1	Phenology and flowering overlap drive specialization in pollinator networks
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50	pollen, seasonality, adaptive foraging

52 Variation in diet breadth and specialization stems from fundamental interactions species 53 have with their environment¹⁻³. Consequently, understanding the drivers of this variation is key to 54 understanding ecological and evolutionary processes, and will facilitate the development of 55 predictive tools as ecological networks respond to environmental change^{4,5}. Diet breadth in wild 56 bees has been an area of focus due to both their close mutualistic dependence on plants, and 57 because both groups are under threat from global biodiversity loss⁶. Though many of the principles 58 governing specialization for pollinators have been identified^{7,8}, they remain largely unvalidated. 59 Using mechanistic models of adaptive foraging in pollinators^{9,10}, we show that while temporal 60 resource overlap has little impact on specialization in pollinators with extended flight periods, 61 reduced overlap increases specialization as pollinator flight periods decrease. These results are 62 corroborated empirically using pollen load data taken from bees with shorter (genus Andrena) and 63 longer (genus Lasioglossum) flight periods across environments with both high and low temporal 64 resource overlap. This approach reveals how interacting phenologies structure plant-pollinator 65 networks and drive pollinator diet breadth via the temporal overlap of floral resources.

66 The dietary specialization of an organism sits at the nexus of numerous fundamental processes in 67 ecology and evolutionary biology. Diet breadth helps define organismal niches¹, mediates migration and 68 dispersal potential^{2,11}, shapes inter- and intraspecific competition^{3,12}, influences species persistence in the face of environmental disturbance⁴, and can affect rates of speciation and extinction⁵. Even the simple 69 70 assembly of the collection of dietary ranges of organisms in a community defines one of the major 71 concepts in ecology, the food web. Therefore, identifying the drivers influencing diet breadth and 72 specialization provides fundamental steps towards understanding a multitude of essential biological 73 questions.

Much of the progress in understanding specialization in consumers comes from studies of
herbivory¹, particularly on insects¹³. Using wild bees and plant-pollinator networks as our study system,
we extend the scope of those studies to include mutualistic interactions. A mutualism is a ubiquitous
ecological interaction in which participant species benefit each other¹⁴, and pollination therefore differs

from other insect herbivory because of the reproductive benefit offered to both consumer and consumed species. Bees depend on removing pollen from flowers to rear their offspring, providing a pollination service at the same time¹⁵. Pollination is also of particular interest given its role in supporting terrestrial biodiversity⁶ and agricultural output¹⁶, a role threatened by widespread declines in both pollinators and insect pollinated plants^{6,17}.

83 Research on insect herbivores has largely focused on how adaptation to plant traits and defenses 84 drives consumer concentration on single plant families^{18,19}. However, other drivers of diet breadth have 85 been identified (ref 7.8, see Table S1). Here, we expand on the fundamental effect of resource density on 86 a consumer's diet breadth and specialization⁷, not in a spatial context, but temporally. This resource 87 density is important in its interaction with consumer phenology. For long-lived organisms (e.g., predatory 88 vertebrates), opportunities to access a single resource type principally depend on the resource's spatial 89 distribution. If a single resource type is physically dense, then a predator can specialize on it without 90 needing to turn to alternatives, whereas if it is patchily distributed then a strategy of specialization is less 91 optimal. However, for a comparatively short-lived organism like a bee whose adult flight period can be 92 measured in days, the distribution of a resource type in time is the principal determinant of accessibility. 93 A plant species that blooms for 30 days is *de facto* dense, continuously available resource for a bee 94 species whose flight period lasts for only 25 days, but sparse and patchily distributed for a species which 95 flies for 90 days. The temporal pattern of flowering can therefore be considered the functional equivalent 96 of resource density for short-lived organisms. These ideas have proven influential but require concrete 97 theoretical and empirical validation to solidify their conclusions.

