

## **Thermal flexibility and a generalist life history promote urban tolerance in butterflies**

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## 1 ABSTRACT

2 Urban expansion poses a serious threat to biodiversity. Given that the expected area of urban  
3 land cover is predicted to increase by 2-3 million km<sup>2</sup> by 2050, urban environments are one of  
4 the most widespread human-dominated land-uses affecting biodiversity. Responses to  
5 urbanization differ greatly among species. Some species are unable to tolerate urban  
6 environments (i.e., urban avoiders), others are able to adapt and use areas with moderate levels  
7 of urbanization (i.e., urban adapters), and yet others are able to colonize and even thrive in urban  
8 environments (i.e., urban exploiters). Quantifying species-specific responses to urbanization  
9 remains an important goal, but our current understanding of urban tolerance is heavily biased  
10 towards traditionally well-studied taxa (e.g., mammals and birds). We integrated a continuous  
11 measure of urbanization — VIIRS night-time lights — with over 900,000 species' observations  
12 from GBIF to derive a comprehensive analysis of species-specific (N=158 species) responses of  
13 butterflies to urbanization across Europe. The majority of butterfly species included in our  
14 analysis avoided urban areas, regardless of whether species' urban tolerances were quantified as  
15 a mean score of urban tolerance across all occurrences (79%) or as a species' response curve to  
16 the whole urbanization gradient (55%). We then used the species-specific responses to  
17 urbanization to assess which life history strategies promote urban tolerance in butterflies. These  
18 trait-based analyses found strong evidence that the average number of flight months, likely  
19 associated with thermal niche breadth, and number of adult food types were positively associated  
20 with urban tolerance, while hostplant specialism was negatively associated with urban tolerance.  
21 Overall, our results demonstrate that specialist butterflies, both in terms of thermal and diet  
22 preferences, are most at risk from increasing urbanization, and should thus be considered in  
23 urban planning and prioritized for conservation.

24 *Keywords:* GBIF; butterflies; lepidoptera; trait-based ecology; climate change; generalism

## 25 INTRODUCTION

26 Anthropogenic habitat modification is, and will continue to be, one of the most significant  
27 drivers of biodiversity declines (Pereira et al. 2010; Barlow et al. 2016; Matuoka et al. 2020). Of  
28 the various anthropogenic stressors, urbanization is one of the most widespread near-term threats  
29 to biodiversity assemblages (McDonald et al. 2019). Urbanization directly leads to habitat loss,  
30 fragmentation, and degradation (Liu et al. 2016). Moreover, urbanization is associated with  
31 increased noise (Francis et al. 2011), light (Hopkins et al. 2018), and chemical (Kabir et al. 2014)  
32 pollution, which also adversely impacts biodiversity (McKinney 2006). With the expected  
33 amount of urban land cover to increase by 2-3 million km<sup>2</sup> by 2050 (Huang et al. 2019), it is  
34 critical to understand how biodiversity responds to urbanization.

35

36 While the evidence is clear that urbanization can significantly alter biological communities  
37 (Fenoglio et al. 2020), in many cases leading to biotic homogenization (McKinney 2006), there  
38 is a large range of responses among species (Threlfall et al. 2012; Lintott et al. 2016; Gippet et  
39 al. 2017). Some species have adapted to (Homola et al. 2019), and are even thriving in, urban  
40 environments (Evans and Gawlik 2020), while others have been extirpated by urbanization  
41 processes (Warren et al. 2019). In general, species can be placed along a continuum according to  
42 their response to urban environments. On one end of this continuum, species preferentially avoid  
43 urban areas leading to displacement in the face of increasing urbanization. And on the other end  
44 of this continuum, species persist in, or even colonize, urban environments to take advantage of  
45 various aspects of the form and structure. Quantifying the extent to which a given species is able  
46 to tolerate urban environments is important for restoration prioritization and for incorporating  
47 biodiversity in future urban planning (e.g., Winchell et al. 2017).

48

49 A species' ability to tolerate urban environments is a result of that species' unique life history  
50 and circumstances, including the species it interacts with (Martin and Bonier 2018), its niche  
51 breadth (Bonier et al. 2007; Palacio 2020), various life history traits (Rodewald and Gehrt 2014;  
52 Lowe et al. 2017; Jung and Threlfall 2018; Callaghan et al. 2019), phylogenetic predisposition  
53 (Sol et al. 2017), or cultural influences (Clucas and Marzluff 2012). This body of previous  
54 research has highlighted the complexity of this question, and results have been largely  
55 inconclusive. However, ecological theory predicts that species traits may be useful predictors to  
56 describe generalities across species (Vallet et al. 2010; Barnum et al. 2017; Jung and Threlfall  
57 2018). Identifying these general patterns in the types of species most at risk from increasing  
58 urbanization will also aid conservation decision-making.

59

60 Our current understanding of urban tolerance and the relationship between urban tolerance and  
61 ecological and life history traits is heavily biased towards traditionally well-studied taxa (e.g.,  
62 mammals and birds). Much is known about the ability of traits to predict urban tolerance in birds  
63 (Callaghan et al. 2019; Palacio 2020), mammals (Santini et al. 2019; Uchida et al. 2020), and  
64 amphibians (Winchell et al. 2020; Martínez-Gómez 2020). For other taxa, such as insects, the  
65 response to urbanization remains poorly quantified, but there is evidence that some taxa are more  
66 affected than others (e.g., Fenoglio et al. 2020). It is increasingly important to better understand  
67 how insects are responding to increasing urbanization, given the potential declines of insects at  
68 various spatial scales (Wepprich et al. 2019; Piano et al. 2019; Didham 2020; Svenningsen et al.  
69 2021).

70

71 Butterflies have large geographic ranges, occupy a number of different niches, are popular with  
72 the general public and hence citizen science monitoring, and can be used as indicators of  
73 environmental change due to their sensitivity to local environmental changes at small scales  
74 (Blair 1999; Essens et al. 2017). These attributes combine to make butterflies an excellent taxa to  
75 quantify responses to urbanization. Although butterflies are negatively impacted by urbanization  
76 (Mata et al. 2014; Tzortzakaki et al. 2019; Fenoglio et al. 2020; Kurlyo et al. 2020), minor  
77 changes in urban greenspace management (e.g., connectivity) can help foster and lead to an  
78 increase in butterfly diversity within urban environments, suggesting species are affected by the  
79 relative amount of urban surfaces compared to green surfaces (Mata et al. 2014; Dylewski et al.  
80 2019; Nagase et al. 2019; Iserhard et al. 2019). But as with other more well-studied taxa, certain  
81 species are more likely to tolerate urban environments than other species and species-specific  
82 responses to urbanization still need to be quantified (Mata et al. 2014).

83  
84 Our objective was to quantify species-specific measures of urban tolerance for European  
85 butterfly species at a macroecological scale (i.e., continental Europe) using a continuous measure  
86 of urbanization. First, we integrated these species-specific measures of urban tolerance with trait  
87 data to test which traits best predict urban tolerance in butterflies. We expected that certain life  
88 history and ecological traits would correlate with urban tolerance (see Table 1 for details on traits  
89 tested and predictions), including climate tolerance such as flight period and overwintering  
90 strategy (Pöyry et al. 2006), the degree of generalism (Bartanova et al. 2014), body size  
91 (Coulthard et al. 2019), microhabitat use (Essens et al. 2017), and general life history traits such  
92 as voltinism and egg laying type (Wepprich et al. 2019). Second, we applied a cluster analysis  
93 across all species to characterize the most typical species' response curves to urbanization and

94 the complex of traits associated with each type of response. Ultimately, these analyses help to  
95 identify the species that are most at risk from increasing urbanization.

96

## 97 METHODS

### 98 *Butterfly observation data from GBIF*

99 We downloaded data from the Global Biodiversity Information Facility (GBIF) for butterfly  
100 occurrence throughout continental Europe (GBIF.org 2020). We downloaded data from 2010 to  
101 2020 and only considered observations of butterflies in Europe (i.e., from Papilionidae,  
102 Hesperiiidae, Pieridae, Riodinidae, Lycaenidae, and Nymphalidae). Only observations that had  
103 coordinates and did not have geospatial issues, as flagged by GBIF, were kept for potential  
104 analysis. We removed possible GBIF duplicates from analysis by removing any observations that  
105 had the same date, latitude, longitude, and species.

106

107 We defined a near-contiguous European region for analysis to account for geographic  
108 heterogeneity in the number of records (see Table S1 with the countries included in the analysis  
109 and the corresponding sample sizes). We trimmed the extent to exclude predominantly offshore  
110 islands and regions with disparate records from the analysis (see Figure S1 for the study extent).

111

### 112 *Urban tolerance of butterflies*

113 We estimated a measure of urban tolerance for each species along a continuum of urbanization.  
114 This approach is similar to what others have used to calculate species' thermal tolerances (e.g.,  
115 Devictor et al. 2012). We overlaid GBIF observations with a continuous measure of  
116 urbanization: VIIRS night-time lights (Elvidge et al. 2017). VIIRS night-time lights measure the

117 radiance in the night-time sky, which strongly corresponds to urbanization processes such as  
118 human population density and urban form and function (Pandey et al. 2013; Zhang and Seto  
119 2013; Stathakis et al. 2015). It has an added advantage that it is globally applicable, and open-  
120 source data, allowing for the applicability of our analysis in other parts of the world. We took the  
121 median values of all images from 2014-2020 at the native resolution of 15 arc-seconds (~ 500 m)  
122 (Evlidge et al. 2017). See Callaghan et al. 2020a for more details about this process. Spatial  
123 analyses were performed in Google Earth Engine (Gorelick et al. 2017). We acknowledge that  
124 butterflies can necessarily select habitat at spatial scales less than 500 m, but our analysis here  
125 was focused on landscape-level responses.

126  
127 After each observation was assigned a measure of VIIRS night-time lights at a continuous scale,  
128 each species had a distribution of their frequency of use along an urbanization gradient (e.g.,  
129 Figure S2). Only species with a minimum of 250 observations were considered for analyses as  
130 this has been shown previously to minimize the variance in response to urbanization among  
131 species and be applicable at localized spatial scales (Callaghan et al. 2020a; Callaghan et al.  
132 2020b). Because each species differs in their geographic extent across Europe (Schweiger et al.  
133 2014) we adjusted the distribution of VIIRS night-time light levels for each species by  
134 standardizing for (1) the available urban habitat in a species' range and (2) the bias in sampling  
135 observations in a species' range relative to urban habitat (Callaghan et al. 2020c, Liu et al. 2021).  
136 To do this, we created a concave hull around the observations for each species using the  
137 concaveman package in R (Gombin 2020). We then subtracted the mean of all VIIRS values for  
138 all observations within a species' range from the mean of all VIIRS observations for a given  
139 species. This provides a value that can be negative (species under-occupy urban areas suggesting

140 they actively avoid them) or positive (species over-occupy urban areas suggesting they prefer  
141 them). This measure of urban tolerance was treated as our response variable in further analyses  
142 and referred to as an urban tolerance score. The urban tolerance score was strongly correlated  
143 with the breadth of urbanization used by a species as well, calculated by the interquartile range  
144 of species' distribution to VIIRS night-time lights (see Figure S3). In other words, species with  
145 higher mean urban tolerance scores also occupied areas with a large range of VIIRS values.

