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1	Climate warming changes synchrony of plants and pollinators in Germany
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3	Running title: Plant pollinator shifts in Germany
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21 Abstract

22 Climate warming changes the timing of many life-history events across the biosphere. When 23 interacting organisms, such as plants and their pollinators, experience these changes differently, climate change may disrupt their interactions and ultimately affect the structure 24 25 and stability of associated communities and ecosystems. However, so far there is little empirical data on the strength and consistency of such phenological mismatches. We used 26 27 occurrence records from the Global Biodiversity Information Facility (GBIF) to examine trends in the phenologies of 1,274 plants and 88 associated insect pollinators 28 29 (butterflies/moths, flies, bees, and beetles) in Germany since the 1960s. We found strong and 30 consistent phenological advances in plants (on average -4.5 days per decade), but differences 31 in the extent of shifts among pollinator groups (-3.2 and +2.0 days per decade for butterflies and moths, and beetles, respectively). The observed temporal trends in plant and insect 32 33 phenologies were generally associated with interannual temperature variation (plants: -7.6 days, butterflies and moths: -4.4 days, beetles: +1.4 days per +1°C), and thus likely driven by 34 climate change. In plants, phenological advancement did not depend on their level of 35 pollinator dependence. When examining the temporal co-occurrence of 1,797 plant-pollinator 36 37 pairs over the decades from 1980 onwards, the temporal trends in their synchrony again 38 strongly depended on the pollinator group: while the synchrony of plant-butterfly interactions 39 remained unchanged during the last decades, interactions with bees and hoverflies tended to 40 become more synchronized, mainly because the phenology of plants responded more strongly 41 to climate change and plants caught up with these pollinators. Still, if the observed trends continue, then these interactions are expected to become more asynchronous again in the 42 43 future. Our study demonstrates that climate change affects the phenologies of different interacting groups of organisms, and that this also influences their synchrony, with potential 44 45 far-reaching ecological consequences.

Keywords: asynchrony, GBIF, mismatch, phenology, pollination mode

47 Introduction

48 Phenological events are periodically occurring events in the life cycle of organisms. The 49 timing of these events often depends on environmental factors such as temperature or photoperiod, and it is well known that climate change affects some of these and thus changes 50 51 the phenologies of many organisms (Cleland et al., 2007). With such phenology shifts, there 52 is increasing risk of phenological mismatches between interacting organisms, potentially 53 exceeding the natural resilience of ecosystems (Memmott et al., 2007). Climate changeinduced phenological shifts have been documented extensively for individual species 54 (Parmesan, 2007), but we still know much less about how these shifts affect ecological 55 56 interactions. Kharouba et al. (2018) recently reviewed 54 published interaction studies across 57 ecosystems and interaction types and found no clear general trend, with about half of the studied interactions becoming more asynchronous but the other half becoming even more 58 59 synchronized through climate change.

60 Plant-pollinator systems are among the biotic interactions expected to suffer most from a mismatch of phenological events (Scheffers et al., 2016). Several previous studies have 61 observed mismatches (Miller-Rushing et al., 2010; Robbirt et al., 2014), but in others 62 63 pollinators and plants seemed to be able to keep up with each other (Bartomeus et al., 2011). 64 An interesting question in this context is also which of the two partners is advancing faster if 65 there is an increasing mismatch. So far, the evidence here is also mixed. For instance, Gordo & Sanz (2006) found pollinators to advance faster than trees, and Parmesan (2007) that 66 67 butterflies advanced faster than herbaceous plants, but in a study of Kudo and Ida (2013) it was the plants – spring ephemerals – that advanced faster than their bee pollinators. 68 69 Mismatches of plant-pollinator interactions can have negative consequences for both 70 partners. For the pollinators, this can include lower survival rates, a decreased overall fitness and higher parasite loads (Schenk et al., 2018). Moreover, mismatches might also impact 71

pollinator demography, the body sizes (Miller-Rushing et al., 2010) and frequencies of sexes,
and thus population viability (Schenk et al., 2018). On the plant side, desynchronized
pollinator interactions are mainly expected to impact plant fitness and thus long-term
population growth and survival. For instance, Kudo and Ida (2013) found that seed counts
were reduced in early flowering spring ephemerals after desynchronization with their bee
pollinators. However, in another study fly-pollinated plants did not show similar responses
(Kudo et al., 2004).

79 Plants differ in their level of dependence on plant pollinators, and an intriguing question therefore is to what extent phenology responses to climate change are linked to the pollinator 80 dependence of plants. Bond (1995) theorized that wind-pollinated plants might experience 81 82 little negative consequences of climate change as they do not depend on interactions with animals. Conversely, insect-pollinated plants may be subject to strong selection toward 83 84 phenologies that are in synchrony with their pollinators. This hypothesis was later 85 corroborated in an empirical study of Fitter and Fitter (2002). A more recent study on orchids (Molnár et al., 2012) found that pollination mode influenced the degree of plant advances in 86 flowering phenology, indicating that self-pollinating and thus pollinator-independent plants 87 88 were not constrained by pollinator phenology. The main idea of these previous studies is that 89 all else being equal, pollinator-independent plants should exhibit stronger phenological shifts 90 in response to the same climate changes.

Testing hypotheses about plant-pollinator responses to climate change is not trivial. Since changes in phenology take place on the scale of decades (Parmesan, 2006), we need long-term data. A possible source of long-term data on plant phenology are herbarium specimens (Jones & Daehler, 2018; Lang et al., 2019), which can indicate the day of year that a specific species was flowering in a given location and year. Herbarium data provide unique historical depth, but they need to be treated with caution because of the sampling biases

associated with them (Daru et al., 2018; Maldonado et al., 2015). In recent years the 97 98 digitization of herbaria as well as other collections and observation data, including on other 99 taxa such as pollinating insects, e.g., from long-term monitoring networks, is creating an 100 increasing number of public data bases that contain vast amounts of natural history data that 101 cover large spatial and temporal scales (Newbold, 2010). These data bases are increasingly 102 being used for analyses of broad ecological trends and global changes (Chapman, 2005; 103 Maldonado et al., 2015). One of the largest and most important hubs of large-scale and long-104 term ecological data sets is the Global Biodiversity Facility (GBIF), an intergovernmental 105 initiative and public data base that provides access to biodiversity data compiled from various 106 individual sources like academic institutions, government agencies or independent collections 107 (GBIF, 2019).

Another matter is finding a measure for changes in phenology. Primack et al. (2004) 108 109 demonstrated that the average collection date of the herbarium specimens of a plant species in 110 a year can be used as a proxy for peak flowering time in that year. The same approach of 111 using occurrence records in natural history collections or other data bases can in principle be used to estimate the activity times of other groups of organisms such as insects (Kharouba et 112 113 al., 2018 and references therein). For instance, analyses of natural history collections in the 114 UK have demonstrated phenology changes in bees (Robbirt et al., 2014) and butterflies 115 (Brooks et al., 2014). Thus, the peak occurrences of plants and insects in GBIF may be used 116 to estimate activity shifts of different groups, as well as their synchrony. When we use the 117 term 'activity' in this paper, we refer to the period in an organism's life when it can interact with its ecological partner. For plants this is the period of flowering, for insect pollinators the 118 119 period of flight.

We used data from GBIF to study phenological mismatches between plants and
pollinators in Germany, at the level of taxonomic groups as well as individual interactions.

122 We asked the following questions: (i) Are there long-term trends in the phenology of plants

123 and pollinators? (ii) If yes, are phenology trends related to climate change? (iii) How are

124 phenological changes of plants related to their pollinator dependencies? (iv) How does

125 climate change affect the synchrony of plant-pollinator interactions?

126

127 Methods

128 Phenology data

129 We worked with occurrence records of plants and insects available from the GBIF database 130 (GBIF: The Global Biodiversity Information Facility, 2019). For the plants, we restricted 131 ourselves to species covered by the BioFlor database of plant traits (Klotz et al., 2002), 132 because we needed to be able to classify plants by their level of pollinator dependence (see below). For the insects we restricted ourselves to beetles (Coleoptera), flies (Diptera), bees 133 134 (Hymenoptera) as well as butterflies and moths (Lepidoptera), as these groups contain most 135 insect pollinators (Kevan & Baker, 1983). We used the R package rgbif (Chamberlain & Boettiger, 2017) to download all available records of the above taxa from GBIF. Our basic 136 criteria for including records were that they originated from Germany, and that they referred 137 138 to either a living specimen (e.g., a captured insect), a human observation, just an observation 139 (i.e., when the exact type of observation was not clear), or a preserved specimen (e.g., an 140 herbarium record). If names of plant species were not accepted names, we used the R package 141 taxsize (Chamberlain & Szöcs, 2013) to check the names against the GBIF backbone 142 taxonomy and determine the actual accepted names (for a list of the data sets used, see Table **S1**). 143

Prior to the data analyses, we subjected the data to several steps of quality control (for a schematic see **Figure S1**). First, we removed all records from before 1960 as these turned out to be too inconsistent, with few records per year and large gaps between years with records.