98 Expanding upon this idea of temporal density, we hypothesize that the degree of temporal overlap 99 in the availability of different resources (flowering overlap among co-occurring plant species) in 100 combination with the fundamental life history phenologies of bees (adult flight period) determines 101 realized diet breadth. Applied to plant-pollinator systems, we hypothesize that reduced flowering overlap 102 will reduce the diet breadth of short-lived but not of long-lived pollinators (Fig. 1). Shorter-lived 103 pollinators will experience markedly different resource availability based on the degree of temporal overlap of co-flowering species. High flowering overlap will allow shorter-lived pollinators a greater
number of options while low overlap will restrict options. Longer-lived pollinators will be less constricted
by the temporal overlap of co-occurring resources, often having equal number of options regardless of the
degree of overlap (Fig. 1). Importantly, our hypothesis does not indicate that flight period will necessarily
correspond with being more or less specialized. Instead, it describes how temporal resource overlap will
affect diet breadth as a function of a pollinator's flight period.

110 We employed a multi-faceted approach to test our hypothesis. First, we leverage advances in

111 modeling adaptive foraging in ecological networks^{9,10} to directly address the phenological mechanism

driving diet breadth across different pollinator flight periods. Second, we utilize our intercontinental

pollen load dataset of shorter lived *Andrena* bees and longer lived *Lasioglossum* bees from both a highly

seasonal and a less seasonal environment which function as low and high temporal resource overlap

treatments, respectively, to empirically validate our model outputs.

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117 Results

118 Overview

119 Plant-pollinator dynamics are modeled using a dynamic consumer-resource approach that incorporates adaptive foraging of pollinators to mechanistically model pollinators' consumption of floral 120 121 rewards and reproductive services to plant species^{9,10} (see Supplementary Methods Table S2). The 122 model's adaptive foraging mechanism serves as a useful tool to implement and measure changing 123 pollinator diet breadth (see Supplementary Methods). Phenology is integrated into the base model through 124 modified sinusoidal wave functions which produce unique phenologies for every plant $i(T_{p_i})$ and animal pollinator $j(T_{a_i})$. These functions T_{p_i} and T_{a_i} regulate floral reward production, adaptive foraging rates, 125 126 pollinator visitation rates, and consequently, plant and pollinator reproduction across time (Tables S3-S4). 127 The malleability of our phenology functions facilitates direct control over the availability of specific 128 resources and the activity patterns of the animal pollinators over time in simulations (see Supplementary

Methods, Fig S1-S5). The quantitative degree of overlap in co-occurring floral resources can be measured as the Total Resource Overlap (TRO) of the entire plant community, and the Averaged Resource Overlap (ARO) per plant species. Activity patterns of animal pollinators are set up to produce a range of different flight period lengths separated by differing lengths in between flight periods (Fig S6). Simulations use three fully connected bipartite network frameworks across 2072 combinations of plant and pollinator phenologies to produce 62160 unique networks used to model over 22 million plant-pollinator interactions (see Supplementary Methods).

136 Our pollen load data (see Supplementary Methods) comprises a large sample set of two bee 137 genera, Andrena (Andrenidae) and Lasioglossum (Halictidae), collected both in the state of Michigan, 138 USA (Nearctic) and the United Kingdom (Palearctic). Restricting our geographic scale to these two 139 specific Holarctic locations aids in our ability to compare across both bee and plant communities due to 140 their shared evolutionary history and overall similarity (see Supplementary Methods). The highly 141 seasonal continental climate of Michigan²⁰ produces lower degrees of overlap and highly seasonal flowering communities, while the more mild oceanic climate of the British Isles²⁰ and its consistent 142 143 temperatures produces longer, overlapping flowering times among plant species²¹.

144 Diet breadth or the degree of specialization in pollinators is measured both in-model with output 145 on pollinator foraging effort and empirically with relative abundance of different pollen in bee pollen 146 loads. We used two metrics that were applicable to both theoretical and empirical results for better direct 147 comparisons, because simply counting unique taxonomic groups in the pollen data will not be applicable 148 to theoretical results. We developed the first metric and labeled it Deviation from Generalism (DFG) (see 149 Supplementary Methods, Table S5). Briefly, DFG is the normalized summation of all pairwise differences 150 of pollinator foraging effort (or relative pollen load) on potential floral resources (see Supplementary 151 Methods, Table S5). It ranges from 0 (perfect generalist) to 1 (perfect specialist). The second metric was 152 the Coefficient of Variation (CV) of a pollinator's foraging effort (or pollen load). Higher CV values 153 occur with outlying values indicating specialization. With model results, both DFG and CV were applied

to final foraging levels at the end of simulations as well as the average foraging levels across the last 1000time steps (see Supplementary Methods, Fig S7).