146

147 To confirm that our measure of urban tolerance captured the continuum in species-specific  
148 responses, and was not driven by detection bias of species towards urban areas, we ran an  
149 additional analysis using occupancy-detection models. We modelled relationships between  
150 species' occurrence patterns and the level of urbanization using species-specific occupancy  
151 models, also allowing urbanization to affect detection probabilities, in the unmarked package  
152 (Fiske and Chandler 2011). There was a strong correlation in the estimated urban tolerances  
153 between these two markedly different approaches, and we therefore focused our analyses on the  
154 urban tolerance score described above, which is a simpler and more generalizable approach (see  
155 details in Figure S4).

156

### 157 *Life history and ecological traits*

158 Based on known relationships in the published literature, we developed a trait framework that  
159 involved five broad categories of traits (i.e., extent of specialization, body size, microhabitat use,  
160 life history, and climate tolerance), each with one or more specific variables to represent these  
161 categories, with a total of 11 different traits (Table 1). The traits investigated were: (1) average  
162 number of flight months; (2) overwintering strategy; (3) mean temperature in a species' range;



163 (4) number of adult food types; (5) hostplant specificity; (6) hostplant specialism index; (7) wind  
164 index; (8) mean voltinism; (9) egg laying type; (10) hostplant growth forms; and (11) number of  
165 egg laying locations. Trait data were extracted from Middleton-Welling et al. 2020 for all traits  
166 besides the mean temperature of a species' range (a measure of thermal preference), which was  
167 extracted from Schweiger et al. 2014. After taxonomic matching (all names were matched to the  
168 taxonomy provided by Middleton-Welling et al. 2020), we were left with 159 species that had  
169 both an urban tolerance score and associated trait data (Table S2). One of these species, however,  
170 Geranium Bronze (*Cacyreus marshalli*) had an urban tolerance score 5x greater than any other  
171 species in our dataset because it is a known invasive pest that often relies on houseplants and has  
172 known synanthropy with novel anthropogenic environments (Quacchia et al. 2008). This was the  
173 only species in the dataset that was not native to our study region within Europe. This species  
174 was regarded as an atypical, outlier species, and thus excluded from our analyses.

175

### 176 *Statistical analysis*

177 We approached our analysis from different angles, using different statistical tools, to provide  
178 complementary evidence and visualizations on how urban tolerance was associated with species'  
179 traits. In brief, this involved: (1) correlation analysis to examine simple correlations among all  
180 traits; (2) multiple regression analysis to focus on understanding variation in urban tolerance and  
181 partial effects of other traits; (3) boosted regression trees to examine non-linearity and account  
182 for interactions among traits; and (4) clustering analysis to visualize the dominant trait clusters  
183 associated with an urbanization gradient.

184

185 *Correlation and Regression modelling.* First, for all numeric predictor variables (N=10), we  
186 assessed the pairwise relationships between urban tolerance and the predictor variables using  
187 Pearson correlation coefficients. Second, to assess the strength of the relationship between a  
188 given predictor variable and urban tolerance, accounting for the relationship of all other predictor  
189 variables, we used multiple linear regression with a Gaussian distribution. The response variable  
190 was urban tolerance, and the predictor variables (N=11) were: the average number of flight  
191 months, wing index, mean temperature in range, the number of adult food types (log<sub>10</sub>  
192 transformed), mean voltinism, the number of hostplant growth forms, the number of egg laying  
193 locations, hostplant specificity, egg laying type, hostplant index (log<sub>10</sub> transformed), and  
194 overwintering stage. Egg laying type was a categorical variable with three levels (single, small,  
195 and large batches) but was dummy-coded in the multiple linear regression because it showed  
196 little correlation with the response variable in exploratory analyses; we therefore did not assess  
197 differences among the levels of egg laying type. Parameter estimates from the model were  
198 standardized by centering and dividing 2 standard deviations (Gelman 2008). In addition to the  
199 large model with all the traits, we ran two separate linear regressions between urban tolerance  
200 and overwintering stage and hostplant growth form, respectively (see Table 1). These two traits  
201 were treated separately as each trait was associated with multiple binomial levels, and we wanted  
202 to avoid over-inflating the number of predictor variables in a single multiple linear regression. In  
203 each instance, the possible overwintering stages (i.e., egg, larval, pupal, adult) and possible  
204 hostplant growth forms (i.e., shrub, tall herb/grass, short herb/grass, and tree) were treated as  
205 binomial predictor variables in separate multiple linear regressions. For all three multiple linear  
206 regression models, we used weights in the model-fitting procedure where more weight was given  
207 to a species based on the number of observations of that species used to derive its urban

208 tolerance score, but the number of observations was capped at 1000 to ensure that our results  
209 were not driven by a few species with high weights.

210  
211 *Boosted regression trees.* We also performed a third analysis, using boosted regression trees  
212 (Elith et al. 2008). This analysis is advantageous because it allows for both linear and nonlinear  
213 relationships between urban tolerance and the ecological and life history traits of butterflies, as  
214 well as complex interactions among the predictor variables themselves. Because of the  
215 robustness of this analysis, we included all possible predictor variables from the three multiple  
216 linear regressions mentioned above (N=19), testing our entire suite of different predictions  
217 (Table 1). Although predictor variables do not need to be transformed for boosted regression  
218 trees (Elith et al. 2008), we kept the log<sub>10</sub>-transformed versions of hostplant index and the  
219 number of adult food types for consistency with the multiple linear regression modelling. First,  
220 we extracted the relative influence for each predictor variable, which shows the effect of each  
221 predictor variable on the response variable normalized to sum to 100 (Friedman 2001; Elith et al.  
222 2008). Second, for any variable that explained >5% of the total relative influence, we produced  
223 partial dependency plots that illustrate the influence of a given predictor variable accounting for  
224 the average effects of other predictor variables (e.g., Vilmi et al. 2019). The boosted regression  
225 tree analysis was performed using the *dismo* package in R (Hijmans et al. 2017). We used a tree  
226 complexity of 5, a learning rate of 0.001, and a bag fraction of 0.5 (e.g., Elith et al. 2008; Buston  
227 and Elith 2011; Vilmi et al. 2019). Exploratory analyses varying the level of tree complexity,  
228 learning rate, and bag fraction showed no difference in the quantitative or qualitative results.  
229

230 *Clustering analysis.* To characterize the trait values associated with the most typical patterns of  
231 species' urban tolerance, we used Generalized Additive Models (gams) in combination with  
232 clustering analysis. We used gams to model the presence/absence of species in 5 x 5 km grids  
233 within their distributional extent (delineated by the convex hull of their occurrence records) with  
234 urban cover in each grid as the predictor, as a spline term. A gam was fit to each species  
235 separately, assuming a binomial error distribution, and VIIRS within each species range was  
236 logged (to the base 10) and scaled between 0 and 1 for each species' gam. We used a spline to  
237 allow a non-linear relationship between species occupancy and VIIRS, and hence accommodate  
238 the diversity of possible species' urban response curves. However, we constrained the spline to a  
239 low number of knots ( $k=5$ ) to minimize biologically unrealistic multi-modal response curves  
240 from being fit. Using the fitted gam, we then predicted the occupancy probability of each species  
241 within grid cells of varying VIIRS values between 0 and 1 (in sequential steps of 0.05). Once we  
242 had characterized the response curve of each species to varying urban cover amounts (VIIRS),  
243 we then identified the most typical response curves using a clustering analysis. We first  
244 calculated a dissimilarity matrix among species' response curves. Since we were not interested in  
245 differences in the mean occupancy of species but rather relative differences in occupancy  
246 according to urban cover, we used a correlation-based dissimilarity metric (Pearson correlation  
247 coefficient). We then used hierarchical partitioning to split the dissimilarity matrix into discrete  
248 groups (i.e., clusters) of species sharing the most similar urban' response curves. To identify the  
249 most appropriate number of clusters, we compared several cluster metrics including Dunn's  
250 index, silhouette widths, and minimum cluster size and meaningful biological interpretation. For  
251 each cluster, we calculated the mean occupancy of species at each VIIRS value and bootstrapped

252 the species values to provide 95% confidence intervals. Finally, we visualized the distribution of  
253 species traits in each cluster to identify the suite of traits values associated with each.

254

#### 255 *Data analysis and availability*

256 All data analysis was conducted in R statistical software and relied heavily on the tidyverse  
257 (Wickham et al. 2019). Statistical significance, in the case of multiple linear regressions, was  
258 concluded at  $\alpha < 0.05$ . Code and data to reproduce these analyses are currently available here  
259 ([https://github.com/coreytcallaghan/butterfly\\_urbanness](https://github.com/coreytcallaghan/butterfly_urbanness)) and will be permanently archived in a  
260 Zenodo repository upon acceptance of this article.

261

## 262 RESULTS

263 We used a total of 922,687 observations for 158 species to position each species' urban tolerance  
264 along an urbanization-tolerance continuum. The mean number of observations per species was  
265 5840 ( $\pm 9748$  SD). A total of 125 species (79%) had an urban tolerance score  $< 0$ , suggesting that  
266 they disproportionately use less urbanized habitat in comparison with that available within their  
267 range. The mean urban tolerance score was  $-0.73$  ( $\pm 1.60$ ) (Figure 1). The most urban tolerant  
268 species was *Polygonia egea* (urban score=5.97), followed by *Satyrium w-album* (urban  
269 score=4.29), *Thecla betulae* (urban score=3.56), and *Pieris rapae* (urban score=3.44). In  
270 contrast, the least urban tolerant species were *Euphydryas maturna* (urban score=-4.37),  
271 *Muschampia proto* (urban score=-3.84), *Hipparchia fidia* (urban score=-3.77), and  
272 *Glaucopsyche melanops* (urban score=-3.63) (Figure 1). For an interactive version, showing the  
273 urban tolerance scores for the 158 species included in analysis, see [here](#).