We also removed the records from 2020 as the year had not been complete at the time of our 147 148 analysis. Second, we removed all records from the first and last days of years because the 149 high number of records on those days indicated that records without a recorded collecting date had been given these as default dates. Next, we removed all records from "GEO Tag der 150 151 Artenvielfalt", a German bioblitz event where large numbers of records are taken on a specific day of the year. Including these data would have strongly biased the intra-annual distributions 152 153 of our records. Finally, we removed the records from several collections which appeared to have misclassified these as being of German origin, probably through a combination of 154 coordinate rounding and determining countries of origin automatically from these 155 156 coordinates. We identified these sets of records by visually inspecting the geographic 157 distributions of the records of each institution; most of these erroneous data sets were from Luxembourg (for a complete list see Table S2). There were a few records just outside the 158 159 boundaries of Germany that we did not remove from our data set because the country 160 information appeared trustworthy and we suspected errors with the recording of the coordinates. Obviously, the latter steps of our quality control were possible only for 161 georeferenced records, which made up 99.97% of the total amount of records. After these data 162 163 curation steps, we maintained around 11 million plant records and over one million insect 164 records for our data analysis. There were large differences between plants and insects not only 165 in the numbers of records but also in their temporal distribution across the studied period (Figure S2). While plants, but also beetles, had relatively even record numbers across 166 167 decades, the other insect groups, in particular flies and bees, were strongly underrepresented in the earlier decades, and record numbers increased rapidly only in the last 20 years, 168 169 probably due to the advent of platforms like iNaturalist.org and naturgucker.de, which allow logging of species occurrences by citizen naturalists, and which make up most of our insect 170

171 data. Beetles were represented, save for one species from the Orsodacnidae, by the

172 Chrysomelidae family.

173

174 Climate data, pollinator dependence, and individual interactions

Besides the main phenology data from GBIF, we obtained several other data sets required for
our analyses. To test for associations with climate, we used climate data from Deutscher
Wetterdienst (DWD, https://www.dwd.de/), specifically the historical (until 2018) and recent
(2019) monthly station observations data set (DWD Climate Data Center [CDC], 2020a,
2020b) to calculate the Germany-wide average annual temperatures for 1960-2019. The exact
climate data sets used are available at the repository under data availability.

181 To classify plants by their level of pollinator dependence we used plant trait data from BiolFlor (Klotz et al., 2002). A species was assigned as pollinator-dependent when it was 182 183 either known to be self-incompatible and pollinated by an insect, dioecious and pollinated by 184 an insect, or protogynous/protandrous while also being pollinated by an insect. In contrast, species that were pollinated abiotically or through selfing, that exclusively reproduced 185 vegetatively, or were apomicts, were classified as pollinator independent. If none of the above 186 187 applied, we assigned an intermediate pollinator dependence. If part of the information above 188 was missing, no pollinator dependence was determined, and the species was excluded from the analyses involving pollinator dependence. 189

Finally, we obtained data on individual plant-pollinator interactions from a UK database on plant-pollinator interactions (Redhead et al., 2018) hosted by the Centre for Ecology and Hydrology (CEH). This database included all known interactions between plants and flowervisiting bees, butterflies, and hoverflies (but unfortunately neither beetles nor moths) in the UK, a country similar to Germany in terms of climate and species composition. While these

interaction data are unlikely to represent all possible species interactions in Germany, wecould not find similar data for our study area.

197

198 Calculation of plant and insect phenology

199 For our analyses of plant flowering phenology and pollinator activity times, we averaged all records of a plant or insect species in a year to calculate each year's mean day of the year 200 (DOY) of the occurrence of a species. As discussed above, this occurrence measure was used 201 as an estimate of each year's peak flowering or peak activity time of plants and insects, 202 203 respectively. Each annual mean DOY was calculated from at least five records of a species 204 per year. To avoid extreme shifts based on too little data, we included only species with records in at least 40% of the years. The median number of records per year for a species in 205 206 our analyses was 47.

Since our analyses of individual plant-insect interactions (see below) were done at the level of decades, we additionally calculated the decadal means, based on nominal decades (0to-9), of species DOYs for each of the included species, and only when at least five records existed per decade. These decadal interaction analyses were done only from 1980 onwards, i.e., for four decades, as too few data were available prior to 1980. To be included in our analyses, an interaction's records needed to span the entire period examined.

After clean-up and averaging, a total of 58,895 annual and 1,336 decadal peak DOYs, with the latter based on a median number of 1686 records, remained in our data set (**Figure S1**). The annual activity data included 1,274 plant and 88 insect species. For 948 of the plant species we had information about pollinator dependence: 144 were pollinator-dependent, 204 pollinator-independent, and 600 were classified as intermediate. The 88 insect species consisted of 40 species of beetles, 44 butterflies and moths, three bees and just one fly

species. The decadal data included 245 plant and 26 insect pollinator species. All data
wrangling and analysis was done in R (R Core Team, 2008).

221

222 Data Analysis

223 To understand phenology changes in plants versus insects, we first estimated the average phenological shifts in each group. We defined phenological shifts as the slope of the linear 224 225 regression linking the peak activity (= mean annual) DOY of an individual species to the year of observation. We visually confirmed approximate normal distribution of the individual-226 species slopes, and that no improbable outliers were present. There were some plants with 227 rather extreme values (Figure S3), however these were mostly early flowering plants which 228 229 likely experience stronger pressures and therefore stronger phenology shifts (Forrest, 2015), and we therefore did not exclude them from our analyses. We compared the mean 230 231 phenological shifts between plants and insects using an independent-sample Welch's t-test, 232 and we further examined the temporal trends between different insect orders and plants in an ANOVA, using a Tukey post-hoc test to determine pairwise differences. We excluded bees 233 234 and flies from the last step as their numbers were too small to be representative for their 235 respective groups.

236 Different climatic factors likely affect the timing of early and late activity periods, 237 which might complicate the interpretation of the peak shifts. We therefore also assessed the 238 extent of shifts of first and last day of activity for each species (and consequently the duration 239 of their activity) to understand how asymmetries in the shifts might affect the peak shifts in phenology. For this we estimated the shifts of the decadal average first and last activity day of 240 241 the year over time in a linear model. We also estimated the shifts of duration of the activity period by first calculating the yearly duration of the activity period as the difference between 242 the last recorded day of activity and the first for each species, taking the decadal average of 243

said duration and then estimating the shift over time in a linear model. We used decadal
averages to ensure the differences were due to long-term trends, as the absolute first and last
day of activity is just the first and last record of a species in that year and therefore subject to
fluctuation.

In addition to the temporal trends in phenology, we also tested for the climate sensitivity of plant and insect phenology. These analyses were analogous to the ones above, except that the explanatory variable was annual mean temperature instead of year of observation, i.e., the data were regression slope parameters of mean annual DOY of a species over the average temperature in that year.

Next, we tested whether phenology trends differed between plant groups with different levels of pollinator dependence. For this, we used the same data as above (slope parameters of individual-species regressions), but we analyzed it with a linear model that included pollinator dependence (dependent, independent, or intermediate) as a fixed factor, and then determined pairwise differences between groups with a Tukey post-hoc test. In addition, we also tested whether mean activity DOY differed significantly between the three pollinator-dependence levels.

260 Finally, we analyzed asynchrony between plants and pollinators using the data on 261 individual plant-pollinator interactions. For each plant and presumed insect pollinator, we 262 calculated the absolute difference in peak activity times for each decade. A value of zero thus indicated perfect asynchrony, and higher values indicated increasing asynchrony. To test 263 264 whether asynchrony changed over time we estimated the slopes of the relationship between differences in peak activities and time (decades) for each plant-insect interaction with a linear 265 266 model. Here, negative slope values indicated a shift towards greater synchrony, and a positive slope a shift towards greater asynchrony. Altogether, there were 1,797 interactions involving 267 245 plants and 26 insect pollinators, one insect usually associated with multiple plants but 268

seldomly plants with multiple insects. To test for differences in average asynchrony and
change of asynchrony between insect groups, we used an ANOVA and assessed pairwise
differences with a Tukey post-hoc test.

