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1	Climate warming changes synchrony of plants and pollinators
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3	Running head: Plant pollinator shifts
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5	Jonas Freimuth ¹ , Oliver Bossdorf ¹ , J. F. Scheepens ² , Franziska M. Willems ¹
6	
7	¹ Plant Evolutionary Ecology, Institute of Evolution and Ecology, University of Tübingen, Auf
8	der Morgenstelle 5, 72076 Tübingen, Germany
9	² Plant Evolutionary Ecology, Institute of Ecology, Evolution and Diversity, Goethe
10	University Frankfurt, Max-von-Laue-Str. 13, 60438 Frankfurt am Main, Germany
11	
12	Correspondence: Franziska Willems, email: franziska.willems@uni-tuebingen.de, phone: +49
13	7071 2972669
14	
15	OrcIDs:
16	Jonas Freimuth: https://orcid.org/0000-0002-9259-3696
17	Oliver Bossdorf: https://orcid.org/0000-0001-7504-6511
18	J.F. Scheepens: https://orcid.org/0000-0003-1650-2008
19	Franziska Willems: https://orcid.org/0000-0002-5481-3686

21 Abstract

22	Climate warming changes the phenology of many species. When interacting organisms
23	respond differently, climate change may disrupt their interactions and affect the stability of
24	ecosystems. Here, we used GBIF occurrence records to examine phenology trends in plants
25	and their associated insect pollinators in Germany since the 1960s. We found strong
26	phenological advances in plants, but differences in the extent of shifts among pollinator
27	groups. The temporal trends in plant and insect phenologies were generally associated with
28	interannual temperature variation, and thus likely driven by climate change. The phenological
29	advancement of plants did not depend on their level of pollinator dependence. When
30	examining the temporal co-occurrence of plant-pollinator pairs from 1980 onwards, the
31	temporal trends in their synchrony again depended on the pollinator group: while the
32	synchrony of plant-butterfly interactions remained unchanged, interactions with bees and
33	hoverflies tended to become more synchronized, mainly because the phenology of plants
34	responded more strongly to climate change and plants caught up with these pollinators. If the
35	observed trends continue, these interactions are expected to become more asynchronous again
36	in the future. Our study demonstrates that climate change affects the phenologies of
37	interacting groups of organisms, and that it also influences their synchrony.
38	
39	Keywords: asynchrony, GBIF, mismatch, phenology, pollination mode
40	
41	Introduction
42	Phenological events are periodically occurring events in the life cycle of organisms. The
43	timing of these events often depends on environmental factors such as temperature or

44 photoperiod, and it is well known that climate change affects some of these and thus changes

45 the phenologies of many organisms [1]. With such phenology shifts, there is increasing risk of

46	phenological mismatches between interacting organisms, potentially exceeding the natural
47	resilience of ecosystems [2]. Climate change-induced phenological shifts have been
48	documented extensively for individual species [3], but we still know much less about how
49	these shifts affect ecological interactions. Kharouba et al. [4] recently reviewed 54 published
50	interaction studies across ecosystems and interaction types and found no clear general trend,
51	with about half of the studied interactions becoming more asynchronous but the other half
52	becoming even more synchronized through climate change.
53	Plant-pollinator systems are among the biotic interactions expected to suffer most from
54	a mismatch of phenological events [5]. Several previous studies have observed mismatches
55	[6,7], but in others pollinators and plants seemed to be able to keep up with each other [8]. An
56	interesting question in this context is also which of the two partners is advancing faster if
57	there is an increasing mismatch. So far, the evidence here is also mixed. For instance [9]
58	found pollinators to advance faster than trees, and Parmesan [3] that butterflies advanced
59	faster than herbaceous plants, but in a study by Kudo & Ida [10] it was the plants – spring
60	ephemerals – that advanced faster than their bee pollinators.
61	Mismatches of plant-pollinator interactions can have negative consequences for both
62	partners. For the pollinators, this can include lower survival rates, a decreased overall fitness
63	and higher parasite loads [11]. Moreover, mismatches might also impact pollinator
64	demography, the body sizes [6] and frequencies of sexes, and thus population viability [11].
65	On the plant side, desynchronized pollinator interactions are mainly expected to impact plant
66	fitness and thus long-term population growth and survival. For instance, Kudo & Ida [10]
67	found that seed counts were reduced in early-flowering spring ephemerals after
68	desynchronization with their bee pollinators. However, in another study fly-pollinated plants
69	did not show similar responses [12].