Model Results

157 Our simulation results (Figs. 2, 3) support our hypothesis (Fig. 1) demonstrating strong effects of 158 temporal resource overlap on the diet breadth of short- but not long-lived pollinators. Fig. 2 shows the 159 results of one network as an illustration of how resource overlap explains large portions of the variation in 160 diet breadth of pollinators with shorter flight periods (Fig. 2a, c), but little for the diet breadth of 161 pollinators with longer flight periods (Fig. 2b, d). The transition from high explanatory power to low 162 explanatory power is apparent as pollinator flight period increases regardless if specialization was 163 measured at the end of simulations or averaged across the last 1000 model time steps (Fig 3). This result 164 is also consistent across varying proportional lengths of time between pollinator flight periods using 165 either the DFG or CV metric across all networks tested (Fig S8-Fig S10). As predicted, low levels of 166 resource overlap drive pollinators with short flight periods to specialize because potential resource 167 options are limited at any given moment in time. As temporal resource overlap increases, potential 168 options increase resulting in a more generalized diet breadth (Fig. 2a, c). Long-lived bees, on the other 169 hand, do not experience the same limitation of potential diet options due to temporal resource overlap, 170 because they are active most of the entire flowering season and can potentially access most or all of the 171 flowering plants. Results were consistent regardless of whether resource overlap was measured as total 172 resource overlap (TRO) or average resource overlap (ARO).

173 Pollen data results

Our empirical results support our hypothesis, showing more specialized diets of short-lived bees in more seasonal environments (Michigan) with lower flowering overlap, than in less seasonal environments (UK) with higher flowering overlap. Our pollen data from Michigan presents more single family specialists than the less seasonal UK dataset (Fig S11). Dietary specialization between bee genera and regions varied significantly at both the level of botanical family (χ^2 =27.1, p<0.001, Fig 4a) and genus (χ^2 =25.3, p<0.001, Fig 4b). In accordance with our hypothesis, there was an effect of region at the

botanical family level, but only for the short-lived *Andrena* where species in Michigan were significantly
more specialized than their British counterparts. In contrast, there were no differences for long-lived *Lasioglossum*. At the botanical genus level, the same trend was more strongly expressed, with *Andrena* in
the UK showing a more generalized diet than their Michigan counterparts, but with no differences from *Lasioglossum* in either Michigan or the UK. Results were consistent when analyzed using the DFG metric
(Fig S12).

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187 Discussion

188 Our model results mechanistically demonstrate how differing phenologies interact with varying 189 degrees of temporal resource overlap to alter pollinator diet breadth. These results are corroborated 190 through our regional pollen load data which also presents the first empirical validation of benchmark 191 theories on adaptive foraging by consumers in general⁷ and pollinators in particular⁸. Furthermore, our 192 results suggest broader implications regarding the drivers of bee diet breadth and diversification. 193 Global patterns of herbivory, particularly in insects, show strong trends of specialization at the botanical family level¹³. These patterns have been linked to phytochemical diversity^{19,22,23} which 194 195 consequently drives an increase of both dietary specialization and species diversification in many 196 herbivorous insect groups across a latitudinal gradient towards the tropics, most clearly seen and well-197 studied in Lepidoptera^{22,24}. However, bees break sharply from this pattern, with a relatively low diversity 198 in the tropics when compared to Mediterranean and xeric environments^{8,25,26}. Tropical environments are dominated by highly generalized, often social species that visit a huge variety of botanical families^{27,28}. 199 200 Tropical areas have the highest levels of global phytochemical diversity, but their flowering patterns have 201 wider phenological variation than temperate areas at both the inter- and intraspecific levels. This results in 202 tropical floral communities that *de facto* show higher resource overlap in any particular year^{29,30} than 203 temperate communities. Our theoretical and empirical results suggest that bees' strong departure from the typical global pattern in herbivory¹³ may be due to a relatively stronger effect of phenology on their diet 204 205 breadth than the phytochemical restrictions suggested for other insect herbivores. In fact, tropical bees

206 have not evolved anything approaching the same high degree of dietary specialization as seen in the