274

275 Pairwise relationships between the urban tolerance score and ecological and life history traits  
276 (Figure 2; Figure 3) showed that urban tolerance was positively correlated with all variables  
277 aside from hostplant specialism index. In particular, urban tolerance was strongly correlated with  
278 the average number of flight months ( $r=0.53$ ) and mean voltinism ( $r=0.45$ ), and less weakly  
279 correlated with the number of adult food types ( $r=0.28$ ). There was weak positive correlation  
280 between mean temperature in range ( $r=0.11$ ) and urban tolerance, and there was a negative  
281 relationship between urban tolerance and hostplant specialism index ( $r=-0.24$ ). Overall, our  
282 predictions matched the expected relationship for our numeric variables (cf Table 1 and Figure  
283 3a).

284

285 Our multiple linear regression explained the variance in urban tolerance reasonably well  
286 ( $R^2=0.38$ ), showing that there was strong evidence (i.e., confidence intervals did not overlap  
287 zero) that the average number of flight months and the number of adult food types were the most  
288 important traits associated with urban tolerance (Figure 3b). The other traits were not  
289 significantly associated after accounting for the effects of these two traits. However, weak  
290 evidence was found for a positive relationship between mean voltinism, wing index, and number  
291 of hostplant growth forms and urban tolerance. A separate multiple linear regression for binomial  
292 traits of hostplant growth forms showed that species associated with all four types of hostplant  
293 growth forms were more likely to be tolerant of urban environments, but there was strong  
294 evidence for species that associated with shrub hostplant and tall herb/grass (Figure S5). For the  
295 overwintering stage, a separate multiple linear regression showed that species overwintering as  
296 adults and pupae showed a positive relationship with urban tolerance, whereas species that  
297 overwinter as larvae or eggs showed a negative relationship with urban tolerance. There was

298 strong evidence that species that overwinter in the larval stage are negatively associated with  
299 urban tolerance (Figure S6).

300  
301 Boosted regression tree analysis showed that our predictor variables explained 23.8% of  
302 deviances in urban tolerance of butterflies. The most important predictor variables — those that  
303 explained >5% of relative influence — were the average number of flight months (35.5%), mean  
304 temperature in a species range (14.5%), hostplant specialism index (11.4%), wing index (10.7%),  
305 overwintering stage as larvae (7.1%), and the number of adult food types (6.3%) (Figure 3c;  
306 Figure 3d). The boosted regression tree analysis showed the non-linear patterns in these predictor  
307 variables. For the average number of flight months, there were marginal gains in urban tolerance  
308 from ~ 4–6 flying months per year, but then from ~6–8 there was a strong increase in the  
309 relationship with urban tolerance. Hostplant specialism index showed a generally smooth decline  
310 in its association with urban tolerance, and wing index showed a non-linear positive response  
311 with urban tolerance. In contrast, mean temperature in a species range showed a non-linear  
312 response with a positive association from about 0 degrees Celsius to 10 degrees Celsius,  
313 followed by a negative association with urban tolerance from about 10 degrees Celsius to 15  
314 degrees Celsius (Figure 3d; Figure 2).

315  
316 Our cluster analysis of responses to urbanization supported an ecological interpretation of three  
317 main clusters generalizing the diversity of species-specific responses to urbanization along a  
318 gradient of urbanization (Figure 4a), showing relatively strong agreement with our urban  
319 tolerance scores (Figure S7). Cluster 1 (N=25 species) grouped together species most common in  
320 high urban areas — i.e., urban exploiters; cluster 2 (N=46 species) grouped together species most

321 common at intermediate levels of urbanization — urban adapters; and cluster 3 (N=87 species)  
322 grouped together species that were most common at low urban areas and rarely occurred outside  
323 of low urban areas — i.e., urban avoiders (Figure 4a; Table S2). When these clusters were  
324 mapped onto species-specific traits, we found a general increase from cluster 3 (least urban  
325 tolerant) to cluster 1 (most urban tolerant) in the number of average food types eaten by adults  
326 (Figure 4b) and the number of average flight months (Figure 4e). Typically, species in cluster 1  
327 had a flight period of ~ 7 months, while species in clusters 2 and 3 were flying ~ 3–5 months  
328 during the year, on average. Also, the number of adult food types was typically 3 for species in  
329 cluster 1, but fewer than 3 types in the other clusters. We also found a general decrease from  
330 cluster 3 to cluster 1 in the hostplant specialism index values (Figure 4c). There were no apparent  
331 differences among clusters for the mean temperature in range, yet the most warm-adapted  
332 species tended to be captured in cluster 1 (Figure 4d).

333

## 334 DISCUSSION

335 We integrated a continuous measure of urbanization — VIIRS night-time lights — with over  
336 900,000 species' observations from GBIF to derive a comprehensive analysis of species-specific  
337 (N=158 species) responses of butterflies to urbanization across continental Europe. The majority  
338 of butterfly species included in our analysis were shown to avoid urban areas (Figure 1; Figure  
339 S7), regardless of whether species' tolerances were quantified as a single mean score (79% of  
340 species avoided urban areas) or as a species' response curve to the whole urbanization gradient  
341 (55% of species). Together, these results help to explain the reduced taxonomic diversity of  
342 butterflies in urban ecosystems (e.g., Pignataro et al. 2020; Fenoglio et al. 2020; Kurlyo et al.  
343 2020) and highlight which species should be the focus of active conservation in urban areas (see



344 Table S2). Still, a reasonable number of species (25 species) were shown to be more common in  
345 urban areas than elsewhere (Figure 4a), hinting at which species might be the winners of  
346 anthropogenic change as urban areas continue to expand. Overall, our results demonstrate that  
347 generalist life histories enable butterfly species to tolerate urban areas, whether generalism is  
348 defined in terms of climate or diet preferences.

349

350 We found support that climate, or thermal, flexibility was linked with urban tolerance among  
351 European butterflies. The average number of flying months was consistently the strongest and  
352 most important predictor of urban tolerance across our different analyses. In temperate Europe,  
353 species with long flight periods during the year, typically over multiple seasons, have to cope  
354 with a range of climatic conditions and hence may have a broader thermal niche breadth. The  
355 relationship between urban tolerance and thermal preferences or flexibility has been found in  
356 other taxa as well, including ants (Diamond et al. 2017), trees (Kendal et al. 2018), lizards  
357 (Campbell-Staton et al. 2020) and birds (Deutsch et al. 2008; Clavero et al. 2011; Barnagaud et  
358 al. 2012). Our results, combined with previous literature, support the general notion that species  
359 with broad environmental tolerance may prosper in urban environments (Bonier et al. 2007):  
360 those species have the necessary flexibility to succeed in the unique and novel environmental,  
361 physiological, and/or ecological attributes of urban environments. We also found that species  
362 with longer flight periods were likely to be bi- or multi-voltine, explaining why voltinism was  
363 also somewhat associated with urbanization in our analysis. More generally, our results support  
364 the hypothesis that human-dominated habitats may pose a thermal challenge for much of  
365 biodiversity (Daily and Ehrlich 1996).

366

367 The mean temperature within a species' range, previously used as a measure of thermal  
368 preference in butterflies (Devictor et al. 2012), explained some variability in urban tolerance  
369 among species. Urban areas are typically warmer than their surroundings because of the urban  
370 heat island effect. Hence, species that tolerate the negative effects of urbanization also have to  
371 tolerate the warmer mean temperatures within urban areas. But because urban areas, as measured  
372 in our analysis, can include urban cold islands as well, thermal flexibility may be more important  
373 than the mean temperature in a species' range. In our clusters of species responses to  
374 urbanization, cluster 1, comprising the most urban tolerant species, also included some of the  
375 most warm-adapted species in our analysis such as *Euchloe belemia*, *Polygonia egea*, and  
376 *Lampides boeticus* (Figure 4). However, the relationship between the mean temperature within a  
377 species' range and urban tolerance was non-linear and inconsistent across our other analyses.  
378 This might be because butterflies can respond to local microclimatic variation (Horner-Devine et  
379 al. 2003), and some species that are warm-adapted but not urban tolerant (e.g., *Charaxes jasius*,  
380 *Aricia cramera*, and *Pseudophilotes panoptes*) are using habitat at a scale not captured by our  
381 analysis. Because small invertebrates are more susceptible to local climatic conditions than  
382 larger-sized taxa, such as birds and mammals, the urban heat island effect may moderate some of  
383 the negative impacts of urbanization (Kaiser et al. 2016), especially in temperate regions where  
384 invertebrates are predicted to commonly experience temperatures below their thermal optimums  
385 (Deutsch et al. 2008). As climate change continues, species living in urban areas will have to  
386 tolerate even warmer temperatures, including heatwave events and summer droughts in Europe.  
387 As a result, voltinism and advances in phenology in butterflies are favored by climate change  
388 (Altermatt 2009). Therefore, associations between climate and urban tolerance suggest that  
389 selection pressures from climate warming may also foster urban tolerant species. Further

390 disentangling the relationship between urban tolerance and thermal tolerance and flexibility will  
391 remain an important goal for understanding the influence of urbanization on butterflies and  
392 identifying the winners and losers of increasing urbanization.

393

394 In addition to the importance of thermal flexibility, we found that urban tolerance in butterflies  
395 was positively associated with diet generalism, confirming previous research that has  
396 demonstrated the link between diet and phenology in butterflies (Altermatt 2010). Diet  
397 generalism at both adult (i.e., the number of adult food types) and larval (i.e., hostplant  
398 generalism) life stages were important for tolerating urban ecosystems (Figure 3, Figure 4). This  
399 suggests that considering the influence of different life history stages (e.g., egg, larval, pupal, or  
400 adult) in how species adapt to urban environments may be important in future work. For most  
401 butterfly species, the larval stage is longer than the adult stage, and often larval food resources  
402 are thus more important in the butterfly life cycle (Altermatt and Pearse 2011). In support of this  
403 general pattern, we found that larval resources (i.e., hostplant specialism) were marginally more  
404 important than adult resources (Figure 3) in predicting urban tolerance. Tolerance to urbanization  
405 may be especially challenging for species that use different resources and habitats during their  
406 life cycle, including holometabolous insects, compared with other taxa with more uniform  
407 resource requirements during their lifespan. These differences among life history strategies could  
408 be linked to the differential impacts of thermal tolerance and local climatic events among life  
409 history strategies (Long et al. 2016). Alternatively, because urban environments can sometimes  
410 have greater species richness in plants, due in part to the prevalence of non-native plant species,  
411 species with diet generalism across life stages may be able to take advantage of this unique  
412 attribute of urban ecosystems.