70 Plants differ in their level of dependence on plant pollinators, and an intriguing question 71 therefore is to what extent phenology responses to climate change are linked to the pollinator 72 dependence of plants. Bond [13] theorized that wind-pollinated plants might experience little 73 negative consequences of climate change as they do not depend on interactions with animals. 74 Conversely, insect-pollinated plants may be subject to strong selection toward phenologies 75 that are in synchrony with their pollinators. This hypothesis was later corroborated in an 76 empirical study by Fitter & Fitter [14]. A more recent study on orchids [15] found that 77 pollination mode influenced the degree of plant advances in flowering phenology, indicating 78 that self-pollinating and thus pollinator-independent plants were not constrained by pollinator 79 phenology. The main idea of these previous studies is that all else being equal, pollinator-80 independent plants should exhibit stronger phenological shifts in response to the same climate 81 changes.

82 Testing hypotheses about plant-pollinator responses to climate change is not trivial. 83 Since changes in phenology take place on the scale of decades [16], we need long-term data. 84 A possible source of long-term data on plant phenology are herbarium specimens [17,18], 85 which can indicate the day of year that a specific species was flowering in a given location 86 and year. Herbarium data provide unique historical depth, but they need to be treated with 87 caution because of the sampling biases associated with them [19,20]. In recent years the 88 digitization of herbaria as well as other collections and observation data, including on other 89 taxa such as pollinating insects, e.g. from long-term monitoring networks, is creating an 90 increasing number of public data bases that contain vast amounts of natural history data that 91 cover large spatial and temporal scales [21]. These data bases are increasingly being used for 92 analyses of broad ecological trends and global changes [20,22]. One of the largest and most 93 important hubs of large-scale and long-term ecological data sets is the Global Biodiversity 94 Facility (GBIF), an intergovernmental initiative and public data base that provides access to

95 biodiversity data compiled from various individual sources like academic institutions,

96 government agencies or independent collections (GBIF, 2019).

97 Another matter is finding a measure for changes in phenology. Primack et al. [23] 98 demonstrated that the average collection date of the herbarium specimens of a plant species in 99 a year can be used as a proxy for peak flowering time in that year. The same approach of 100 using occurrence records in natural history collections or other data bases can in principle be 101 used to estimate the activity times of other groups of organisms such as insects (4 and 102 references therein). For instance, analyses of natural history collections in the UK have 103 demonstrated phenology changes in bees [7] and butterflies [24]. Thus, the peak occurrences 104 of plants and insects in GBIF may be used to estimate activity shifts of different groups, as 105 well as their synchrony. When we use the term 'activity' in this paper, we refer to the period 106 in an organism's life when it can interact with its ecological partner. For plants this is the 107 period of flowering, for insect pollinators the period of flight. 108 We used data from GBIF to study phenological mismatches between plants and 109 pollinators in Germany, at the level of taxonomic groups as well as individual interactions. 110 We asked the following questions: (i) Are there long-term trends in the phenology of plants 111 and pollinators? (ii) If yes, are phenology trends related to climate change? (iii) How are 112 phenological changes of plants related to their pollinator dependencies? (iv) How does 113 climate change affect the synchrony of plant-pollinator interactions? 114

115 Methods

116 Phenology data

117 We worked with occurrence records of plants and insects available from the GBIF database

118 [25–36]. For the plants, we restricted ourselves to species covered by the BioFlor database of

119 plant traits [37], because we needed to be able to classify plants by their level of pollinator