- 207 tropical Lepidoptera despite existing in the same extremely phytochemically diverse landscapes. For
- 208 further discussion of this point, see Supplementary Discussion.
- 209 Our modeling developments present a flexible operational basis going forward. Integration of
- 210 time dependent functional components into species behavior and traits show that even high dimensional
- 211 network models can be further expanded in tractable ways to include the dynamics of organismal
- 212 phenology. Additionally, our metrics for diet breadth complement each other (see Supplementary
- 213 Methods) and reflect past methods (Fig S10). These metrics also integrated well with empirical pollen
- 214 load data sets that can be used to vet theoretical predictions. By expanding pollen datasets globally and
- 215 incorporating empirically vetted plant-pollinator phenology, future research can address how foraging
- 216 competition occurs across overlapping phenologies and what effects changing climatic conditions can
- 217 have on network stability.
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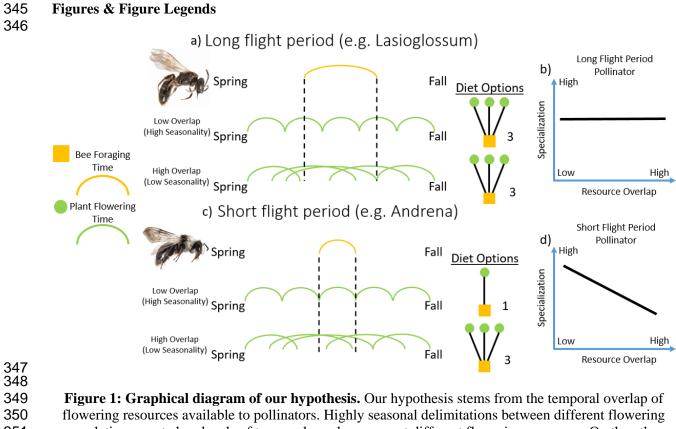
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- Author Contributions: F.S.V. conceived the project. T.J.W., P.G., F.S.V., and J.R.M. developed the
 conceptual basis for the project. T.J.W. identified suitable collection sites and bee genera for the empirical
 study design, collected bee and pollen data, and identified specimens. P.G. and F.S.V. developed dynamic
 model, simulation design, code, and diet breadth metrics. P.G. implemented simulations and analyses.
 T.J.W. and P.G. wrote the first draft, and all authors edited and revised the manuscript.
- Supplementary Information: Supplementary Methods, Supplementary Discussion, Supplementary
 Figures and Tables.
- **233 Data Availability:** Pollen load data is available as Supplementary Table S9.
- 234235 Code Availability: Simulation code and simulation data are available at the repository:
- https://github.com/fsvaldovinos/Phenology. Phenology parameters used in the simulation portion of our
 study are available in Table S10.

239 References

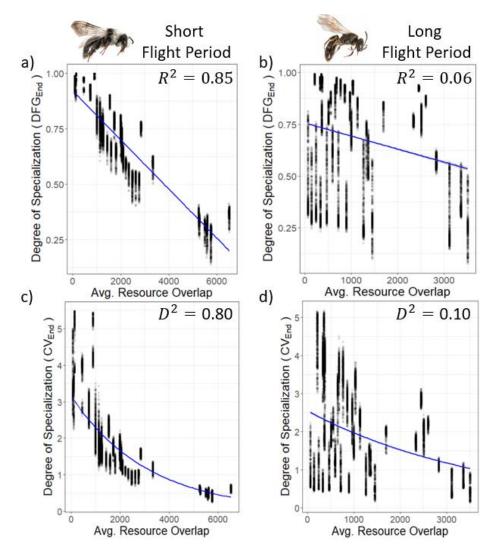
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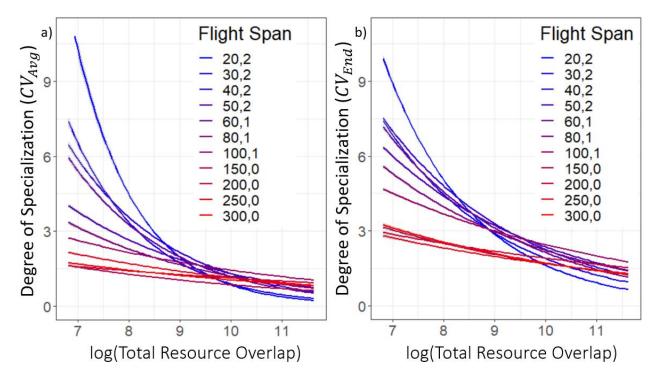