413  
414 Overwintering strategy and dispersal ability played more minor roles in a species' ability to use  
415 urban environments. Butterfly species' responses to climate has been previously shown to  
416 depend on their overwintering strategy (Long et al. 2016). We found that species overwintering  
417 as adults were more urban-tolerant than species overwintering as eggs (Figure 3b). And a  
418 separate analysis showed that species overwintering as adults or pupae were positively associated  
419 with urban tolerance whereas overwintering as eggs and larvae were negatively associated with  
420 urban tolerance (Figure S6). Species that overwinter as adults are typically those able to begin  
421 reproducing earlier in the season, whereas those overwintering as larvae must first undergo  
422 metamorphosis. Hence, this result is also consistent with the positive effect of the number of  
423 flight months on species tolerance to urbanization. Our results also showed that body size, as  
424 measured by wing index, was somewhat positively associated with urban tolerance. Body size in  
425 butterflies is linked to dispersal ability (Stevens et al. 2011; Sekar et al. 2011; Middleton-Welling  
426 et al. 2020) and climate tolerance (Klockmann et al. 2016), suggesting that both these traits  
427 probably interact to explain the moderate evidence we found that body size predicts urban  
428 tolerance among butterflies.

429  
430 Our analysis was focused on butterfly responses to urbanization at a macro-ecological scale,  
431 using a globally-applicable remotely-sensed product of urbanization at a native resolution of  
432 ~500 meters (Elvidge et al. 2017). However, urbanization processes happen at multiple spatial  
433 scales, ranging from local to landscape levels (Concepción et al. 2015; Piano et al. 2019), and  
434 biodiversity responses to urbanization may differ among these spatial scales (Merckx and Van  
435 Dyck 2019). Butterflies can select habitat at fine-grained spatial scales within urban ecosystems

436 smaller than 500 meters (e.g., Kaiser et al. 2016), such as urban meadows (Dylewski et al. 2019)  
437 or revegetated road verges (Saarinen et al. 2005; Valtonen et al. 2007). Indeed, the spatial  
438 resolution of our analysis likely explains why we found weak support for micro-scale habitat  
439 predictors such as the habitat of hostplant types or egg-laying location types. These traits may be  
440 important for predicting space use within urban areas, but not urban tolerance as measured in our  
441 current analysis. Future work should formally test how species-specific responses to urbanization  
442 varies among spatial scales in butterflies (e.g., Moll et al. 2020; Callaghan et al. 2020). In  
443 addition to our limitations in the spatial resolution, we highlight that we only looked at urban  
444 preferences in butterflies averaged across the full annual cycle, but some species may increase  
445 their use of urban areas during certain times of the year. For example, some species may move  
446 into urban areas during late fall when the surrounding temperatures drop, taking advantage of the  
447 urban heat island effect (Kaiser et al. 2016). Future work should investigate patterns in urban  
448 tolerance of butterflies across the full annual cycle (Marra et al. 2015). Finally, we treated  
449 phenology as a fixed trait in our analysis but in reality, species' phenology can vary among years  
450 and places. Indeed, phenology might also vary with urbanization, with warmer temperatures  
451 within urban areas allowing some butterflies to appear earlier in the year (but see Diamond et al.  
452 2014).

453  
454 Butterflies are popular with the non-scientific public and provide many cultural ecosystem  
455 services (e.g., McGinlay et al. 2017), particularly within urban ecosystems where they are most  
456 likely to be encountered even by casual observers. Butterflies, therefore, might play important  
457 roles in minimizing 'extinction of experience' for humans who are becoming increasingly  
458 concentrated in urban areas (Soga and Gaston 2016). Conserving urban biodiversity, including

459 butterflies, is increasingly important in urban conservation planning. An important first step in  
460 this process is understanding the species that are tolerant and intolerant of urban ecosystems. We  
461 provide a method to efficiently quantify the urban tolerance of butterflies at a macro-ecological  
462 scale and accomplished this for 158 species of European butterflies. As data in GBIF continues  
463 to grow, largely due to citizen science efforts (Chandler et al. 2017), our analysis here can be  
464 updated for the remaining European butterfly species. Nonetheless, we provide strong evidence  
465 that generalism, in terms of both climate and diet, is inherently linked with urban tolerance and  
466 that generalist species are best-adapted to urban ecosystems. Our findings suggest that the  
467 majority of European butterfly species avoid highly urbanized areas, highlighting the need to  
468 include greening strategies in urban planning and conservation decisions (Ramírez-Restrepo and  
469 MacGregor-Fors 2017).

470

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476

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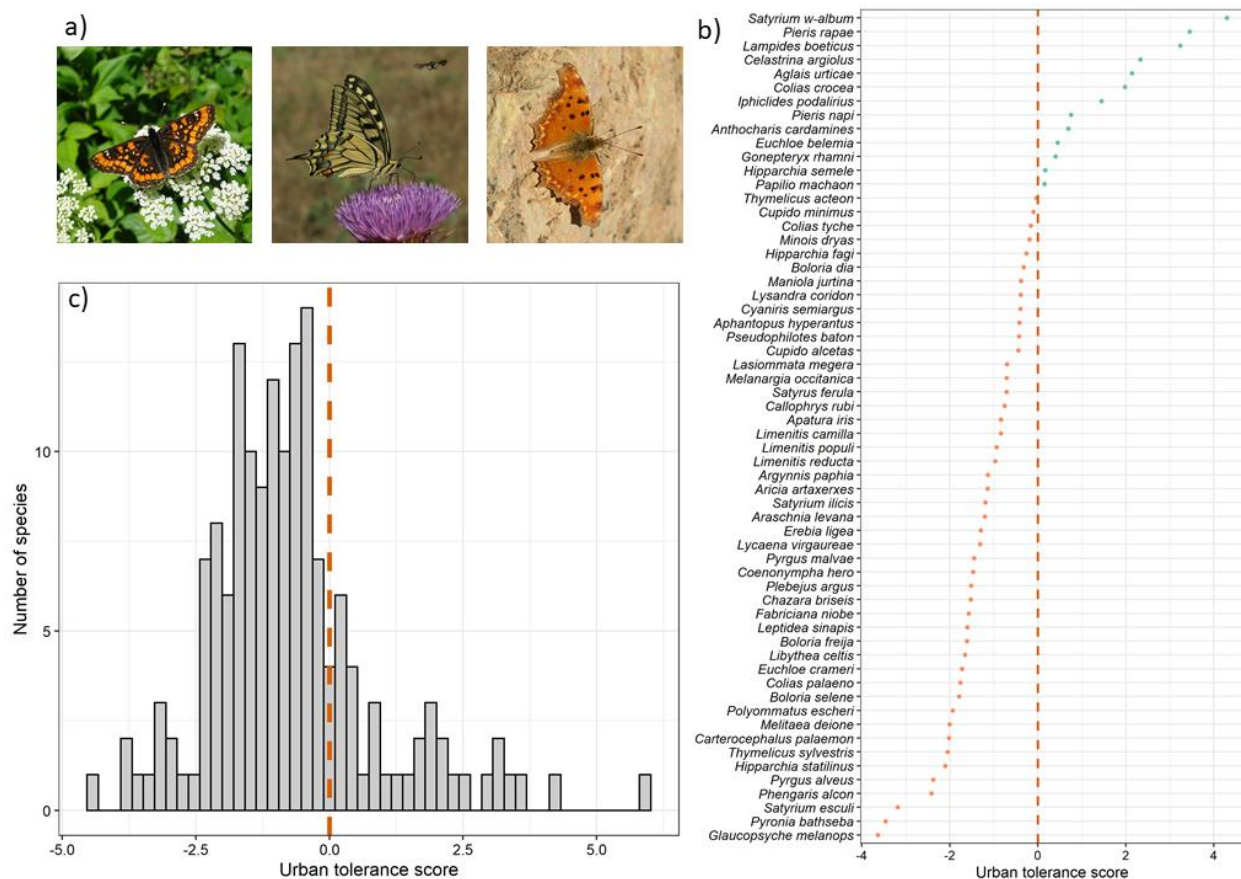
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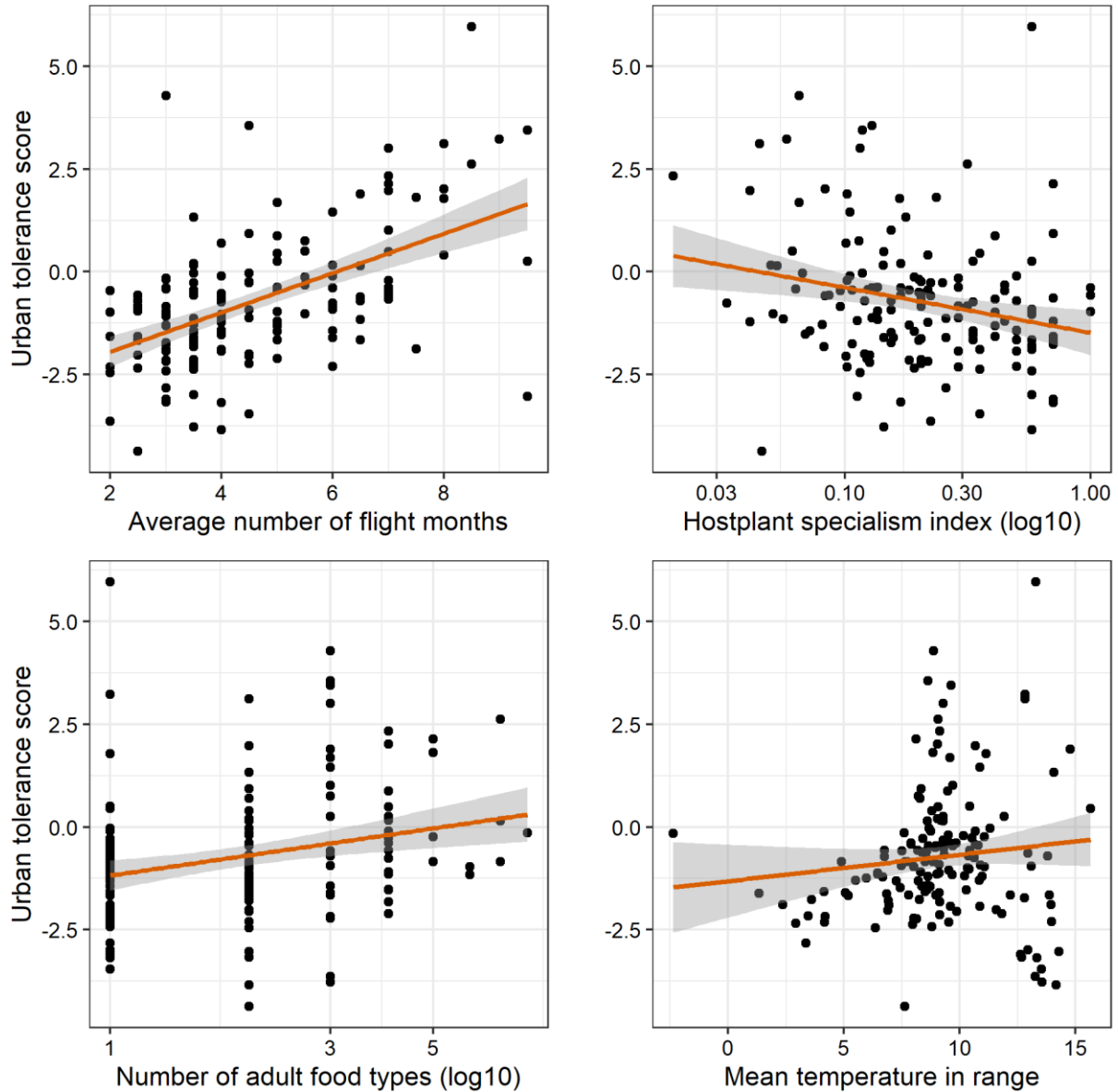
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## 728 FIGURES