120 dependence (see below). For the insects we restricted ourselves to beetles (Coleoptera), flies 121 (Diptera), bees (Hymenoptera) as well as butterflies and moths (Lepidoptera), as these groups 122 contain most insect pollinators [38]. We used the R package rgbif [39] to download all 123 available records of the above taxa from GBIF. Our basic criteria for including records were 124 that they originated from Germany, and that they referred to either a living specimen (e.g., a 125 captured insect), a human observation, just an observation (i.e., when the exact type of 126 observation was not clear), or a preserved specimen (e.g., an herbarium record). If names of 127 plant species were not accepted names, we used the R package *taxsize* [40] to check the 128 names against the GBIF backbone taxonomy and determine the actual accepted. 129 Prior to the data analyses, we subjected the data to several steps of quality control 130 (Figure S1). First, we removed all records from before 1960 as these turned out to be too 131 inconsistent, with few records per year and large gaps between years with records. We also 132 removed the records from 2020 as the year had not been complete at the time of our analysis. 133 Second, we removed all records from the first and last days of years because the high number 134 of records on those days indicated that records without a recorded collecting date had been 135 given these as default dates. Next, we removed all records from "GEO Tag der Artenvielfalt", 136 a German bioblitz event where large numbers of records are taken on a specific day of the 137 year. Including these data would have strongly biased the intra-annual distributions of our 138 records. Finally, we removed the records from several collections which appeared to have 139 misclassified these as being of German origin, probably through a combination of coordinate 140 rounding and determining countries of origin automatically from these coordinates. We 141 identified these sets of records by visually inspecting the geographic distributions of the 142 records of each institution; most of these erroneous data sets were from Luxembourg (Table 143 **S1**). There were a few records just outside the boundaries of Germany that we did not remove 144 from our data set because the country information appeared trustworthy and we suspected

145	errors with the recording of the coordinates. Obviously, the latter steps of our quality control
146	were possible only for georeferenced records, which made up 99.97% of the total amount of
147	records. After these data curation steps, we maintained around 11 million plant records and
148	over one million insect records for our data analysis. There were large differences between
149	plants and insects not only in the numbers of records but also in their temporal distribution
150	across the studied period (Figure S2). While plants, but also beetles, had relatively even
151	record numbers across decades, the other insect groups, in particular flies and bees, were
152	strongly underrepresented in the earlier decades, and record numbers increased rapidly only in
153	the last 20 years, probably due to the advent of platforms like iNaturalist.org and
154	naturgucker.de, which allow logging of species occurrences by citizen naturalists, and which
155	make up most of our insect data. Beetles were represented, save for one species from the
156	Orsodacnidae, by the Chrysomelidae family.
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170 were apomicts, were classified as pollinator-independent. If none of the above applied, we

171 assigned an intermediate pollinator dependence. If part of the information above was missing,

172 no pollinator dependence was determined, and the species was excluded from the analyses

173 involving pollinator dependence.

Finally, we obtained data on individual plant-pollinator interactions from a UK database on plant-pollinator interactions hosted by the Centre for Ecology and Hydrology (CEH). This database included all known interactions between plants and flower-visiting 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, we could not find similar data for our study area.

181

182 Calculation of plant and insect phenology

183 For our analyses of plant flowering phenology and pollinator activity times, we averaged all 184 records of a plant or insect species in a year to calculate each year's mean day of the year 185 (DOY) of the occurrence of a species. As discussed above, this occurrence measure was used 186 as an estimate of each year's peak flowering or peak activity time of plants and insects, 187 respectively. Each annual mean DOY was calculated from at least five records of a species 188 per year. To avoid extreme shifts based on too little data, we included only species with 189 records in at least 40% of the years. The median number of records per year for a species in 190 our analyses was 47.

191 Since our analyses of individual plant-insect interactions (see below) were done at the 192 level of decades, we additionally calculated the decadal means, based on nominal decades (0-193 to-9), of species DOYs for each of the included species, and only when at least five records 194 existed per decade. These decadal interaction analyses were done only from 1980 onwards,

195 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.

197 After clean-up and averaging, a total of 58,895 annual and 1,336 decadal peak DOYs, 198 with the latter based on a median number of 1,686 records, remained in our data set (). The 199 annual activity data included 1,274 plant and 88 insect species. For 948 of the plant species 200 we had information about pollinator dependence: 144 were pollinator-dependent, 204 201 pollinator-independent, and 600 were classified as intermediate. The 88 insect species 202 consisted of 40 species of beetles, 44 butterflies and moths, three bees and just one fly 203 species. The decadal data included 245 plant and 26 insect pollinator species. All data 204 wrangling and analysis was done in R [43].