272

273 **Results**

274 Temporal trends in plant and insect phenology

275 The analysis of the peak activity data showed a strong difference in the average temporal 276 shifts of plant and insect phenology (Welch's $t_{100,929} = 6.644$, P < 0.001). The phenology of plants generally advanced much more strongly, with an average shift of -4.5 ± 0.2 days per 277 decade (mean \pm SE), while across all insects the shift was only -0.4 \pm 0.6 days per decade. 278 279 84.8% of all plant species but only 56.8% of all insect species advanced their phenology (Figure 1). However, these numbers across all insects obscured different trends among the 280 281 insect orders: when considered separately, butterflies/moths exhibited a strong phenology 282 shift of -3.2 ± 0.8 days per decade (mean \pm SE), with 79.5% of the species advancing, whereas the beetles in contrast delayed their peak activity on average by 2.0 ± 0.7 days per 283 decade, with 65.0% of the species following this trend. When plants, butterflies/moths and 284 285 beetles were analyzed as separate groups, ANOVA indicated significant differences among 286 them ($F_{2,1355} = 25.16$, P < 0.001), with significant pairwise differences (Tukey post-hoc, $\alpha =$ 287 0.05) between the phenology shifts of beetles and plants, and beetles and butterflies/moths, respectively. (For an overview over all groups, see Figure S4A and Figure S5.) 288 289 We found asymmetries between the slopes of first and last day of activity over time (Figure S6). In plants, the symmetry was generally skewed towards a stronger shift of the 290 291 first day of activity (First: -1.2 ± 0.0 mean days/decade \pm SE, Last: 0.5 ± 0.0 mean days/decade \pm SE) with butterflies/moths behaving similarly (First: -1.4 \pm 0.1 mean 292 days/decade \pm SE, Last: 0.5 \pm 0.2 mean days/decade \pm SE), whereas in beetles the last day of 293

activity shifted more strongly (First: 0.1 ± 0.1 mean days/decade \pm SE, Last: 0.5 ± 0.1 mean days/decade \pm SE). It is also notable that the plants' and butterflies/moths' day of first activity generally advanced while the beetles' day of first activity was rather delayed (**Figure S7**).

297

298 Climate sensitivity of plant and insect phenology

299 The climate sensitivities of the phenologies of plants, butterflies/moths and beetles generally 300 resembled their temporal trends (Figure 1), and group differences in climate sensitivities 301 matched those in temporal trends described above. Again, there was a significant difference 302 between plants and all insects (Welch's $t_{96.026} = 8.027$, P < 0.001), with plants showing a 303 strong negative association between peak activity and temperature, but a much weaker 304 association for all insects together. On average, plant peak flowering shifted by -7.6 ± 0.2 days per $^{\circ}$ C (mean \pm SE), and 92.5% of the individual species showed earlier flowering with 305 306 increasing temperature, whereas for insects it was only -1.3 ± 0.8 days per °C, and 63.6% 307 showing a trend towards earlier peak activity (Figure 1). When the butterflies/moths were 308 considered separately, however, they showed a strong association with temperature, with an average peak activity shift of -4.4 ± 0.8 days per °C (mean \pm SE) and 80% of the individual 309 310 species advancing, whereas the beetles showed an opposing trend of delayed peak activity, 311 with an average of $+1.4 \pm 1.1$ days per °C temperature change. There were significant differences among the three groups (ANOVA, $F_{2,1355} = 45.701$, P < 0.001), with significant 312 313 differences between all pairwise combinations (Tukey post-hoc, $\alpha = 0.05$). (For an overview 314 over all groups, see Figure S4B and Figure S8.)

315

316 *Pollinator dependence*

The phenology of plants, and its temporal trends, differed very little among plant groups of

318 different levels of pollinator dependence (Figure S9). The peak flowering of pollinator-

independent plants (average DOY 199.5) advanced on average by -3.9 days per decade, while
pollinator-dependent plants (average DOY 196.2) advanced by -5.1 days per decade, and
intermediate plants (average DOY 199.5) advanced by -4.5 days per decade. In all three
groups, the percentage of plants advancing was 85-86%. None of the differences between
groups was statistically significant.

324

325 Synchrony of plant-pollinator interactions

When examining the synchrony of individual plant-pollinator interactions, we found that the 326 three pollinator groups differed in their average levels of asynchrony with the plants, but that 327 328 interactions did not become more asynchronous but rather more synchronized during the last 329 decades (Figure 2A). The temporal trends differed strongly among the pollinator groups (ANOVA, $F_{2,2522} = 67.750$, P < 0.001; Tukey's post-hoc test significant at $\alpha = 0.05$ for all 330 331 pairwise comparisons: Figure 2C): while the synchrony of plant-butterfly interactions 332 remained on average unchanged, plant-pollinator interactions involving bees shifted on average by -2.7 days per decade, with 68% of individual interactions decreasing asynchrony 333 over time. The strongest shifts were in plant-hoverfly interactions which shifted by -6.2 days 334 335 per decade, with 89% of all interactions showing decreasing asynchrony (Figure 2A, C). In 336 all three plant-pollinator groups, asynchrony was mostly due to earlier peak activity of the 337 insects (Figure 2B). Interestingly, however, there was a tendency for these patterns to disappear in all three groups over time, presumably because of the stronger phenology shifts 338 339 of plants (**Figure 1**). Plant-hoverfly interactions ($n_{Insect} = 1$, $n_{Plant} = 132$, $n_{total} = 132$) became on average synchronous in the last decade. For the plant-butterfly interactions ($n_{Insect} = 36$, 340 341 $n_{Plant} = 231$, $n_{total} = 1,819$) the linear model predicts the point of synchrony to be reached in 2029, and for the plant-bee interactions ($n_{Insect} = 4$, $n_{Plant} = 214$, $n_{total} = 574$) in 2050. 342

343

344 **Discussion**

345 In this study, we took advantage of large collections of occurrence records to examine 346 phenological trends of flowering plants and insect pollinators in Germany. We asked whether 347 phenology changes affected the synchrony of plants and insects, and whether observed 348 changes in phenology, and variation therein, were related to the different groups' responses to 349 climate warming. We also examined whether the phenology responses of plants depended on 350 their levels of pollinator dependence. Our results showed that the phenological shifts of plants 351 and insects indeed differed, with plants shifting by several days per decade while insects on average shifting hardly at all. As peak flowering historically occurred after peak insect 352 353 activity, these trends imply an increase in plant-pollinator synchrony during the last decades, 354 but a potential for future desynchronization if climate change continues.

Plants and insects also differed in their overall temperature sensitivity. While plants 355 356 shifted on average by over a week per degree of warming, insects shifted by only one day. 357 There were large differences between insect orders in their phenology trends and temperature 358 sensitivities. As groups with greater temperature sensitivity also showed larger phenology shifts over time, it seems likely that the two are causally related, i.e., that anthropogenic 359 360 climate warming is responsible for the observed phenology shifts. Lastly, there were no 361 differences between pollinator-dependent and -independent plants, suggesting that plants 362 either responded passively to temperature, with advanced flowering in warmer years irrespective of pollinator dependence, or that most plants have sufficient generalist pollinators 363 364 that can fill in for other, desynchronized pollinator species and thereby reduce selection pressure on plant phenology. 365

366

367 Caveats

When interpreting the results of our study, it is important to consider some caveats of the 368 collections data and occurrence records we used. For instance, the temporal distribution of 369 370 collections data is usually quite heterogenous, and so was our data (Figure S2). Our analysis of the shifts of the first and last days of activity may thus be influenced by varying 371 372 observation efforts over the years. In particular the increasing popularity of nature observation 373 platforms such as www.naturgucker.de, whose records are contained in GBIF, may have resulted in higher probability of detecting early and late occurrences. Besides temporal 374 375 heterogeneity, occurrence records are usually also not homogenously represented in space. 376 Our study's measure of phenology, peak occurrence time, does not account for temporal variation of spatial representation of records within Germany, although some areas might be 377 378 over- or underrepresented in some parts of the studied period. Moreover, our study also does 379 not account for spatiotemporal variation in macro- and microclimate which can influence 380 intraspecific variation in phenology shifts (Song et al., 2020) and could therefore potentially 381 induce local mismatches.