351 populations create low levels of temporal overlap amongst different flowering resources. On the other 352 hand, fewer seasonal delimitations between different flowering populations create higher levels of 353 temporal overlap amongst different flowering resources. a) In long flight-period pollinators (e.g. 354 Lasioglossum), differing levels of temporal resource overlap have limited to no effect on the degree of 355 diet options available to the pollinator. b) This is predicted to lead to a non-significant relationship 356 between overlap and specialization in long flight period pollinators. c) In short flight-period pollinators (e.g. Andrena), low and high temporal resource overlap can limit or expand pollinator diet options 357 358 respectively. d) This is predicted to lead to a significant relationship between overlap and specialization in 359 short flight-period pollinators.



361 362

363 Figure 2: Temporal resource overlap drives specialization more strongly in short-lived than in long-364 lived pollinators. Model output on pollinator specialization taken across average plant overlap (ARO) for 365 short flight period (flight period = 20 time steps) and long flight period pollinators (flight period = 300 366 time steps). Each dot represents the diet breadth metric of a single pollinator population based on their 367 foraging effort per plant species. Diet breadth was measured at the end of simulations using a) DFG in a 368 short flight period bee, F-statistic: 1.05e+05 on 1 and 18498 DF, p-value: < 2.2e-16, b) DFG in a long flight period bee, F-statistic: 1089 on 1 and 18498 DF, p-value: < 2.2e-16, c) CV in a short flight period 369 370 bee, Residual deviance: 671.94 on 18498 degrees of freedom, d) CV in a long flight period bee, Residual 371 deviance: 8244.4 on 18498 degrees of freedom. Example data taken from a network framework with 30 372 plant and 50 pollinator species. Higher values of both DFG and CV metrics indicate higher levels of 373 specialization (narrower diet breadth) while lower values indicate a more generalized feeding range 374 (wider diet breadth).





378 Figure 3: Interactive effect of temporal resource overlap and pollinator flight period on pollinator 379 diet breadth. Representative sample of generalized linear models (Gamma distribution with log link 380 functions) showing change in effect of Total Resource Overlap (TRO) on pollinator diet breadth in model 381 simulations. Flight period labels indicate the length of the flight period and the number of phenology 382 cycles between subsequent flight periods (see Supplementary Methods). Diet breadth shown as the degree 383 of specialization measured by the CV of pollinator foraging effort per plant species taken \mathbf{a}) at the end of 384 simulations and **b**) averaged across the last 1000 time steps. Higher CV values correspond with a higher 385 level of specialization (narrower diet breadth) while lower CV values indicate a more generalized feeding 386 range (wider diet breadth). Example data taken from a network framework with 30 plant and 50 pollinator 387 species. 388

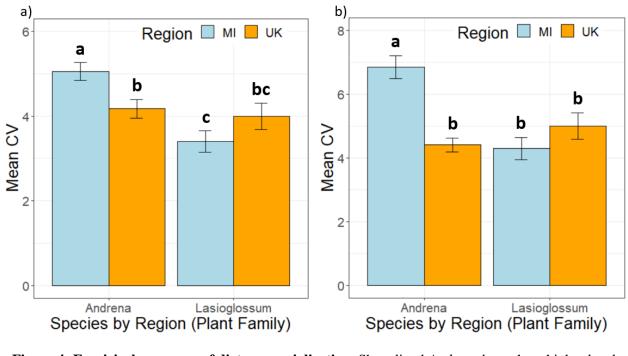


Figure 4: Empirical measures of dietary specialization. Short-lived Andrena bees show higher levels
 of specialization in highly seasonal (Michigan) compared to the weakly seasonal (UK), with no
 differences for long-lived Lasioglossum bees using the CV metric. Moreover, only in Michigan is there a
 difference in specialization between the two bee genera, showing a minimal impact of flight period in
 weakly seasonal environments. Effects are consistent at both the (a) botanical family and (b) botanical
 genus level. Bars headed by different letters are significantly different (p<0.05). Error bars show standard

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