205

206 Data Analysis

207 To understand phenology changes in plants versus insects, we first estimated the average 208 phenological shifts in each group. We defined phenological shifts as the slope of the linear 209 regression linking the peak activity (= mean annual) DOY of an individual species to the year 210 of observation. We visually confirmed approximate normal distribution of the individual-211 species slopes, and that no improbable outliers were present. There were some plants with 212 rather extreme values (Figure S3), however these were mostly early-flowering plants which 213 likely experience stronger pressures and therefore stronger phenology shifts [44], and we 214 therefore did not exclude them from our analyses. We compared the mean phenological shifts 215 between plants and insects using an independent-sample Welch's t-test, and we further 216 examined the temporal trends between different insect orders and plants in an ANOVA, using 217 a Tukey post-hoc test to determine pairwise differences. We excluded bees and flies from the 218 last step as their numbers were too small to be representative for their respective groups.

219 Different climatic factors likely affect the timing of early and late activity periods, 220 which might complicate the interpretation of the peak shifts. We therefore also assessed the 221 extent of shifts of first and last day of activity for each species (and consequently the duration 222 of their activity) to understand how asymmetries in the shifts might affect the peak shifts in 223 phenology. For this we estimated the shifts of the decadal average first and last activity day of 224 the year over time in a linear model. We also estimated the shifts of duration of the activity 225 period by first calculating the yearly duration of the activity period as the difference between 226 the last recorded day of activity and the first for each species, taking the decadal average of 227 said duration and then estimating the shift over time in a linear model. We used decadal 228 averages to ensure the differences were due to long-term trends, as the absolute first and last 229 day of activity is just the first and last record of a species in that year and therefore subject to 230 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.

243	Finally, we analyzed asynchrony between plants and pollinators using the data on
244	individual plant-pollinator interactions. For each plant and presumed insect pollinator, we
245	calculated the absolute difference in peak activity times for each decade. A value of zero thus
246	indicated perfect asynchrony, and higher values indicated increasing asynchrony. To test
247	whether asynchrony changed over time we estimated the slopes of the relationship between
248	differences in peak activities and time (decades) for each plant-insect interaction with a linear
249	model. Here, negative slope values indicated a shift towards greater synchrony, and a positive
250	slope a shift towards greater asynchrony. Altogether, there were 1,797 interactions involving
251	245 plants and 26 insect pollinators, one insect usually associated with multiple plants but
252	seldomly plants with multiple insects. To test for differences in average asynchrony and
253	change of asynchrony between insect groups, we used an ANOVA and assessed pairwise
254	differences with a Tukey post-hoc test.

256 **Results**

257 Temporal trends in plant and insect phenology

258	The analysis of the peak activity data showed a strong difference in the average temporal
259	shifts of plant and insect phenology (Welch's $t_{100.929} = 6.644$, $P < 0.001$). The phenology of
260	plants generally advanced much more strongly, with an average shift of -4.5 ±0.2 days per
261	decade (mean \pm SE), while across all insects the shift was only -0.4 \pm 0.6 days per decade.
262	84.8% of all plant species but only 56.8% of all insect species advanced their phenology
263	(Figure 1). However, these numbers across all insects obscured different trends among the
263 264	(Figure 1). However, these numbers across all insects obscured different trends among the insect orders: when considered separately, butterflies/moths exhibited a strong phenology
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268	beetles were analyzed as separate groups, ANOVA indicated significant differences among
269	them ($F_{2,1355} = 25.16$, $P < 0.001$), with significant pairwise differences (Tukey post-hoc, $\alpha =$
270	0.05) between the phenology shifts of beetles and plants, and beetles and butterflies/moths,
271	respectively. (Figure S4A and Figure S5.)
272	We found asymmetries between the slopes of first and last day of activity over time
273	(Figure S6). In plants, the symmetry was generally skewed towards a stronger shift of the
274	first day of activity (First: -1.2 \pm 0.0 mean days/decade \pm SE, Last: 0.5 \pm 0.0 mean
275	days/decade \pm SE) with butterflies/moths behaving similarly (First: -1.4 \pm 0.1 mean
276	days/decade \pm SE, Last: 0.5 \pm 0.2 mean days/decade \pm SE), whereas in beetles the last day of
277	activity shifted more strongly (First: 0.1 ± 0.1 mean days/decade \pm SE, Last: 0.5 ± 0.1 mean
278	days/decade \pm SE). It is also notable that the plants' and butterflies/moths' day of first activity
279	generally advanced while the beetles' day of first activity was rather delayed (Figure S7).