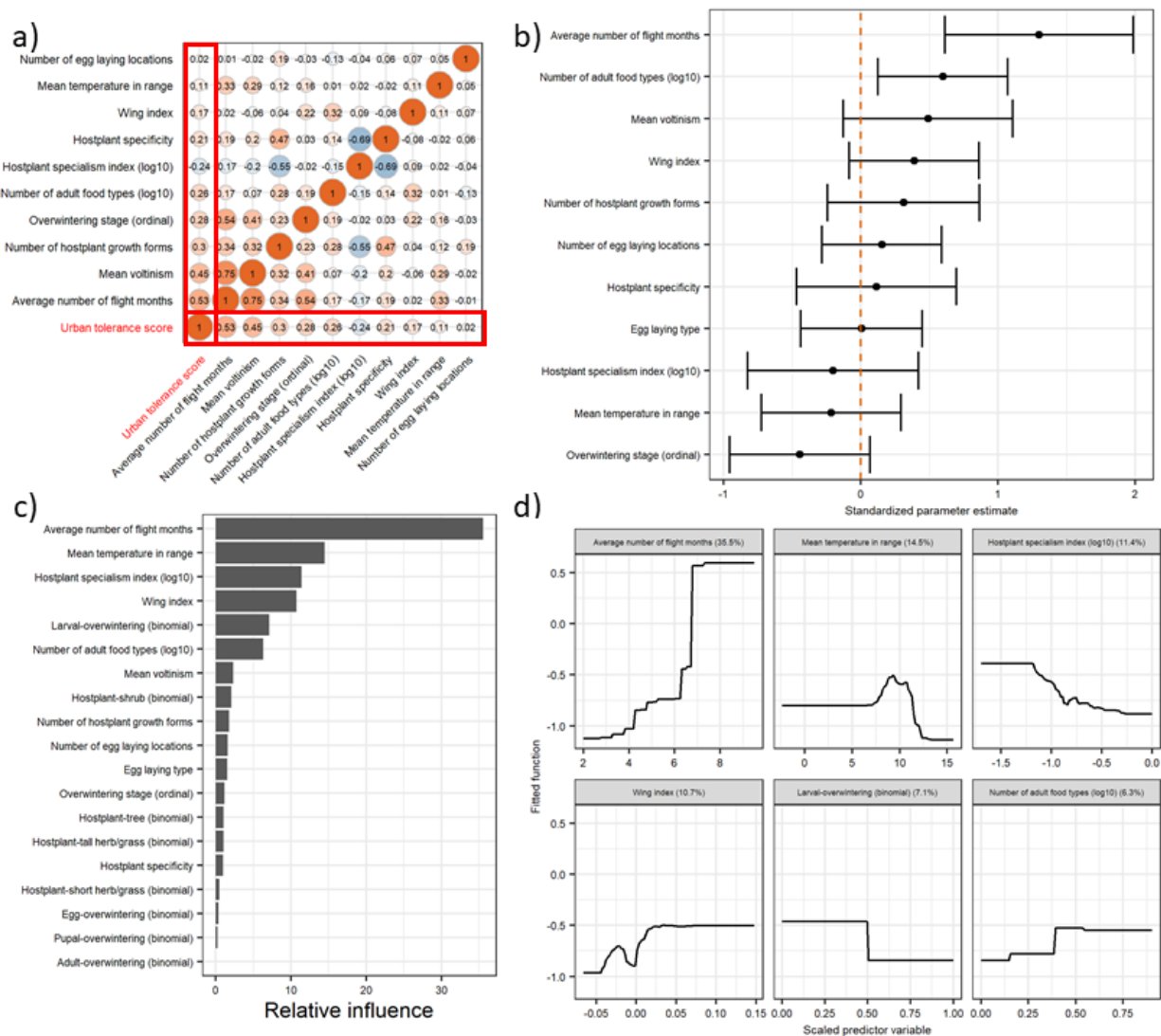


729  
 730 **Figure 1.** a) Three species included in our analysis, ordered from left to right in terms of their  
 731 urban tolerance scores: Scarce Fritillary (*Euphydryas maturna*) with an urban tolerance score of -  
 732 4.37; Old World Swallowtail (*Papilio machaon*) with an urban tolerance score of 0.15; Southern  
 733 Comma (*Polygonia egea*) with an urban tolerance score of 5.97. All photos by Julia Wittman  
 734 (@birdingjulia) and are CC-BY-NC. b) Example of the rankings for 60 randomly chosen  
 735 butterflies, ranked from those that were found proportionately in more urbanized areas (above 0)  
 736 to those found proportionately in less urbanized areas (below 0). For a full interactive figure  
 737 showing all 158 species considered in analysis [see here](#). c) A histogram of the urban tolerance  
 738 scores for all 158 species included in the analysis.  
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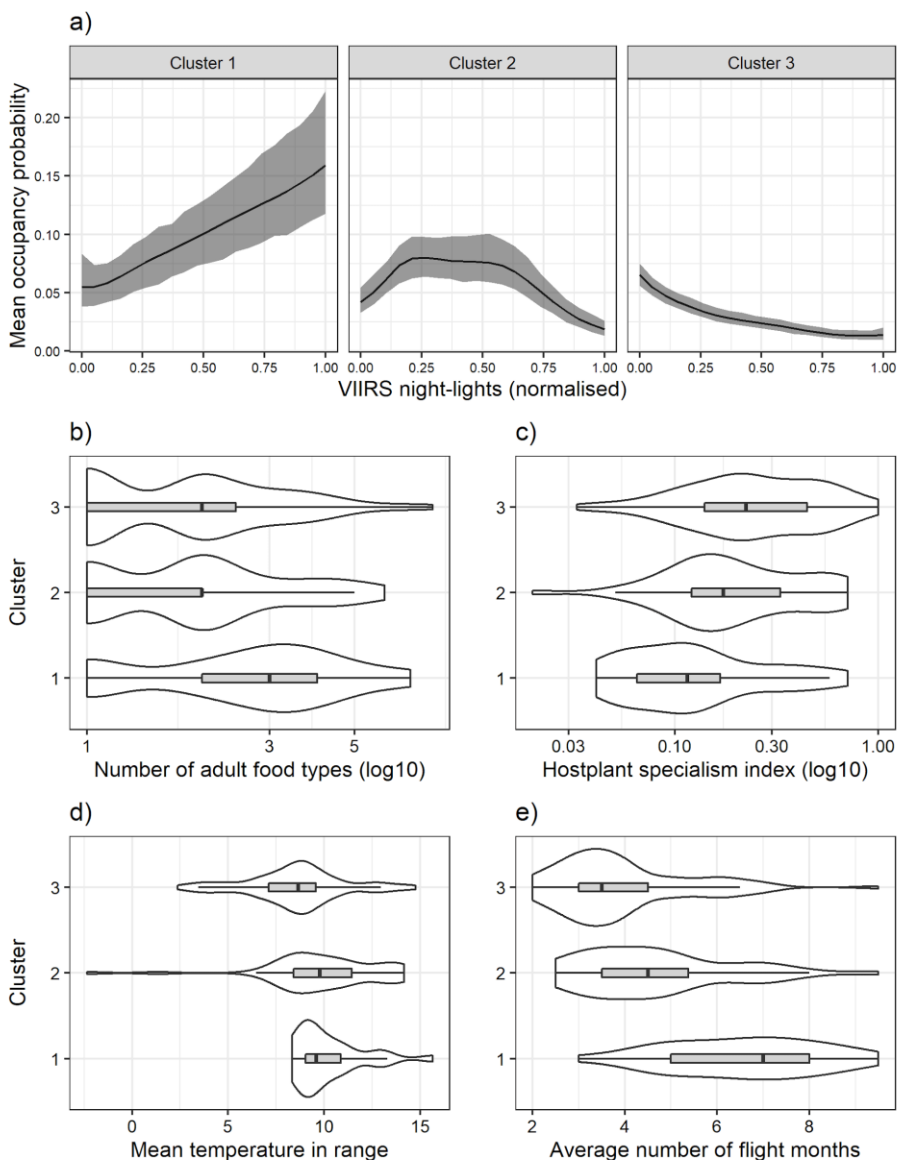
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**Figure 2.** The relationship between our urban tolerance score for N=158 species of butterfly, and the average number of flight months (top left), hostplant specialism index (top right), number of adult food types (bottom left), and mean temperature in a species' range (bottom right). The orange line represents a simple linear model fit, and the shaded gray area represents a 95% confidence interval around the linear model fit.



747

748 **Figure 3.** Results of our statistical analysis quantifying the relationship between urban tolerance  
 749 score of butterflies (N=158) and various predictor variable (see Table 1). a) Correlation plot of  
 750 all numeric predictor variables (N=10) and our response variable (in red text). Variables are  
 751 ordered left to right by the strength of their pairwise relationship with the response variable. b)  
 752 Results of our multiple linear regression and standardized parameter estimates with 95%  
 753 confidence intervals. Variables to the right of the vertical orange line positively interacted with  
 754 urban tolerance whereas variables to the left of the orange line negatively interacted with urban  
 755 tolerance. c) and d) Results from our boosted regression tree analysis, with c) representing the  
 756 relative influence of all predictor variables (N=19) included in the model, ordered from the  
 757 variable with the most relative influence to the least, and d) shows the partial dependence plots  
 758 for all predictor variables that had >5% relative influence on the urban tolerance of butterflies.



