Ecology Letters

Ecology Letters, (2020) 23: 1107-1116

doi: 10.1111/ele.13510

LETTER

Trait matching and phenological overlap increase the spatiotemporal stability and functionality of plant-pollinator interactions

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The peer review history for this article is available at https://publons. com/publon/10.1111/ele.13510

Abstract

Morphology and phenology influence plant–pollinator network structure, but whether they generate more stable pairwise interactions with higher pollination success remains unknown. Here we evaluate the importance of morphological trait matching, phenological overlap and specialisation for the spatio-temporal stability (measured as variability) of plant–pollinator interactions and for pollination success, while controlling for species' abundance. To this end, we combined a 6-year plant–pollinator interaction dataset, with information on species traits, phenologies, specialisation, abundance and pollination success, into structural equation models. Interactions among abundant plants and pollinators with well-matched traits and phenologies formed the stable and functional backbone of the pollination network, whereas poorly matched interactions were variable in time and had lower pollination success. We conclude that phenological overlap could be more useful for predicting changes in species interactions than species abundances, and that non-random extinction of species with well-matched traits could decrease the stability of interactions within communities and reduce their functioning.

Keywords

Abundance, interaction frequency, pollination success, pollinator impact, spatial variability, specialisation, temporal variability.

Ecology Letters (2020) 23: 1107-1116

INTRODUCTION

Understanding the mechanisms that determine species interactions and interaction network patterns is critical for predicting changes in community structure and ecosystem function. Research on networks of species interactions has moved from assessing static interaction patterns towards evaluating the temporal variability of species interactions (Trøjelsgaard & Olesen 2016). For instance, recent studies on plant-pollinator interactions show that interactions frequently vary across space and time (Petanidou et al. 2008; Trøjelsgaard et al. 2015), and that this variability is underpinned by species turnover and interaction rewiring (Trøjelsgaard et al. 2015; Burkle et al. 2016; Rabeling et al. 2019). However, the mechanisms that make individual plant-pollinator interactions more or less persistent in space and time (i.e. what makes an interaction more stable), and whether the mechanisms determining species interaction frequencies are also responsible for determining the stability of interactions, remain uncertain.

Among the drivers of plant–pollinator interactions, species phenology, morphology and abundance have extremely important roles (Stang *et al.* 2007; Krishna *et al.* 2008; Vázquez *et al.* 2009b). Abundant species, for example, are

more likely to randomly encounter each other and interact (Canard et al. 2012, 2014), hence increasing species interaction probability among potentially interacting species. However, if species are abundant at different times (i.e. species have nonoverlapping phenologies), then the likelihood of them interacting is likely to decline. Therefore, the influence of species abundance on interactions may be better represented by the phenological overlap of the interacting species (i.e. the amount of time interacting species co-occur), rather than total abundance. Phenological overlap is not just a result of species abundances (abundant species have longer phenologies and hence increase the chances of phenological overlap, Gillespie et al. 2017), but also depends on species characteristics (i.e. the timing of phenological stages, Pleasants 1990), and on the interaction between species characteristics and abundance (i.e. changes in the timing of phenological stages induced by changes in species abundance, Wolf et al. 2017). Phenological overlap can, therefore, be an extremely important predictor of the occurrence and frequency of species interactions, but could also play an important role in the temporal and spatial dynamics of species interactions. For instance, if a plant and a pollinator species have highly overlapping phenologies across multiple years, it will be more likely that the

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interaction between those two species will occur in multiple years (i.e. will be more stable across time), compared with two interacting species whose phenologies have little overlap (Fig. 1a, b and e).

Phenological overlap between plant and pollinator species can persist over long periods (evolutionary time) and, by allowing species to consistently interact more strongly, it can facilitate evolution of trait matching among interacting species (Lomáscolo et al. 2019) (Fig. 1b and c). Furthermore, it has been suggested that trait matching might occur or coevolve to improve efficiency in the removal and deposition of pollen and/or as a way to favour the acquisition of greater rewards. For instance, if a pollinator obtains greater feeding rewards from flowers with traits matching the pollinator's, then the pollinator might preferentially visit that plant species and thus increase the interaction frequency between those species (Klumpers et al. 2019). Therefore, trait matching is an important predictor of network architecture and species interactions (Santamaría & Rodríguez-Gironés 2007; Stang et al. 2009). In addition, networks may contain a persistent (i.e. occur consistently through time and space) core of well-matched high-reward coevolved interactions, with additional interactions between poorly matched species being more opportunistic and easily disrupted (Tylianakis & Morris 2017; Chacoff et al. 2018), and, hence, more variable (Fig. 1c and e).