281 Climate sensitivity of plant and insect phenology

282 The climate sensitivities of the phenologies of plants, butterflies/moths and beetles generally 283 resembled their temporal trends (Figure 1), and group differences in climate sensitivities 284 matched those in temporal trends described above. Again, there was a significant difference 285 between plants and all insects (Welch's $t_{96,026} = 8.027$, P < 0.001), with plants showing a 286 strong negative association between peak activity and temperature, but a much weaker 287 association for all insects together. On average, plant peak flowering shifted by -7.6 ± 0.2 288 days per °C (mean \pm SE), and 92.5% of the individual species showed earlier flowering with 289 increasing temperature, whereas for insects it was only -1.3 ± 0.8 days per °C, and 63.6% 290 showing a trend towards earlier peak activity (Figure 1). When the butterflies/moths were 291 considered separately, however, they showed a fairly strong association with temperature, 292 with an average peak activity shift of -4.4 ± 0.8 days per °C (mean \pm SE) and 80% of the

293	individual species advancing, whereas the beetles showed an opposing trend of delayed peak
294	activity, with an average of +1.4 \pm 1.1 days per °C temperature change. There were
295	significant differences among the three groups (ANOVA, $F_{2,1355} = 45.701$, $P < 0.001$), with
296	significant differences between all pairwise combinations (Tukey post-hoc, $\alpha = 0.05$). (For an
297	overview over all groups, see Figure S4B and Figure S8.)
298	
299	Pollinator dependence
300	The phenology of plants, and its temporal trends, differed very little among plant groups of
301	different levels of pollinator dependence (Figure S9). The peak flowering of pollinator-
302	independent plants (average DOY 199.5) advanced on average by -3.9 days per decade, while
303	pollinator-dependent plants (average DOY 196.2) advanced by -5.1 days per decade, and
304	intermediate plants (average DOY 199.5) advanced by -4.5 days per decade. In all three
305	groups, the percentage of plants advancing was 85-86%. None of the differences between
306	groups was statistically significant.
307	

308 *Synchrony of plant-pollinator interactions*

309 When examining the synchrony of individual plant-pollinator interactions, we found that the 310 three pollinator groups differed in their average levels of asynchrony with the plants, but that 311 interactions did not become more asynchronous but rather more synchronized during the last 312 decades (Figure 2A). The temporal trends differed strongly among the pollinator groups 313 (ANOVA, $F_{2, 2522} = 67.750$, P < 0.001; Tukey's post-hoc test significant at $\alpha = 0.05$ for all 314 pairwise comparisons: **Figure 2C**): while the synchrony of plant-butterfly interactions 315 remained on average unchanged, plant-pollinator interactions involving bees shifted on 316 average by -2.7 days per decade, with 68% of individual interactions decreasing asynchrony 317 over time. The strongest shifts were in plant-hoverfly interactions which shifted by -6.2 days

318 per decade, with 89% of all interactions showing decreasing asynchrony (Figure 2A, C). In 319 all three plant-pollinator groups, asynchrony was mostly due to earlier peak activity of the 320 insects (Figure 2B). Interestingly, however, there was a tendency for these patterns to 321 disappear in all three groups over time, presumably because of the stronger phenology shifts 322 of plants (Figure 1). Plant-hoverfly interactions ($n_{Insect} = 1$, $n_{Plant} = 132$, $n_{total} = 132$) became 323 on average synchronous in the last decade. For the plant-butterfly interactions ($n_{Insect} = 36$, 324 $n_{Plant} = 231$, $n_{total} = 1,819$) the linear model predicts the point of synchrony to be reached in 325 2029, and for the plant-bee interactions ($n_{\text{Insect}} = 4$, $n_{\text{Plant}} = 214$, $n_{\text{total}} = 574$) in 2050. 326