382 When estimating insect peak activity, we did not account for the earlier life stages of 383 insects appearing in the data, despite being not important for pollination. This bias could be 384 most relevant for butterflies and moths, as their larval stages are more conspicuous than fly and beetle larvae. Butterflies/moths are, however, the group with the latest peak activity times 385 386 for large parts of the studied period, so this bias is either not strong or we are underestimating how late in the year butterflies and moths occur. Similarly, some plants occurrences may have 387 388 been recorded when plants were not flowering. Flowers are important for plant species identification, and herbarium records are usually made from flowering specimen, but we 389 390 cannot rule out that some plant occurrence records were based on vegetative plants alone.

Finally, in our analyses we focused on peak activity and therefore did not consider the degree of overlap between the flight times of pollinators and the flowering of plants. However, if the durations of activity periods change, then the relative overlap of two interacting groups could change in spite of identical activity peaks, or vice versa. Testing such possibilities with occurrence data, however, requires even higher-resolution data for individual species than in our study.

397

398 Phenological shifts over time

The general differences between plants and insects in their advancement of phenology seem to indicate a shift in the synchrony between plants and their pollinators, with plants generally advancing faster than insects. However, the insect groups differed strongly in the extent of their shifts of activity over time, and the overall pattern of a slower phenological shifts was largely driven by the beetles, whereas butterflies/moths kept pace with the phenology changes of plants.

405 The extent to which plants advanced their phenology in our data is comparable to that 406 found by Fitter and Fitter (2002) in their long-term observation study of changes in first 407 flowering dates of hundreds of plant species in England. They compared flowering during 408 1991-2000 to that between 1954 and 1990 and found an average advancement of 4.5 days. 409 This is surprisingly congruent with our observation of 4.5 days advancement per decade over 410 the whole period from 1960 to 2019. A more recent long-term analysis of phenology changes 411 in subalpine meadow plants in the Rocky Mountains was undertaken by CaraDonna et al. (2014) who found an even stronger average advancement of first flowering of 6.4 days per 412 413 decade. Since CaraDonna et al. (2014) also analyzed peak floral abundance, their data should be particularly comparable to our estimation of peak flowering through the DOY of peak 414 occurrence. They found a rate of advancement of 5.3 days per decade in spring peak 415

416 abundance but only 3.3 days for the summer peak floral abundance. Our results of peak

417 occurrence across the whole year thus fall in between these two estimates.

418 For insects, previous studies seem to be less consistent, with widespread but not universal advances in springtime phenology (mostly associated with warming) over the last 419 420 decades (Forrest, 2016). For butterflies, long-term records showed that their times of first flights (correlated with peak appearance) advanced on average by -3.7 days per decade in the 421 422 2000s compared to the previous decades in England (Roy & Sparks, 2000), and by -7.7 days per decade in California (Forister & Shapiro, 2003). The magnitude of the shifts observed in 423 England is similar to what we estimated for butterflies/moths in Germany (on average -3.2 424 425 days per decade).

426

427 Temperature sensitivities of plant and insects

428 We found that associations between temperature and phenology differed among groups but 429 that the magnitude of these associations generally reflected the different groups' phenology 430 shifts observed over time. This strongly suggests a link between the phenology shifts and 431 climate change, corroborating previous studies such as the ones by CaraDonna et al. (2014) 432 and Song et al. (2020). We found that plants were generally more sensitive to temperature, 433 i.e., their phenology advanced more strongly, than insect pollinators. Previous studies on 434 insect phenology in the temperate zone (reviewed in Forrest, 2016) have shown that increased 435 spring temperatures are often associated with earlier insect emergence, but that this pattern 436 cannot be generalized as easily as for the plants, as temperature-phenology relationships of insects are more complex. While many insects plastically respond to warmer temperatures by 437 438 speeding up their rates of development (and thus potentially emerge earlier), others have been found to respond in counterintuitive ways and delay their phenology. This might be due to 439 440 dependence on other cues such as rainfall (Bonal et al., 2015), due to cold period

requirements of insects during their diapause (climate warming can cause a loss or reduction 441 442 of this chilling period, and this tends to increase the amount of warming required for 443 subsequent emergence; Forrest, 2016), or because species overwinter in a diapause state in which they are not temperature sensitive (Fründ et al., 2013). Fründ et al. (2013) also showed 444 445 that bees overwintering in larval stages responded to higher winter temperatures with delayed emergence, while bees overwintering as adults showed advanced emergence (but had greater 446 weight losses during overwintering). We did see delayed phenology in some of our data, 447 particularly for beetles and bees. This also connects well to some of the findings reviewed by 448 449 Forrest (2015), for instance that during winter above-ground nesting bees experience different 450 temperatures than the plants they feed on during the summer. Such microclimate differences 451 between insects and plants during overwintering may sometimes explain contrasting climate responses. In other cases, delays in the first appearance of adults may result from longer 452 453 growing seasons. For example, longer growing seasons have reduced selection for rapid 454 development in some high-elevation grasshoppers, in such a way that they reach maturity 455 later — but at a larger size — than in the past (Buckley et al., 2015). Furthermore, warming 456 can change the number of generations per year (voltinism; Forrest, 2016). All of the above-457 mentioned mechanisms can cause variation in the phenology shifts of insects with climate 458 warming and may therefore explain why climate change is not always accompanied by 459 phenological advances but might also cause delays – as we observed for the beetles.

Another interesting idea is that the phenological advancement of the plants itself could cause delayed phenology of some pollinators. Wallisdevries and van Swaay (2006) found that advanced plant growth led to delayed development of butterflies since the cooling created by shading leaves worsened foraging conditions for the larvae. However, in our study we did not see this effect for butterflies/moths as their phenology shifts closely tracked the shifts of

plants, perhaps because of the high levels of specialization of many butterfly larvae (Gilbert& Singer, 1975).

467

468 Pollinator-dependence of plants

469 We did not find any differences in the phenological changes of pollinator-dependent versus 470 pollinator-independent plants. This result is consistent with Rafferty and Ives (2011) who 471 found that the phenology shifts of plants were not constrained by their pollinators, because 472 these kept pace with the plants. In contrast, Kudo et al. (2004) found a negative effect of 473 flowering advancement in bee-pollinated but not fly-pollinated plants. Fitter and Fitter (2002) 474 found significant differences between insect-pollinated plants (-4.8 days shift in day of first 475 flowering) versus wind-pollinated plants (-3.5 days shift) and suggested this was because shifting pollinator activity forced plants to flower earlier. In our study we did not find any 476 477 such differences, indicating that plant responses to temperature are either entirely passive, or 478 that most plants have generalist pollinators with a long period of activity, so that there is little 479 selection pressure on plant phenology. The data set used in our analysis is larger than those used in the studies cited above, so our results may be regarded as more conclusive and more 480 481 general, bearing the limitations of the collections data in mind.

482

483 *Changes in plant-pollinator synchrony*

When we analyzed the synchrony of plant-pollinator interactions, we found clear trends in shifting synchrony, but they strongly varied among insect pollinator groups. Since the phenology of plants generally advanced faster than that of the insects during the last decades, but plants had generally been the later partner in most plant-pollinator interactions, these shifts lead to greater synchrony overall. However, if the observed trends continue, then many of the studied interactions will soon reach points of perfect synchrony, and after that the

interactions may become more asynchronous again, albeit in the other direction. For plant-490 491 hoverfly interactions this point has already been reached. With linear trends and if we assume 492 that observed trends will continue, the points of reversals are expected in approximately 10 years for plant-butterfly interactions and in around 30 years for plant-bee interactions. If 493 494 interactions will become more asynchronous again in the future, then resilience of pollinator networks, in particular through pollinator generalism, could buffer some of the impact of 495 phenological mismatches (Miller-Rushing et al., 2010), and our finding of no differences 496 497 between pollinator-dependent and pollinator-independent plants support this idea. However, 498 while generalist pollinators make up the larger part of the interactions in most pollination 499 networks, some plant-pollinator interactions are highly specialized, and these might be the 500 ones suffering most from future mismatches (Bascompte & Jordano, 2007). 501

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505

506 Authors contributions

507 JF and FMW conceived the study; JF collected and analyzed the data, and wrote the first draft 508 of the manuscript, with guidance from FMW. JFS and OB provided input to data analysis and 509 manuscript writing. All authors read and approved the final manuscript.