The evolution of trait matching among plants and pollinators seems to be affected by species' trophic specialisation and the asymmetries in specialisation among interacting species (Anderson *et al.* 2010; Klumpers *et al.* 2019). For example, if a pollinator forages on different plant species, it may be less likely to match a particular focal plant species than if it were foraging on that species alone (Anderson *et al.* 2010). Hence, specialisation of both interacting partners might indirectly increase the spatial and temporal stability of interactions by promoting trait matching (Fig. 1d, c and e). However, these specialised interactions may be more vulnerable to extinction (Aizen *et al.* 2012), which would make them less stable. Nevertheless, assessing the effect of specialisation on interaction stability requires controlling for the effects of species abundances (Winfree *et al.* 2014), as rare (less abundant) species could present apparent specialisation (Dorado *et al.* 2011).

Trait matching could also play an important role in the functional outcome of plant-pollinator interactions, such as pollination success. For instance, although imperfect trait matching does not imply a complete functional mismatch (Memmott *et al.* 2004; Solís-Montero & Vallejo-Marín 2017), strong trait matching between interacting plant and pollinator species increases pollinator foraging efficiency (Klumpers *et al.* 2019), and can favour fruit set in crop plants (Garibaldi *et al.* 2015). Furthermore, pollination success relies on *per capita* (i.e. per-visit) efficacy of each pollinator species on plant reproduction and interaction frequency (Vázquez *et al.* 2012), both of which could be influenced by trait matching (Klumpers *et al.* 2019). Specifically, if increased trait matching results in a

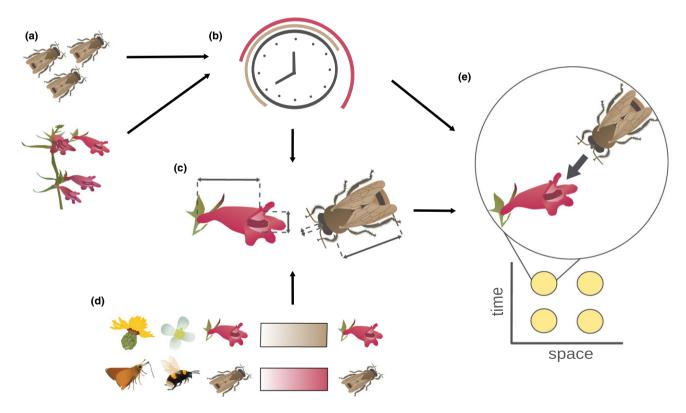


Figure 1 Conceptual figure showing the potential determinants of plant–pollinator interaction frequency and the spatial and temporal variability of interactions. (a) Abundance of a plant species and a pollinator species; (b) plant–pollinator phenological overlap; (c) plant and pollinator morphological traits, used to estimate trait matching between species; (d) plant and pollinator species specialisation; (e) plant–pollinator interaction frequency and spatio-temporal variability of interactions. Filled circles in (e) represent sampling sites and the arrow between plant and pollinator species represent their interaction frequency.

more efficient interaction (e.g. more efficient pollen deposition) (Nagano et al. 2014; Solís-Montero & Vallejo-Marín 2017), it would increase pollination success via the per capita effect of pollinator species on pollination success. In addition, specialisation of interacting species could strengthen trait matching (and vice versa) via coevolution, thereby increasing interaction frequency and, indirectly, pollination success. Similarly, because phenological overlap influences the frequency of interactions among species, it could indirectly affect interaction-dependent functions, such as pollination success.