327 Discussion

328 In this study, we took advantage of large collections of occurrence records to examine 329 phenological trends of flowering plants and insect pollinators in Germany. We asked whether 330 phenology changes affected the synchrony of plants and insects, and whether observed 331 changes in phenology, and variation therein, were related to the different groups' responses to 332 climate warming. We also examined whether the phenology responses of plants depended on 333 their levels of pollinator dependence. Our results showed that the phenological shifts of plants 334 and insects indeed differed, with plants shifting by several days per decade while insects on 335 average shifting hardly at all. As peak flowering historically occurred after peak insect 336 activity, these trends imply an increase in plant-pollinator synchrony during the last decades, 337 but a potential for future desynchronization if climate change continues. 338 Plants and insects also differed in their overall temperature sensitivity. While plants 339 shifted on average by over a week per degree of warming, insects shifted by only one day. 340 There were large differences between insect orders in their phenology trends and temperature 341 sensitivities. As groups with greater temperature sensitivity also showed larger phenology 342 shifts over time, it seems likely that the two are causally related, i.e., that anthropogenic

343	climate warming is responsible for the observed phenology shifts. Lastly, there were no
344	differences between pollinator-dependent and -independent plants, suggesting that plants
345	either responded passively to temperature, with advanced flowering in warmer years
346	irrespective of pollinator dependence, or that most plants have sufficient generalist pollinators
347	that can fill in for other, desynchronized pollinator species and thereby reduce selection
348	pressure on plant phenology.
349	
350	Caveats
351	When interpreting the results of our study, it is important to consider some caveats of the
352	collections data and occurrence records we used. For instance, the temporal distribution of
353	collections data is usually quite heterogenous, and so was our data (Figure S2). Our analysis
354	of the shifts of the first and last days of activity may thus be influenced by varying
355	observation efforts over the years. Particularly, the increasing popularity of nature observation
356	platforms such as www.naturgucker.de, whose records are contained in GBIF, may have
357	resulted in higher probability of detecting early and late occurrences. Besides temporal
358	heterogeneity, occurrence records are usually also not homogenously represented in space.
359	Our study's measure of phenology, peak occurrence time, does not account for temporal
360	variation of spatial representation of records within Germany, although some areas might be
361	over- or underrepresented in some parts of the studied period. Moreover, our study also does
362	not account for spatiotemporal variation in macro- and microclimate which can influence
363	intraspecific variation in phenology shifts [45] and could therefore potentially induce local
364	mismatches.
365	When estimating insect peak activity, we did not account for the earlier life stages of
366	insects appearing in the data, despite being not important for pollination. This bias could be

367 most relevant for butterflies and moths, as their larval stages are more conspicuous than fly

368	and beetle larvae. Butterflies/moths are, however, the group with the latest peak activity times
369	for large parts of the studied period, so this bias is either not strong or we are underestimating
370	how late in the year butterflies and moths occur. Similarly, some plants occurrences may have
371	been recorded when plants were not flowering. Flowers are important for plant species
372	identification, and herbarium records are usually made from flowering specimen, but we
373	cannot rule out that some plant occurrence records were based on vegetative plants alone.
374	Finally, in our analyses we focused on peak activity and therefore did not consider the degree
375	of overlap between the flight times of pollinators and the flowering of plants. However, if the
376	durations of activity periods change, then the relative overlap of two interacting groups could
377	change in spite of identical activity peaks, or vice versa. Testing such possibilities with
378	occurrence data, however, requires even higher-resolution data for individual species than in
379	our study.
380	

381 *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.

The extent to which plants advanced their phenology in our data is comparable to that found by Fitter & Fitter [14] in their long-term observation study of changes in first flowering dates of hundreds of plant species in England. They compared flowering during 1991-2000 to that between 1954 and 1990 and found an average advancement of 4.5 days. This is surprisingly congruent with our observation of 4.5 days advancement per decade over the