759  
 760 **Figure 4.** Results of our cluster analysis, and the three normalized responses to urbanization (a),  
 761 for each cluster respectively. The clusters mapped to four traits (b-e), confirming the importance  
 762 of these traits for urban tolerance among the species within each respective cluster. Cluster 1  
 763 (N=25 species) grouped together species most common in high urban areas — i.e., urban  
 764 exploiters; cluster 2 (N=46 species) grouped together species most common at intermediate  
 765 levels of urbanization — urban adapters; and cluster 3 (N=87 species) grouped together species  
 766 that were most common at low urban areas and rarely occurred outside of low urban areas — i.e.,  
 767 urban avoiders (see Table S2 for the species corresponding to each cluster).  
 768

769 TABLES

770

771 **Table 1.** A summary of the traits included in analyses, as well as our prediction for each trait. All data were extracted from Middleton-  
 772 Welling et al. 2020 except for the mean temperature in a species range which was extracted from Schweiger et al. 2014.

773

Category	Trait	Description	Prediction
Climate tolerance	Average number of flight months	The average number of months of the year a species is observed flying, taken as the average of the minimum and maximum number of flight months observed for each species.	We expected that species that had a greater number of flight months would be positive associated with urban tolerance.
	Overwintering stage (ordinal)	Originally a categorical variable, corresponding to the overwintering stage for a species, where the options are egg, larva, pupa, or adult. We converted the possible combinations of these categorical variables into an ordinal variable ranging from 1 (egg) to 4.5 (adult).	We expected that the ordinal overwintering stage variable would be positive associated with urban tolerance, as species that overwintered as adults would be more likely to be urban tolerant.
	Overwintering stage (binomial)	We also treated overwintering stage in a separate analysis where each categorical option was treated as a binomial predictor variable.	We expected that species which overwinter as adults would be the most urban tolerant, followed by species that overwinter as pupae, larvae, and eggs.
	Mean temperature in range	The mean temperature within a species range.	We expected that species with a higher mean temperature in their range would be positively associated with urban tolerance.
Extent of specialization	Number of adult food types	Eight possible adult food types were presented by Middleton-Welling et al. 2020: herbs, flowers, ergot, shrub/tree flower, honeydew, sap, decaying plant, animal, and mineral. We used the total number of categories an adult species feeds on, with a highest possible value of 8, and lowest of 1.	We expected a positive relationship between the number of adult food types and urban tolerance.
	Hostplant specificity	An ordinal variable corresponding with the range of host plants a species can use, ordered as monophagous species (1), narrow oligophagous	We expected a positive relationship between hostplant specificity and urban tolerance.

		(2), broad oligophagous (3), and polyphagous (4). See details in Middleton-Welling et al. 2020.	
	Hostplant index	An index ranging from 0 to 1, providing a quantitative measure of overall hostplant specificity, where 1 is most specific. See Middleton-Welling et al. 2020 for details of this calculation.	We expected a negative relationship between the hostplant index and urban tolerance: more specialized species would be least urban tolerant.
Body size	Wing index	A composite variable representing a single measurement of overall size for all butterfly species generated from forewing length and wingspan measures, for both males and females. See Middleton-Welling et al. 2020 for details of this calculation.	We expected a positive relationship between wing index and urban tolerance.
Life history	Mean voltinism	A measure of the number of generations a species has in a year. We took the mean value between the minimum and maximum voltinism measures provided by Middleton-Welling et al. 2020.	We expected a positive relationship between mean voltinism and urban tolerance.
	Egg laying type	A categorical variable representing three types of egg-laying strategies: single egg, small batch, and large batch. Some species may lay single eggs or small batches, and we used the largest possible category for each species.	We expected a positive relationship between the number of eggs a species lays with urban tolerance.
Microhabitat use	Number of hostplant growth forms	A variable representing the total number of growth forms of a species' hostplants, ranging from 1 to 5. The five categories of species' hostplants were short herb/grass (<1m), tall herb/grass (>1m), shrub, tree, and liana.	We expected a positive relationship between the number of hostplant growth forms and urban tolerance.
	Hostplant growth form (binomial)	We also treated hostplant growth form in a separate analysis where each categorical option was treated as a binomial predictor variable. But because so few species in our analysis used liana, this was not included as a variable.	We expected that species which use herbs/grass would be more positively associated with urban tolerance.



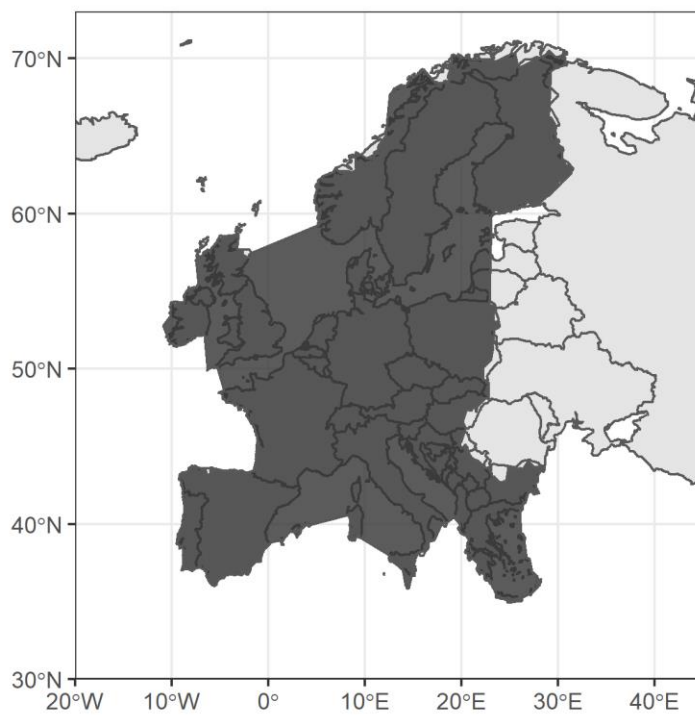
	Number of egg laying locations	A variable representing the total number of unique structures that eggs are laid on by a particular species, with a highest possible value of 7 for the most general, and 1 for the most specific. The 7 categories provided by Middleton-Welling et al. 2020 are bare ground, short turf/herbs/grass (<1m), tall herbs/grass (>1m), shrub, tree trunk, canopy, and liana.	We expected a positive relationship between the number of egg laying locations with urban tolerance.
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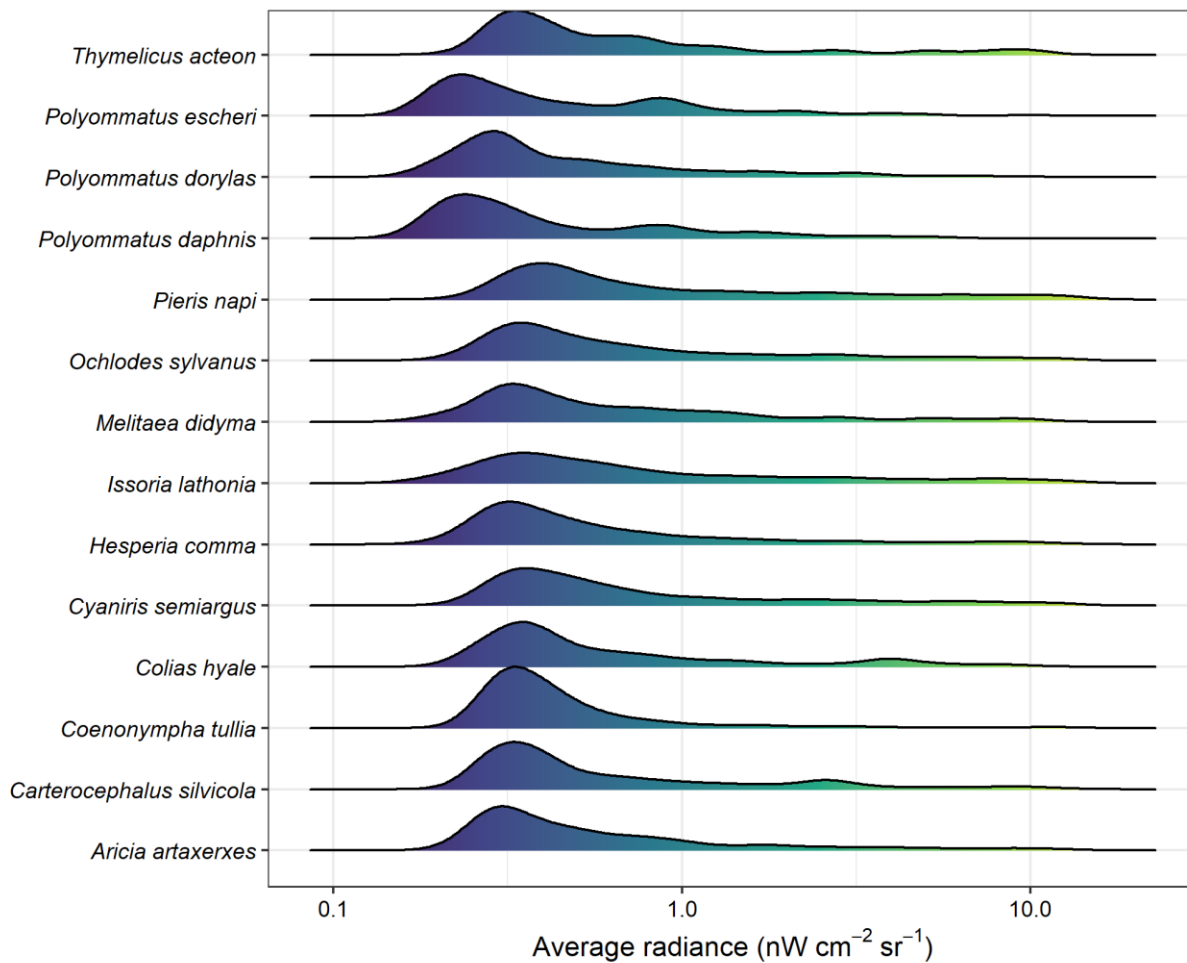
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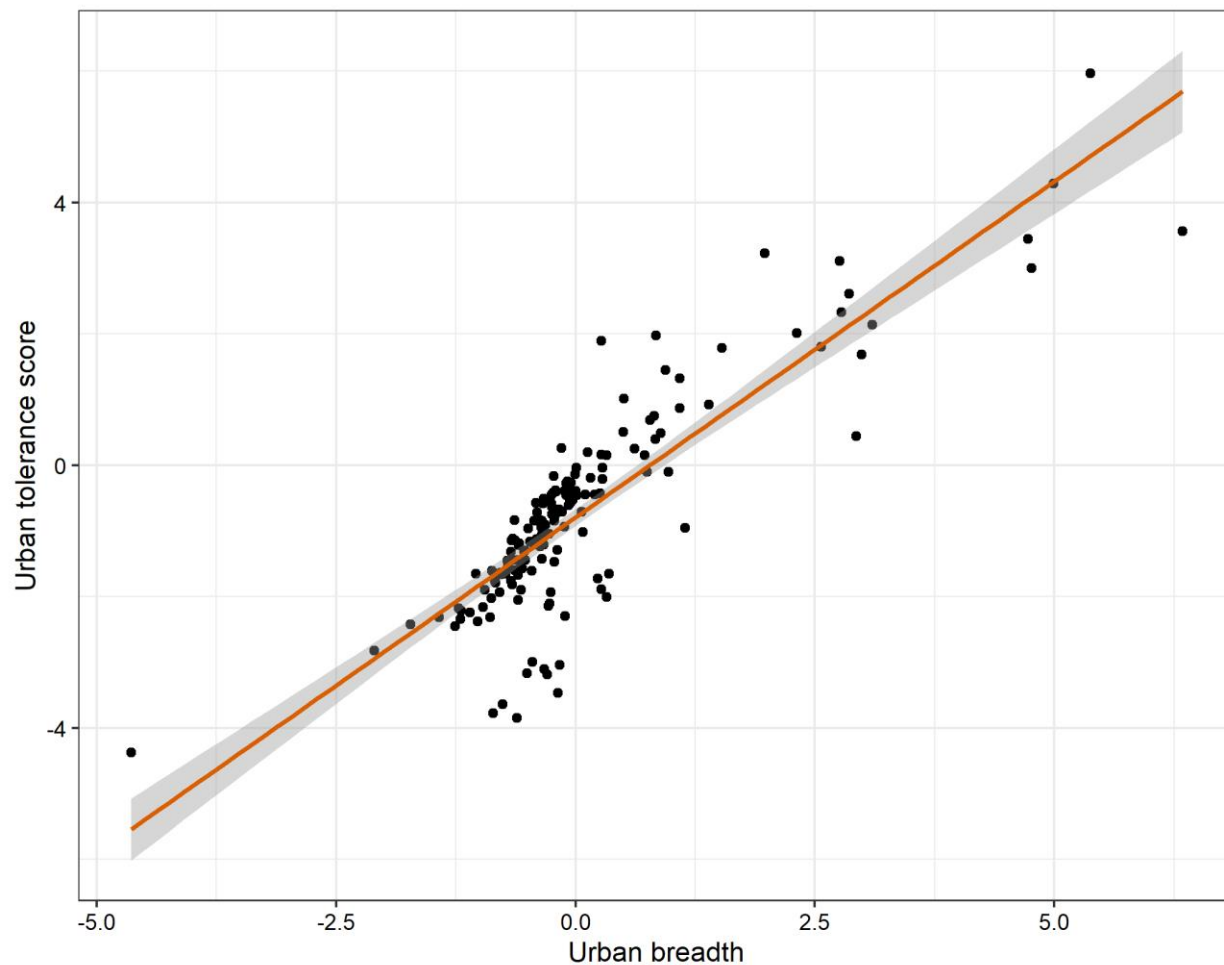
## SUPPLEMENTARY FIGURES