Here we use temporal and spatial data on plant and pollinator interactions, traits and pollination success to evaluate the relevance of trait matching and phenological overlap for the spatio-temporal stability (measured as the variability of interaction frequency) and functionality (measured as pollination success, i.e. the number of pollen tubes growing after pollen deposition) of plant-pollinator interactions. Although morphological trait matching among plant and pollinator species has been usually evaluated as a univariate metric (Solís-Montero & Vallejo-Marín 2017; Klumpers et al. 2019), for example the match (or mismatch) of the pollinator proboscis length with the floral nectar tube length, species interactions depend on the matching of multiple non-independent traits (Vázquez et al. 2009a). Moreover, multiple pollinator traits are known to be important for pollination success (Hoehn et al. 2008), such that a trait-interaction-function framework will likely need to incorporate multiple traits. Therefore, we assessed the congruence of multiple traits simultaneously (in multivariate space). Furthermore, because species abundance is an important determinant of interaction frequencies (Vázquez et al. 2009b), we incorporated plant and pollinator species abundances in our analyses to assess the effects of trait matching and phenological overlap after controlling for abundance. We hypothesise that: (1) interaction frequency increases with increasing trait matching and phenological overlap between the interacting species; (2) spatio-temporal stability of interactions increases with increasing trait matching and phenological overlap between the interacting species; (3) trait matching between interacting species increases with specialisation, hence increased specialisation indirectly increases the spatio-temporal stability of interactions; (4) pollination success increases with increasing trait matching and phenological overlap between the interacting species.

METHODS

Plant-pollinator visitation networks

During 2006–2011 plant–pollinator interactions were recorded in the Monte Desert of Villavicencio Nature Reserve (Mendoza, Argentina, 32°32′ S, 68°57′ W). The maximum number of sites sampled per year was four (each of 1-ha area) and they were separated by at least 1 km. Vegetation across all sites was c. 2 m tall shrubland, dominated by Larrea divaricata, Condalia microphylla, Prosopis flexuosa and Opuntia sulphurea (further study site and sampling methodology details can be found in Chacoff et al. 2012). In 2006 we sampled four sites (sites 1, 2, 3 and 4), in 2007 three sites (sites 1, 2 and 4) and from 2008 to 2011 we sampled two sites (sites 1 and 4).

We pooled all interactions sampled across sites per year to obtain a plant–pollinator visitation network per year. We combined the 6 years of visitation data to obtain a single plant–pollinator visitation network, and used the total number of flower visits by each pollinator species across the 6 years to represent interaction frequency.

Although the number of sites sampled and sampling effort differed across years, the percentage of links observed (i.e. connectance), interaction evenness, network specialisation and the percent of observed links relative to the estimated total links (estimated by the Chao2 statistic) were not significantly related to sampling effort (Chacoff *et al.* 2018). In addition, the species interaction composition (i.e. the identity and frequency of the pairwise interactions forming the network) did not differ significantly across years, and did not depend on the number of sites sampled each year (Supporting Information, Fig. S1).

Species abundance

To estimate the abundance of flowers in the sites where species interactions were sampled, we counted the number of flowers per plant species per site using transects and quadrats. In 2006 we sampled forty 2×2 m quadrats per site (640 m² total area), in 2007 five 50×2 m transects (500 m² total), and in 2008–2011 two 50×2 m transects and four 20×8 m quadrats (840 m² total per year). We summed the abundance of each flower species across years to obtain an overall measure of flower abundance per plant species: number of flowers per species over the entire sampling period (i.e. 4500 m² sampled over the 6 years).

Because empirical measures of pollinator abundance (measured independently of interactions) were not available, we estimated them from quantitative interaction data using the mass action principle (Staniczenko et al. 2013, 2017). The mass action principle originated in chemistry, and states that the rate of a chemical reaction is proportional to the product of the concentration or mass of the reactants. From an ecological perspective, it is assumed that the frequency of interactions (F_{ii}) between species is proportional to the product of the species relative abundances (x_i = pollinator species i abundance; x_i = plant species *j* abundance) due to random encounters, but also depends on the interaction preferences between interacting species (γ_{ij}) , that is $F_{ij} = x_i x_j \gamma_{ij}$. Therefore, when sufficient species interaction frequencies (F_{ij}) data are available (such as in our dataset), it is possible to infer interaction preferences (γ_{ii}) and then the effective relative abundance of the interacting partners (x_i and x_i) (Staniczenko *et al.* 2013).

Species traits and trait matching. We selected plant and pollinator morphological traits that are considered important for the occurrence of plant–pollinator interactions (Eklöf et al. 2013; Junker et al. 2013; Coux et al. 2016; Johnson & Raguso 2016). For plant species, we measured corolla length, corolla aperture and mean flower height (see Supporting Information for more details). We sampled between 3 and 24 flowers per species and calculated the mean value of each trait per species. For pollinators, we measured proboscis length and width, body width, body thickness and body length (see Supporting Information for explanation of how plant and

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pollinator traits are related, i.e. matched). We measured up to 10 individuals per species and used the mean value of each trait per species; details of the measured traits can be found in Lomáscolo *et al.* (2019).