393	whole period from 1960 to 2019. A more recent long-term analysis of phenology changes in
394	subalpine meadow plants in the Rocky Mountains was undertaken by CaraDonna et al. [46]
395	who found an even stronger average advancement of first flowering of 6.4 days per decade.
396	Since CaraDonna et al. [46] also analyzed peak floral abundance, their data should be
397	particularly comparable to our estimation of peak flowering through the DOY of peak
398	occurrence. They found a rate of advancement of 5.3 days per decade in spring peak
399	abundance but only 3.3 days for the summer peak floral abundance. Our results of peak
400	occurrence across the whole year thus fall in between these two estimates.
401	For insects, previous studies seem to be less consistent, with widespread but not
402	universal advances in springtime phenology (mostly associated with warming) over the last
403	decades [47]. For butterflies, long-term records showed that their times of first flights
403	decades [47]. For butterflies, long-term records showed that their times of first flights
403 404	decades [47]. 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 2000s
403 404 405	decades [47]. 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 2000s compared to the previous decades in England [48], and by -7.7 days per decade in California
403 404 405 406	decades [47]. 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 2000s compared to the previous decades in England [48], and by -7.7 days per decade in California [49]. The magnitude of the shifts observed in England is similar to what we estimated for

409 *Temperature sensitivities of plant and insects*

410 We found that associations between temperature and phenology differed among groups but 411 that the magnitude of these associations generally reflected the different groups' phenology 412 shifts observed over time. This strongly suggests a link between the phenology shifts and 413 climate change, corroborating previous studies such as the ones by CaraDonna et al. [46] and 414 Song et al. [45]. We found that plants were generally more sensitive to temperature, i.e., their 415 phenology advanced more strongly, than insect pollinators. Previous studies on insect 416 phenology in the temperate zone (reviewed in [47]) have shown that increased spring 417 temperatures are often associated with earlier insect emergence, but that this pattern cannot be

418 generalized as easily as for the plants, as temperature–phenology relationships of insects are 419 more complex. While many insects plastically respond to warmer temperatures by speeding 420 up their rates of development (and thus potentially emerge earlier), others have been found to 421 respond in counterintuitive ways and delay their phenology. This might be due to dependence 422 on other cues such as rainfall [50], due to cold period requirements of insects during their 423 diapause (climate warming can cause a loss or reduction of this chilling period, and this tends 424 to increase the amount of warming required for subsequent emergence; [47]), or because 425 species overwinter in a diapause state in which they are not temperature sensitive [51]. Fründ 426 et al. [51] also showed that bees overwintering in larval stages responded to higher winter 427 temperatures with delayed emergence, while bees overwintering as adults showed advanced 428 emergence (but had greater weight losses during overwintering). We did see delayed 429 phenology in some of our data, particularly for beetles and bees. This also connects well to 430 some of the findings reviewed by Forrest [44], for instance that during winter above-ground 431 nesting bees experience different temperatures than the plants they feed on during the 432 summer. Such microclimate differences between insects and plants during overwintering may 433 sometimes explain contrasting climate responses. In other cases, delays in the first appearance 434 of adults may result from longer growing seasons. For example, longer growing seasons have 435 reduced selection for rapid development in some high-elevation grasshoppers, in such a way 436 that they reach maturity later — but at a larger size — than in the past [52]. Furthermore, 437 warming can change the number of generations per year (voltinism; [47]). All the above-438 mentioned mechanisms can cause variation in the phenology shifts of insects with climate 439 warming and may therefore explain why climate change is not always accompanied by 440 phenological advances but might also cause delays – as we observed for the beetles. 441 Another interesting idea is that the phenological advancement of the plants itself could 442 cause delayed phenology of some pollinators. Wallisdevries & van Swaay [53] 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 [54].

448 Pollinator-dependence of plants

449 We did not find any differences in the phenological changes of pollinator-dependent versus 450 pollinator-independent plants. This result is consistent with Rafferty & Ives [55] who found 451 that the phenology shifts of plants were not constrained by their pollinators, because these 452 kept pace with the plants. In contrast, Kudo et al. [56] found a negative effect of flowering 453 advancement in bee-pollinated but not fly-pollinated plants. Fitter & Fitter [14] found 454 significant differences between insect-pollinated plants (-4.8 days shift in day of first 455 flowering) versus wind-pollinated plants (-3.5 days shift) and suggested this was because 456 shifting pollinator activity forced plants to flower earlier. In our study we did not find any 457 such differences, indicating that plant responses to temperature are either entirely passive, or 458 that most plants have generalist pollinators with a long period of activity, so that there is little 459 selection pressure on plant phenology. The data set used in our analysis is larger than those 460 used in the studies cited above, so our results may be regarded as more conclusive and more 461 general, bearing the limitations of the collections data in mind.

