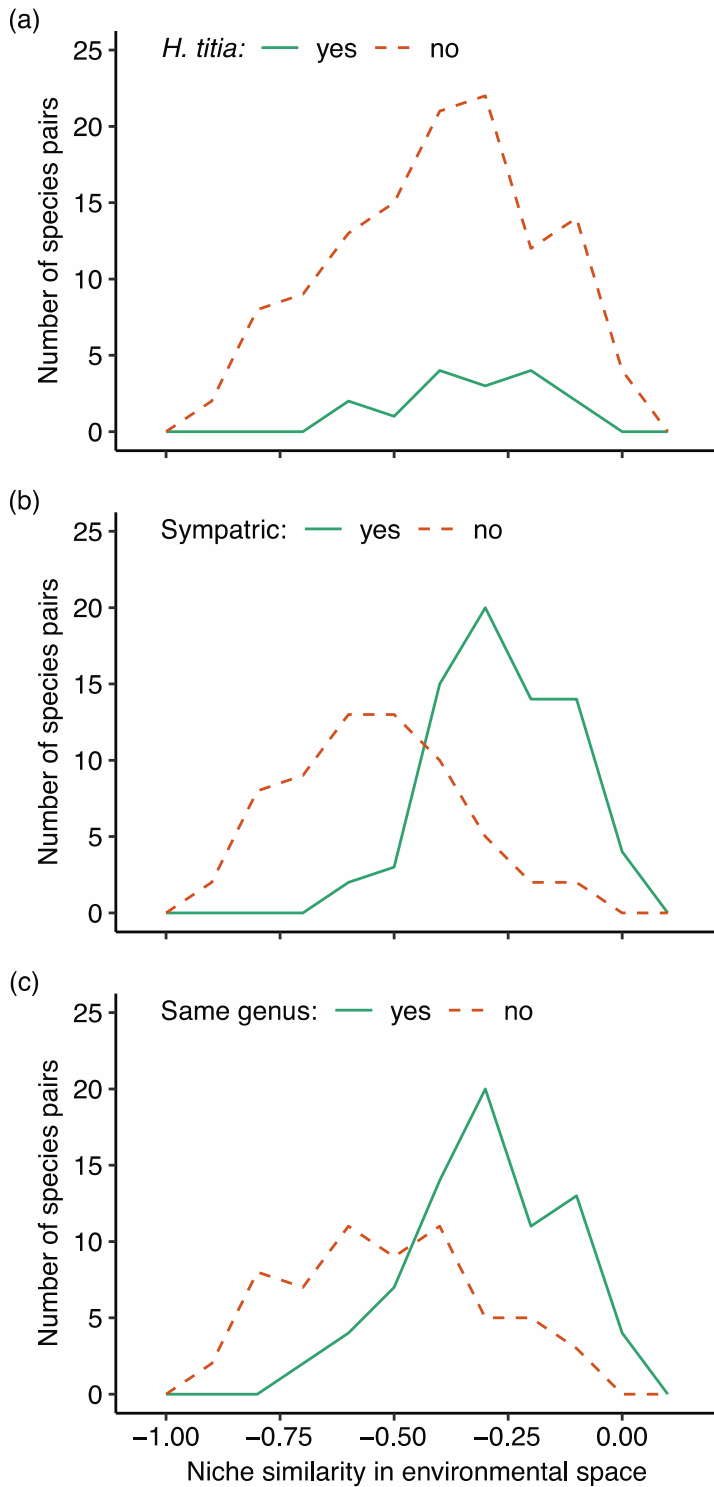
829 records. Key to variable names: diurnal temperature range (BIO2), annual temperature range

830 (BIO7), mean temperature of wettest quarter (BIO8), precipitation seasonality (BIO15),

831 precipitation of warmest quarter (BIO18), precipitation of coldest quarter (BIO19), normalized