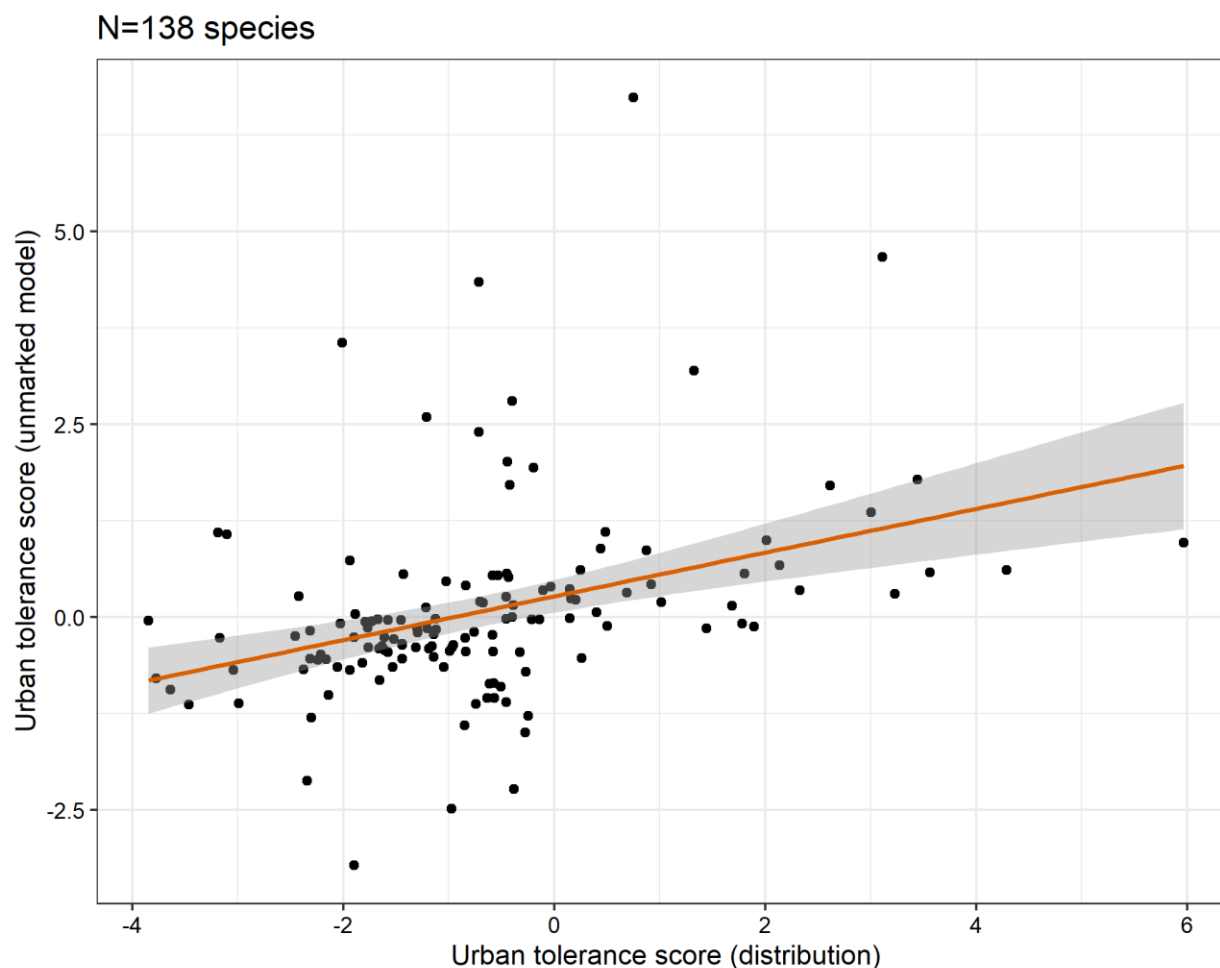
**Figure S1.** A map of the study extent that was manually delineated using the observations from GBIF. The concave map was made using the ‘concaveman’ package in R which is a R port for a mapbox library of the same name, and the default concavity of 2 was used to make a polygon surrounding our point observations from GBIF.



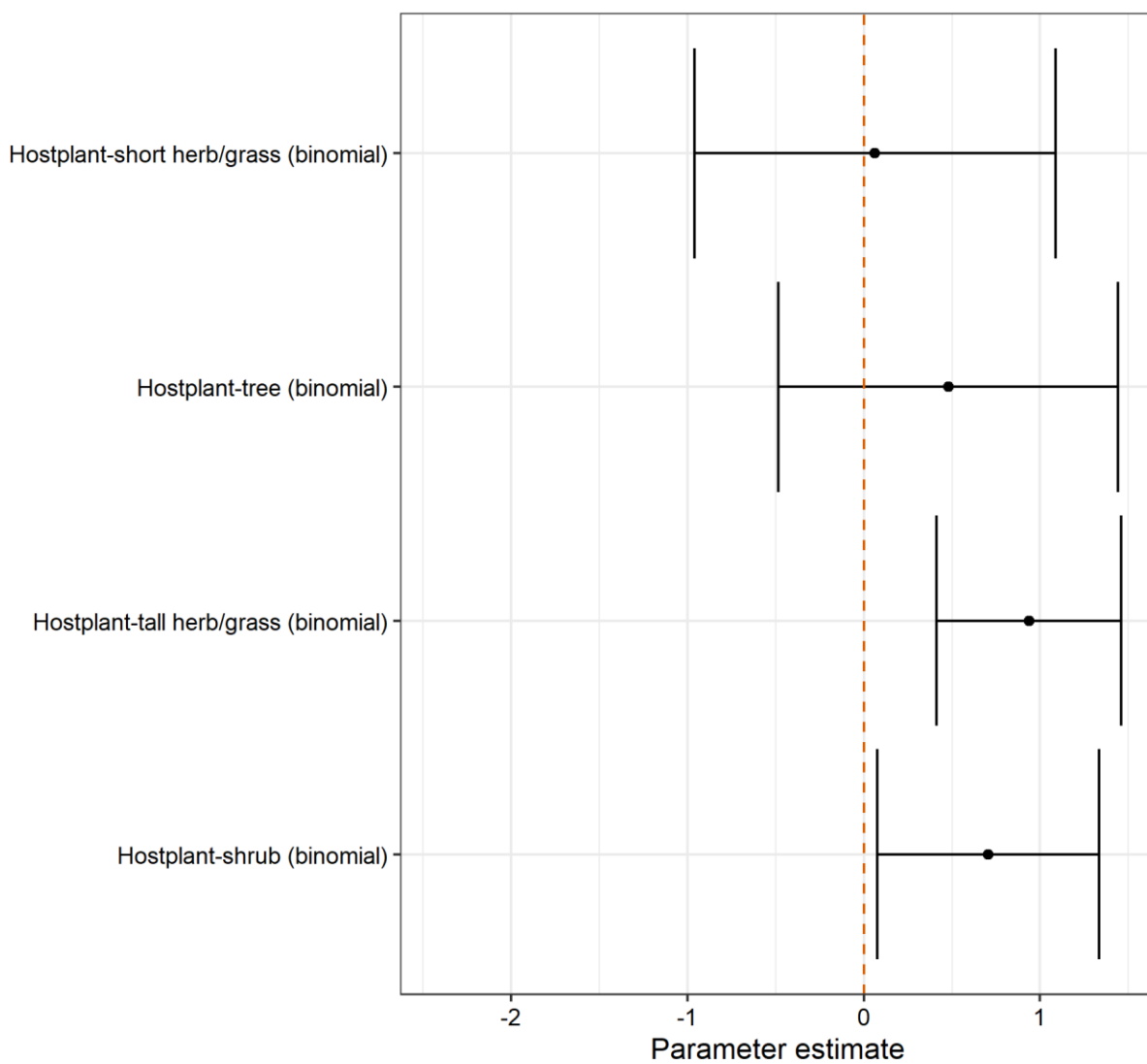
**Figure S2.** Fifteen example species, and their distribution in response to VIIRS night-time lights.