510

511 Data availability

512 The climate data that support the findings of this study are openly available in the DWD Climate Data513 Center at

514 https://opendata.dwd.de/climate_environment/CDC/observations_germany/climate/monthly/kl/historic
515 al/.

- 516 The occurrence data are openly available from GBIF at the following URLs:
- 517 https://doi.org/10.15468/dl.uu63tf, https://doi.org/10.15468/DL.HY4ETB,
- 518 https://doi.org/10.15468/DL.AB8B4Q, https://doi.org/10.15468/DL.BDEN5T,
- 519 https://doi.org/10.15468/DL.EBANT3, https://doi.org/10.15468/DL.3CVYTS,
- 520 https://doi.org/10.15468/DL.TRZY5J, https://doi.org/10.15468/DL.UQBXG2,
- 521 https://doi.org/10.15468/DL.KSQEVW, https://doi.org/10.15468/DL.RHAR75,
- 522 https://doi.org/10.15468/DL.7U4BE8.
- 523 The plant-pollinator interaction data is available at the UK Centre for Ecology and Hydrology at
- 524 https://doi.org/10.5285/6d8d5cb5-bd54-4da7-903a-15bd4bbd531b.
- 525 The plant trait data used in this study is available at https://www.ufz.de/biolflor/index.jsp.
- 526 The code required to run the analysis is available at https://github.com/jonasfreimuth/Phenological-
- 527 shifts-germany.
- 528
- 529 **Conflicts of interest**
- 530 The authors declare no conflict of interest.

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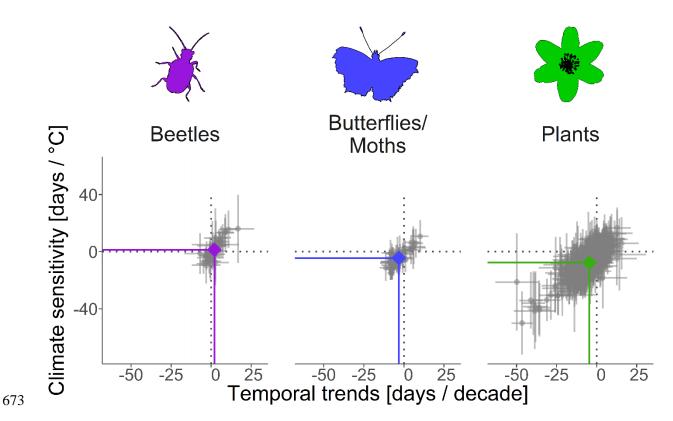
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674Figure 1. Temporal trends (days per decade) versus climate sensitivities (days per °C temperature675change) of the phenology (peak flowering/activity) of plants, beetles, and butterflies/moths, with the676colored lines indicating the averages for each group. Grey dots indicate individual species means with677the vertical and horizontal bars representing the 95% confidence intervals. For all three groups the678relationship between temporal trend and climate sensitivity is highly significant at with r > 0.8 and a P679< 0.001.</td>

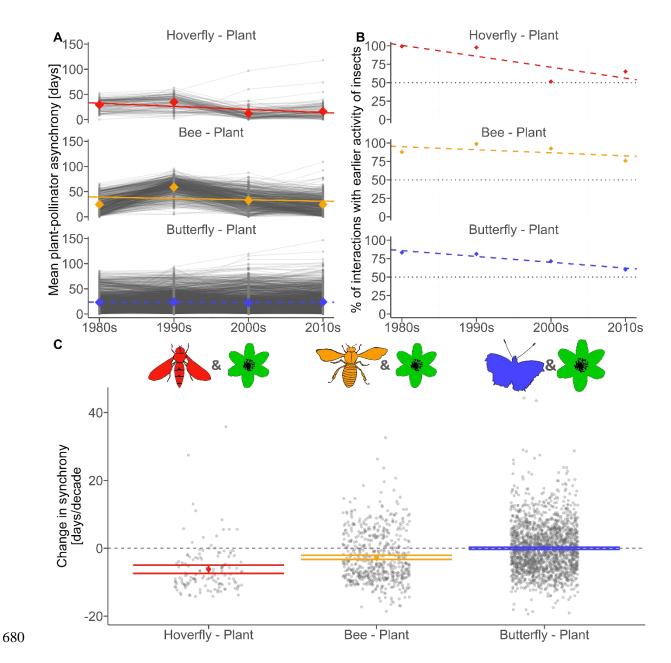
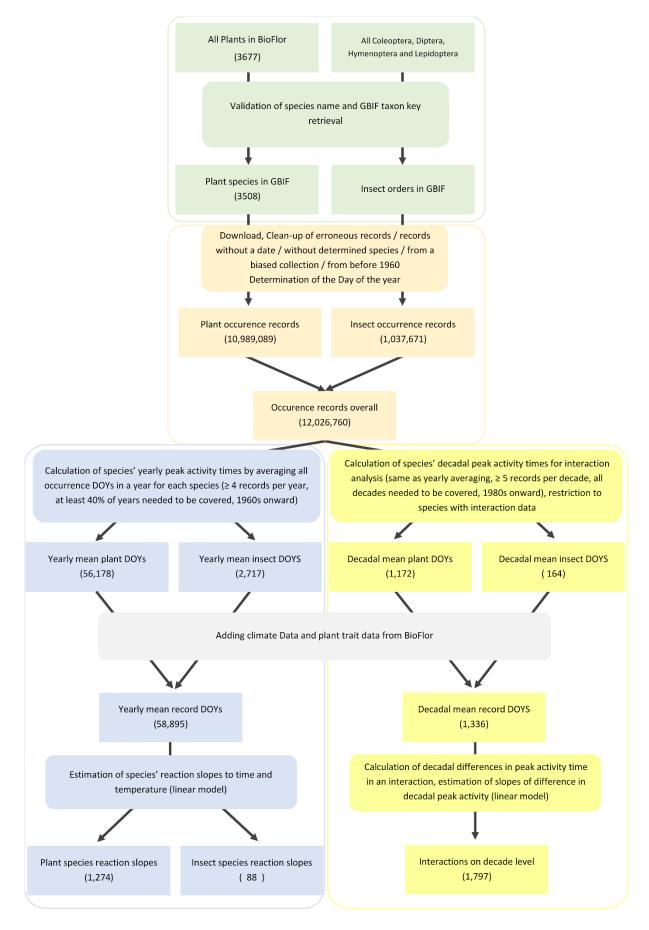


Figure 2. Asynchrony of individual plant-pollinator interactions, and their temporal trends,
separated by pollinator groups. (A) Decadal changes of asynchrony (grey dots/lines:
individual interactions; colored diamonds/lines: linear regression for each group. (B) Fraction
of interactions with earlier insect activity. (C) Average decadal synchrony changes of
individual interactions (grey dots), and the means for each group (colored dots and 95% CI
whiskers). Solid lines in (A) and (B) indicate significant linear regressions, dashed lines nonsignificant ones.

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689

690 Figure S1. Schematic of data collection and the different steps of quality control, data

- 691 selection and data aggregation. First, we selected and aggregated the plant and insect data
- 692 (green), then these data were cleaned (orange), and after that we created a data set on activity
- 693 shifts of individual species using the yearly data (blue) and another data set on species
- 694 interactions using decadal data (yellow).

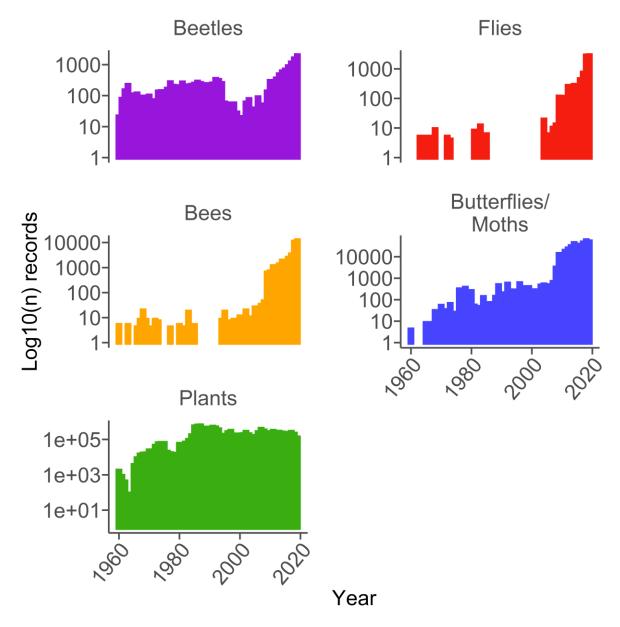
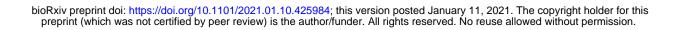


Figure S2. The numbers of occurrence records per year for each of the studied taxonomic

697 groups. Note the different scales of the y-axes, and their log-transformation.