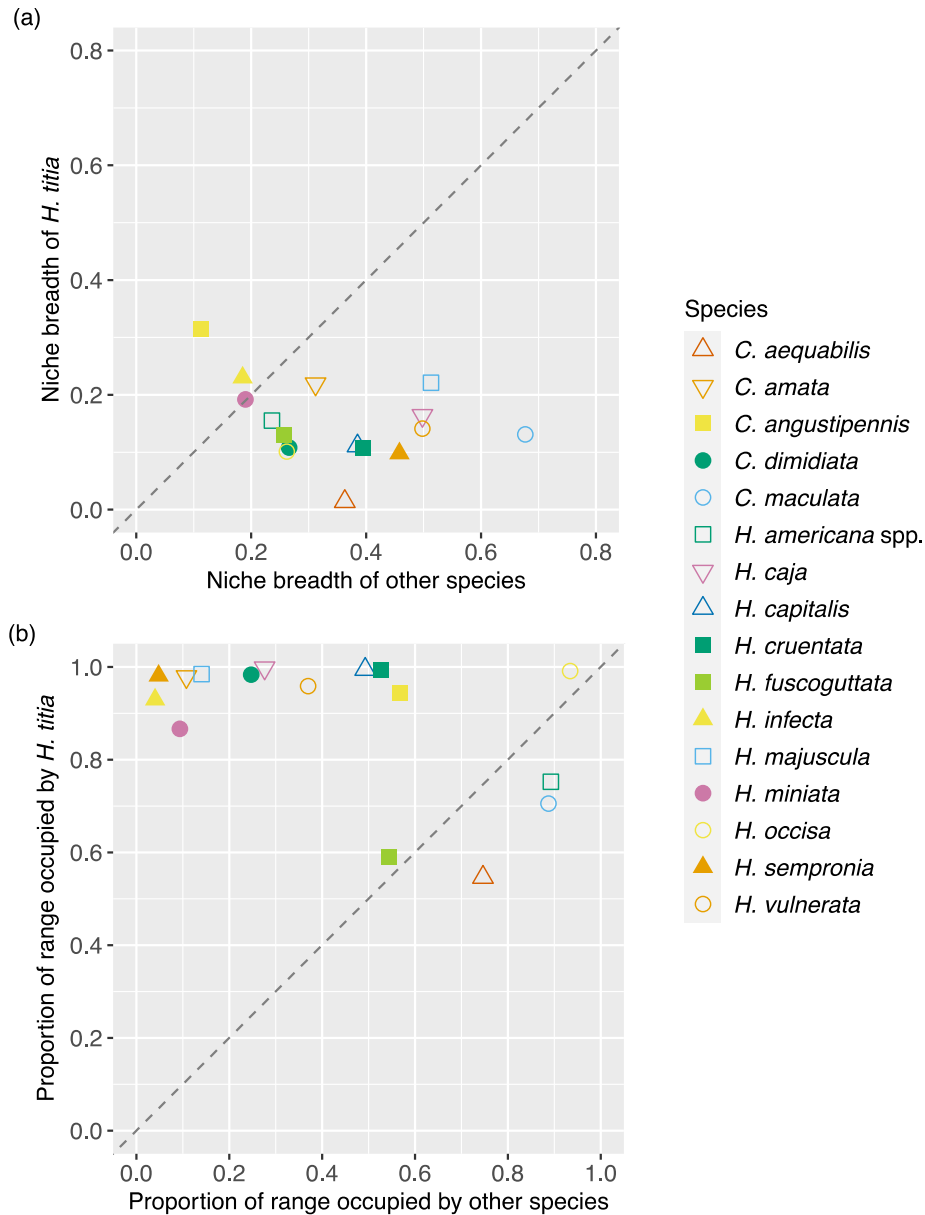
832 difference vegetation index (NDVI).

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834
 835 **Figure 5.** Frequency polygons of niche similarity in environmental space for different groupings
 836 of calopterygid damselfly species. (a) Species pairs that include *H. titia* versus all other species
 837 pairs. (b) Sympatric species pairs versus allopatric species pairs. (c) Species in the same genus

838 versus species in different genera (i.e., *Hetaerina* and *Calopteryx*). Niche similarity was
839 estimated from Maxent ENMs in the full geographic extent using index I_{env} (results for index
840 D_{env} were similar). To remove sample-size effects, the null mean I_{env} was subtracted from the
841 observed I_{env} , and thus values farther below 0 on the horizontal axis represent species pairs with
842 lower niche similarity. Differences less than ~ -0.15 were significant with the identity test (Table
843 S4).
844



845

846 **Figure 6.** Comparisons between *H. titia* and other species of calopterygid damselflies in the
 847 shared extents (i.e., rectangular areas of geographic overlap). (a) Niche breadth in environmental
 848 space (B_{env}) estimated from separate Maxent models for each shared extent. (b) Proportion of the
 849 shared extents occupied (1 km² resolution). The dashed lines have a slope of 1, and thus points
 850 above (below) the lines are cases in which *H. titia* had a larger (smaller) value than the other
 851 species.



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