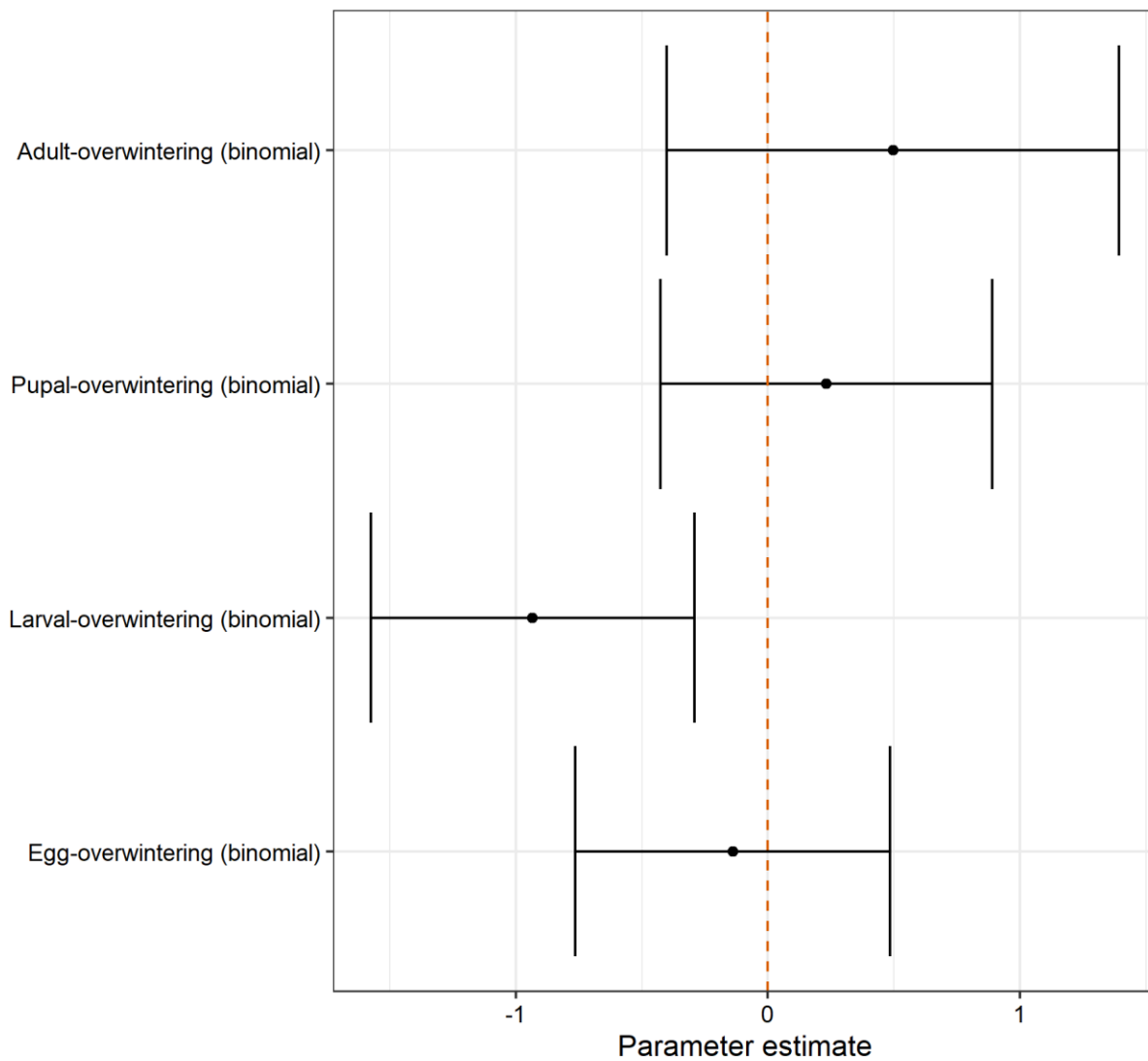
**Figure S3.** The relationship between urban tolerance (the difference between the mean of all observations within a species' range and the observations of each species within their range) and urban breadth (the difference between the interquartile range of all observations within a species' range and the observations of each species within their range) showed a strong positive relationship. Shown here are all species (N=158) included in analysis. Because of this relationship, our analysis focused on the urban tolerance score throughout.



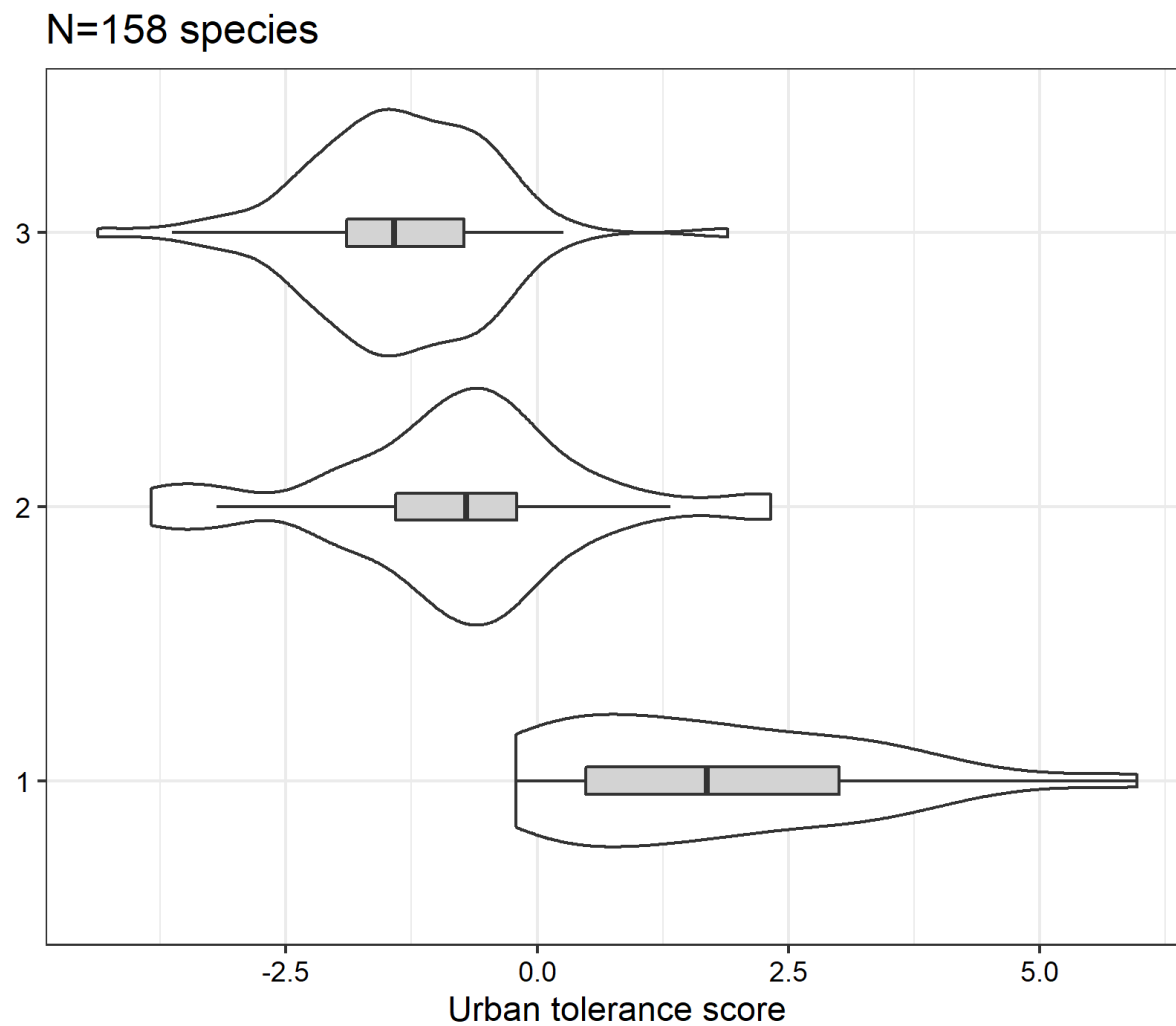
**Figure S4.** To confirm the reliability of our distributional approach, we performed a separate analysis to rank species along a continuum of urban tolerance/preference using unmarked models. For this analysis, species were grouped by 5km grid across Europe, and week of the year, into presence/absence, and each species was only calculated within its concave hull range. The mean VIIRS night-time lights level was also calculated within each 5km grid. An unmarked model was ran which accounted for the effects of urban cover (VIIRS) on the detection probability of species. Only species with a standard error of their modelled response to urbanization  $< 2$  was included in the comparison between the two approaches (N=138 species).



**Figure S5.** Results of a separate multiple linear regression which investigated the binomial predictor variables of hostplant growth form with the urban tolerance score response variable.



**Figure S6.** Results of a separate multiple linear regression which investigated the binomial predictor variables of overwintering stage form with the urban tolerance score response variable.



**Figure S7.** The results of our two different methods to quantify urban tolerance of butterflies throughout Europe.



## SUPPLEMENTARY TABLES

**Table S1.** The countries included in analysis and the number of observations included for potential analyses in each country.

Country	ISO country code	Number of samples	Species richness
Sweden	SE	565214	126
Ireland	IE	144570	43
France	FR	83491	247
Germany	DE	51761	149
Spain	ES	36110	221
Portugal	PT	20614	116
Italy	IT	12791	220
Switzerland	CH	11128	162
Greece	GR	10762	173
Austria	AT	7113	149
Bulgaria	BG	5340	164
Slovenia	SI	4949	128
Norway	NO	4385	90
North Macedonia	MK	2919	140
Poland	PL	2919	91
United Kingdom	GB	2422	44
Hungary	HU	2411	108
Luxembourg	LU	1777	54
Croatia	HR	1492	107
Finland	FI	1205	69
Czechia	CZ	894	60
Denmark	DK	713	58
Slovakia	SK	626	72
Andorra	AD	562	115
Montenegro	ME	418	91
Albania	AL	253	76
Bosnia & Herzegovina	BA	139	56
Serbia	RS	135	57
Malta	MT	117	13
Liechtenstein	LI	63	26
Gibraltar	GI	50	17
Belgium	BE	17	7
Kosovo	XK	15	12
Monaco	MC	11	7
San Marino	SM	2	1

**Table S2.** Uploaded separately - table of raw data used for modelling in the analysis.