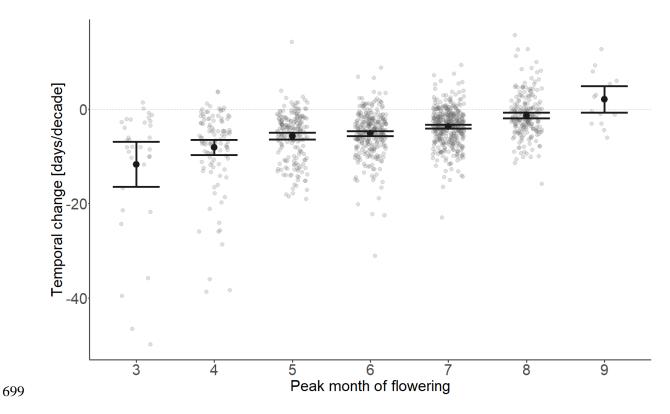
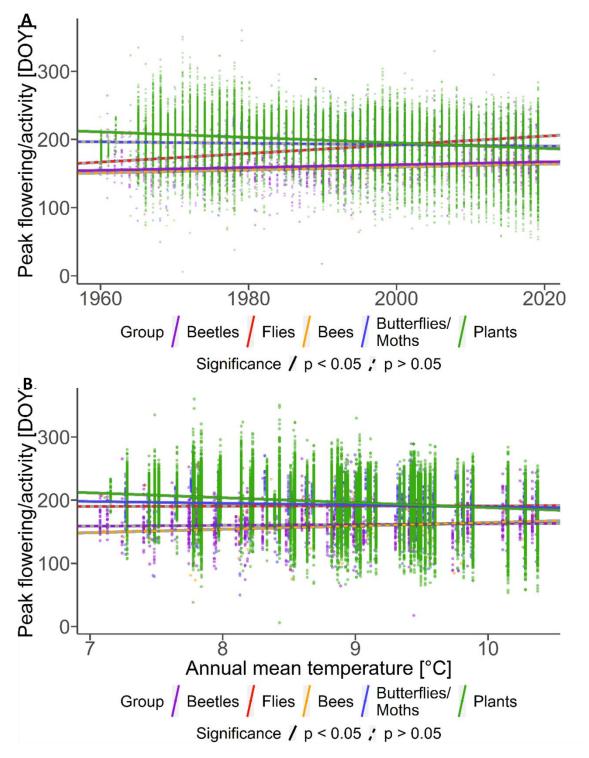
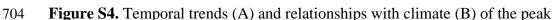


Figure S3. Temporal changes of plant phenology, with species grouped by their peak month of
flowering. Grey dots are regression slopes (= days/decade) of individual species; black dots with
whiskers are group means with 95% confidence intervals.





flowering/activities of plants (green) and insects (purple: beetles; red: flies; orange: bees;
blue: butterflies/moths), with each dot representing the average peak activity of an individual

- 707 species in a specific year. Solid lines represent significant linear regressions, dashed lines
- 708 non-significant trends for taxonomic groups.

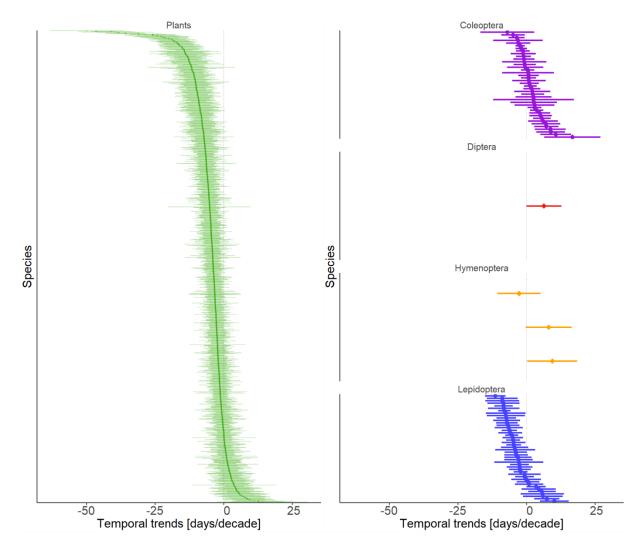


Figure S5. Forest plot of plant and insect species' temporal trends ordered by the strength of

the trend. Lines represent 95% confidence intervals.

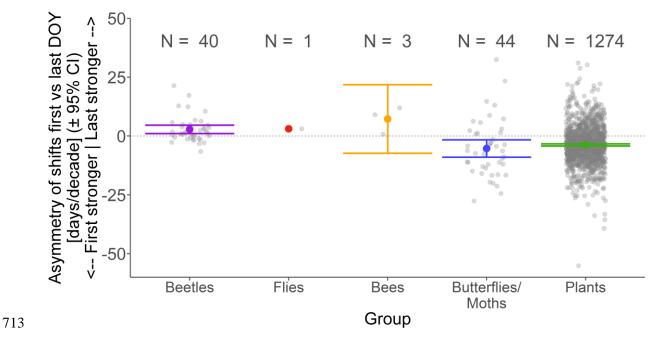
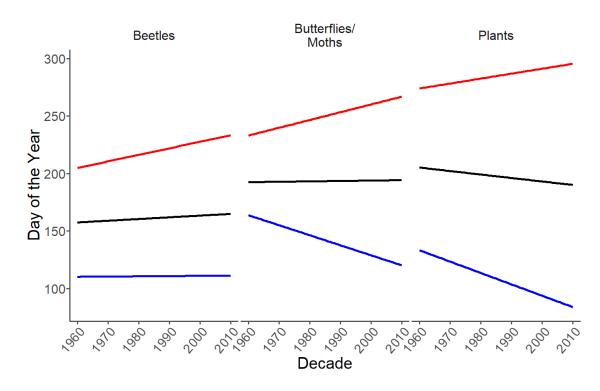


Figure S6. Asymmetry in slope of the first and last day of occurrence over time by species
group (colored dots with 95% confidence intervals) and by individual species (grey dots).
Asymmetry is the difference between the slope of the decadal average first and last
occurrence of a species in a year over time, with values close to zero representing low
asymmetry.

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720 Figure S7. Mean shifts in first (blue) and last (red) occurrences of beetles, butterflies/moths,

- and plants over the studied decades. Black lines represent trends of the peak
- 722 flowering/activity times.

723

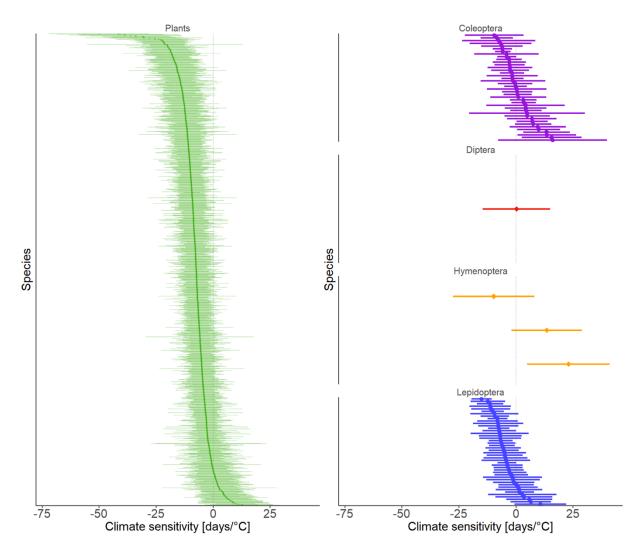
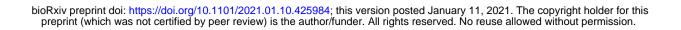
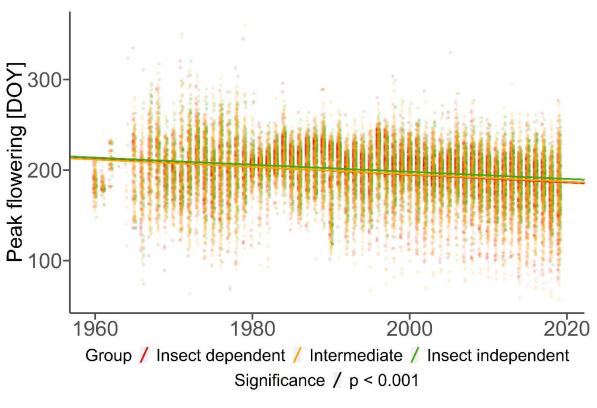


Figure S8. Forest plot of plant and insect species' sensitivity to climate ordered by the

strength of the sensitivity. Lines represent 95% confidence intervals.