462

463 *Changes in plant-pollinator synchrony*

464 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

466 phenology of plants generally advanced faster than that of the insects during the last decades,

467 but plants had generally been the later partner in most plant-pollinator interactions, these

468	shifts lead to greater synchrony overall. However, if the observed trends continue, then many
469	of the studied interactions will soon reach points of perfect synchrony, and after that the
470	interactions may become more asynchronous again, albeit in the other direction. For plant-
471	hoverfly interactions this point has already been reached. With linear trends and if we assume
472	that observed trends will continue, the points of reversals are expected in approximately 10
473	years for plant-butterfly interactions and in around 30 years for plant-bee interactions. If
474	interactions will become more asynchronous again in the future, then resilience of pollinator
475	networks, in particular through pollinator generalism, could buffer some of the impact of
476	phenological mismatches [6], and our finding of no differences between pollinator-dependent
477	and pollinator-independent plants support this idea. However, while generalist pollinators
478	make up the larger part of the interactions in most pollination networks, some plant-pollinator
479	interactions are highly specialized, and these might be the ones suffering most from future
480	mismatches [57].
481	
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485	
186	Authors contributions

486 Authors contributions

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

490

491 Data availability

492 The R code used to conduct the analysis is available at https://github.com/jonasfreimuth/Phenological-

493 shifts-germany.

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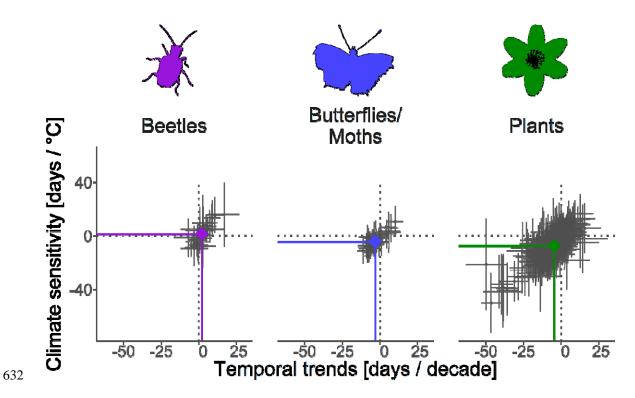


Figure 1. Temporal trends (days per decade) versus climate sensitivities (days per °C temperature change) of the phenology (peak flowering/activity) of plants, beetles, and butterflies/moths, with the colored lines indicating the averages for each group. Grey dots indicate individual species means with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

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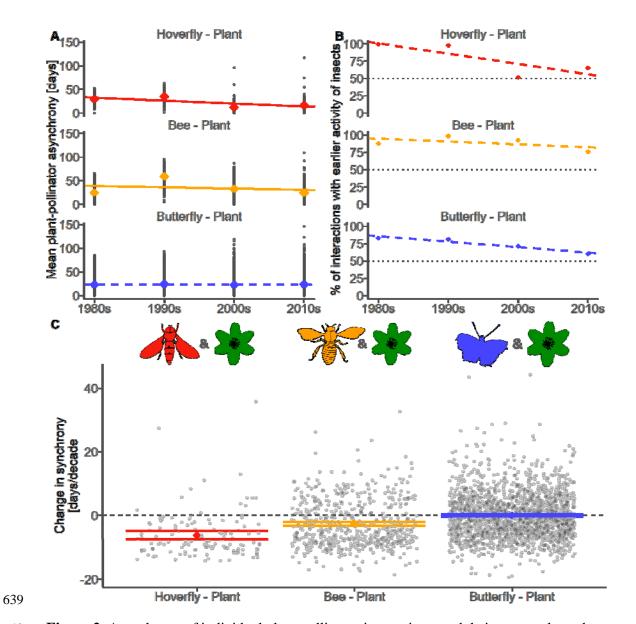


Figure 2. Asynchrony of individual plant-pollinator interactions, and their temporal trends,
separated by pollinator groups. (A) Decadal changes of asynchrony (grey dots/lines:
individual interactions; colored diamonds/lines: linear regression for each group. (B) Fraction
of interactions with earlier insect activity. (C) Average decadal asynchrony 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.