730Figure S9. Temporal trends in plant phenology, separated by levels of pollinator dependence of the731plants. Red dots: pollinator-dependent plants; green dots: pollinator-independent plants; yellow dots:732intermediate levels of pollinator dependence. The lines are linear regressions; all are significant at P <7330.001.

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Table S1. GBIF datasets used in our study.

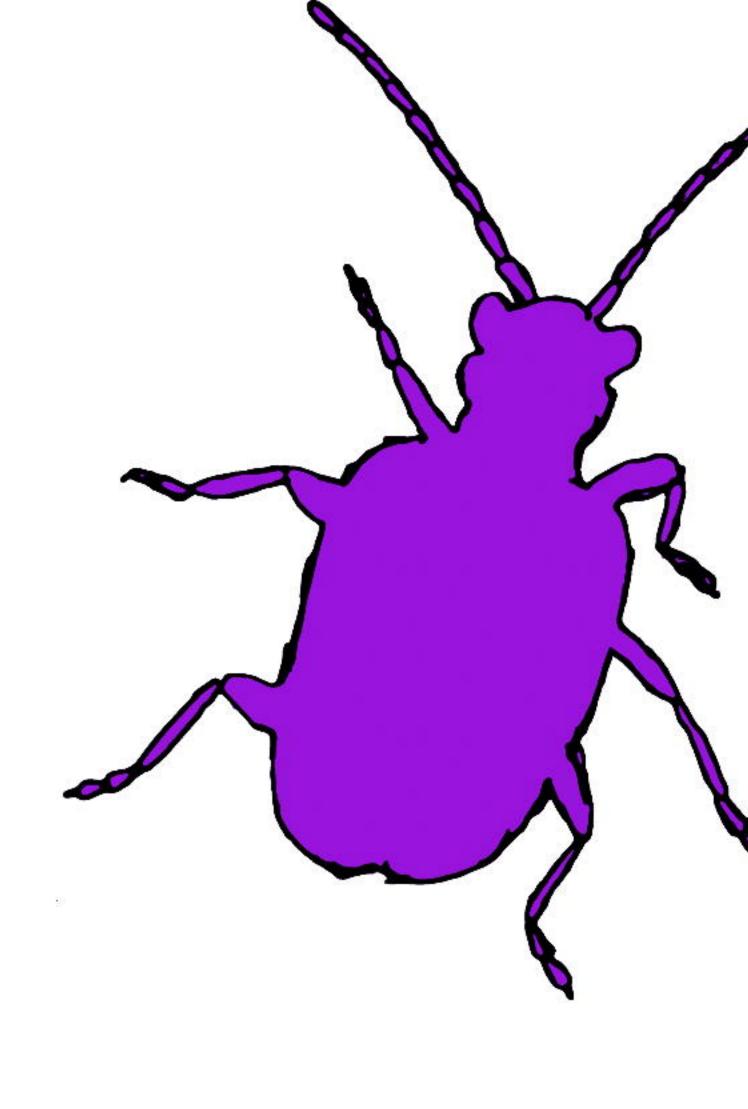
Group	Citation	DOI
Insects	GBIF.org (2020f)	10.15468/DL.UU63TF
Plants	GBIF.org (2020h)	10.15468/DL.HY4ETB
	GBIF.org (2020e)	10.15468/DL.AB8B4Q
	GBIF.org (2020d)	10.15468/DL.BDEN5T
	GBIF.org (2020i)	10.15468/DL.EBANT3
	GBIF.org (2020g)	10.15468/DL.3CVYTS
	GBIF.org (2020c)	10.15468/DL.TRZY5J
	GBIF.org (2020b)	10.15468/DL.UQBXG2
	GBIF.org (2020a)	10.15468/DL.KSQEVW
	GBIF.org (2020j)	10.15468/DL.RHAR75
	GBIF.org (2020k)	10.15468/DL.7U4BE8

Table S2. Institutions whose records we excluded from our analyses. For "GEO Tage der Artenvielfalt" the reason was that there was an extreme

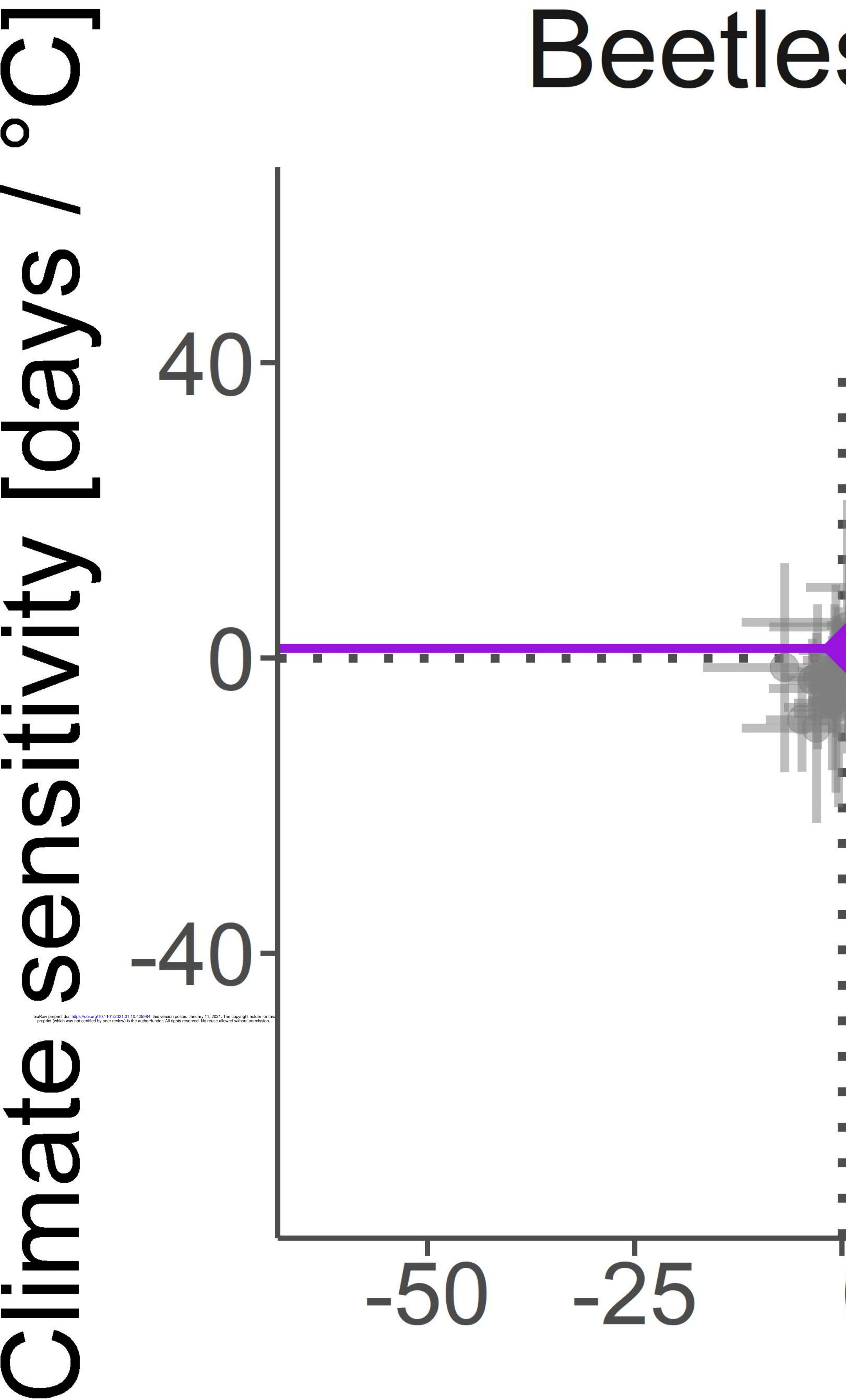
number of records on a single day; for all other institutions the reason was that many records listed as being of German origin were probably

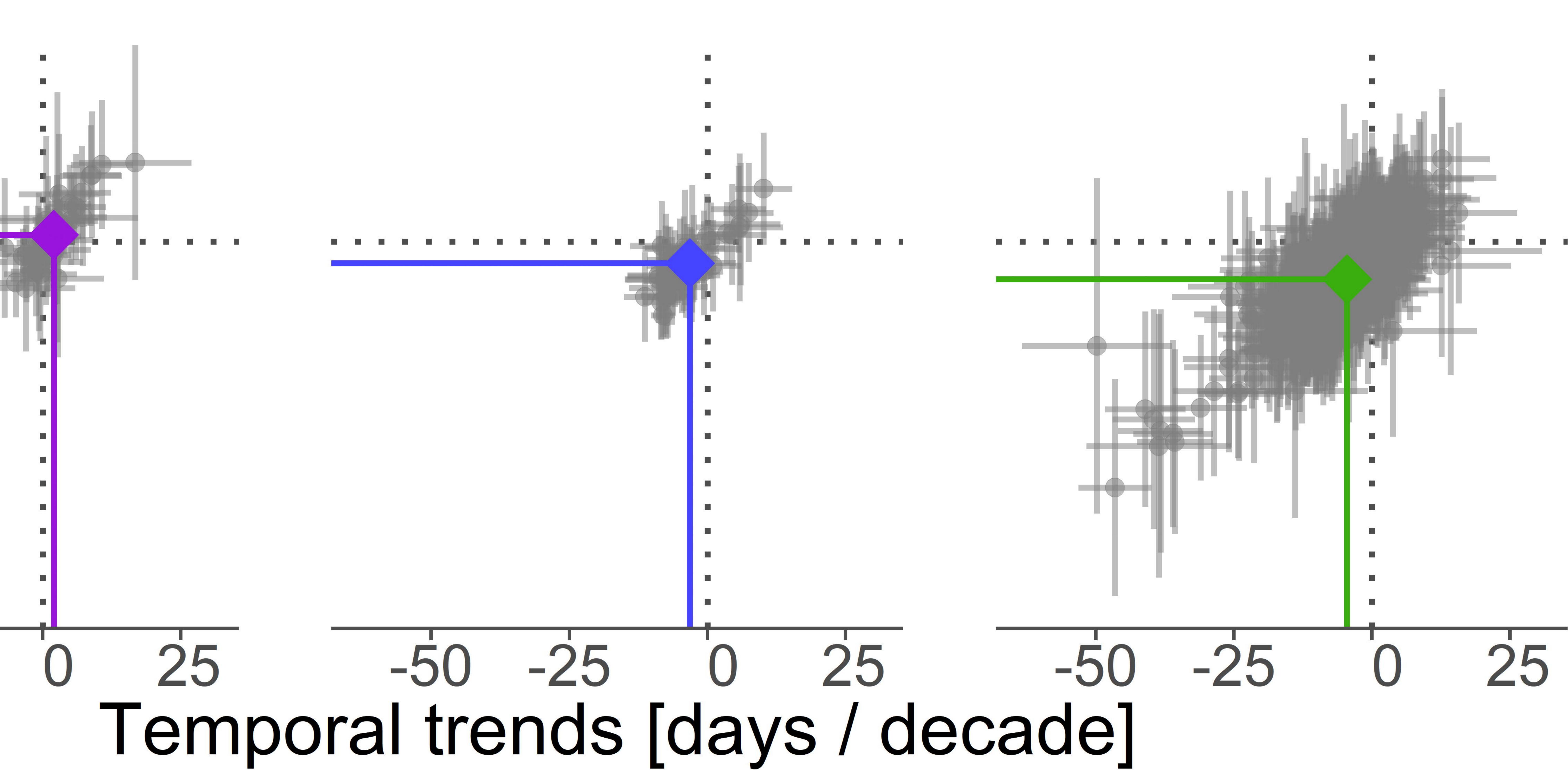
738	misclassified and actually not from Germany.					
	Institution	Description				

Institution	Description
Administration de la Gestion de l'Eau (AGE)	Water administration of Luxembourg
Administration de la nature et des forêts (ANF)	Nature and forest administration of Luxembourg
GEO Tag der Artenvielfalt	Annual Biodiversity Day of the GEO magazine in Germany
Ministère de l'Environnement, du Climat et du	Ministry of Environment, Climate and Sustainable Development of
Développement durable (MECDD)	Luxembourg
Musée national d'histoire naturelle du Luxembourg (MnhnL)	National Museum of Natural History, Luxembourg
Naturpark Öewersauer	Upper Sûre Natural Park, Luxembourg
SICONA - Naturschutzsyndikat	Communal organization for nature conservation in Luxembourg
SPW-DEMNA	Department of Natural and Agricultural Environment, Wallonia, Belgium
STOWA	Foundation for Applied Water Research, Netherlands



Beetles

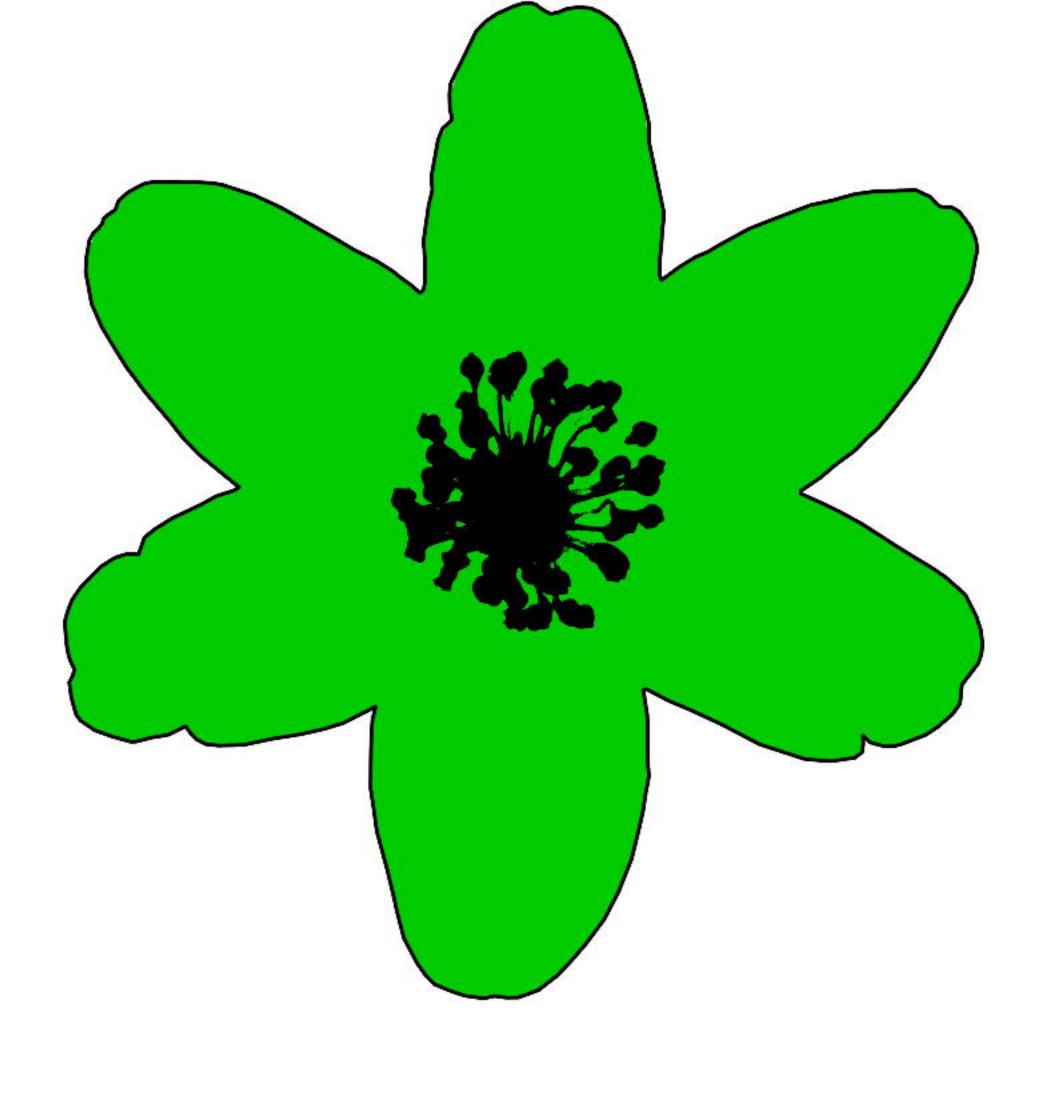








Butterflies/ Moths



Plants