We measured traits from 45 (of 59; 76.3%) plant species and 135 (of 196; 68.9%) pollinator species present in the interaction networks. Because some specimens were missing from the collection, we could not measure traits from all species. Therefore, we used only the measured subset of species (and their interactions) in our analyses. This subset represents 73% of the species sampled in the interaction networks, 84% of the binary links (i.e. 881 links) and 97% of all interaction events recorded (i.e. 27 074 interactions). Although species not included in our analyses were rare, their specialisation did not differ systematically from that of the species in our analyses (Supporting Information, Fig. S2). In addition, although the phenology of the species excluded from our analyses tended to be short (between 1-13 weeks and 1-8 weeks for plants and pollinators respectively), plant and pollinator species with similar phenology duration were included in our analyses (i.e. species with short phenologies were represented in the analyses).

As discussed earlier, the linking of traits, interactions and functioning likely requires the inclusion of multiple non-independent traits. For this reason, we used a multivariate trait congruence metric to quantify trait matching, which also accounts for the correlation between traits that occurs within species (Cariveau et al. 2016; Melin et al. 2019). First we calculated similarity in morphological traits between each pair of plant species (i.e. a plant species trait similarity matrix) and between each pair of pollinator species (i.e. a pollinator species trait similarity matrix) using Gower's similarity coefficient (Gower 1971) and the vegdist function from the vegan v 2.5-5 R package (Oksanen et al. 2019). Then, we performed a Parafit test (Legendre et al. 2002) using the plant and pollinator trait similarity matrices and the binary species interaction matrix, using the parafit function from the ape v 5.1 R package (Paradis et al. 2004). The Parafit test quantifies the congruence between two topological trees (one for plants and one for pollinators) defined by the species trait similarity matrices. Parafit tests the null hypothesis that species interact randomly with respect to trait differences at each trophic level, that is that pollinator species with similar traits do not interact with plant species with similar traits more often than expected at random. If the null hypothesis is rejected, that is pollinator species with similar traits interact with plant species with similar traits, there is significant congruence between plant and pollinator traits. Given significant congruence, the contribution of each interaction to this pattern (i.e. how strongly pollinator traits predict their interactions with plants based on their traits) can be estimated. A large contribution to the community trait-matching congruence pattern (as measured by the statistic ParafitLink1) therefore provides strong evidence of trait matching for that interaction.

Pollination success. To estimate pollination success, we used data on pollen deposition for four of the most abundant plant species, Larrea divaricata and L. nitida (Zygophyllaceae), Opuntia sulphurea (Cactaceae) and Zuccagnia punctata (Fabaceae), from a previous study (Vázquez et al. 2012).

Pollen deposition on these four plant species was sampled in 2008, at the same sites as plant-pollinator interactions were sampled. Two of these plant species had many flower visitor species (L. divaricata had 52 and Z. punctata 56 visitor species), whereas the others had fewer flower visitors (L. nitida had 24 and O. sulphurea 33). We measured pollen deposition by three to seven pollinator species for each plant species (10 pollinator species, 19 different interactions [i.e. links]). We acknowledge that by focusing on plant species with intermediate and large number of flower visitor species we are omitting more specialised plant species. Nevertheless, the plant-pollinator interactions for which we measured pollen deposition span almost the entire range of phenological overlap, trait matching and interaction frequency across all the interactions sampled during the 6 years (Supporting Information, Fig. S3).

Pollination success was estimated as the impact that each pollinator species has on plant reproduction relative to the impact of all pollinator species (Vázquez et al. 2012). Pollinator species impact depends on the per capita pollinator's effect (per-visit effect) and the frequency of interaction between the plant and pollinator species (Vázquez et al. 2005; Wootton & Emmerson 2005). Therefore, a pollinator species' impact on plant reproduction (t_{ij}) was calculated as $t_{ij} = P_{ij} I_{ij} / T_{i.}$, where P_{ij} is the per capita pollinator's effect; I_{ij} is the frequency of interaction (number of visits) of pollinator species j to plant species i across the entire sampling period (6 years); and T_i the impact of all pollinator species to plant j reproductive success $(T_{i.} = \sum I_{ii} |P_{ii}|)$. As an estimate of the per capita pollinator's effect (P_{ii}) , we counted the number of pollen tubes growing below the tip of the pistil (as a measure of plant reproductive success) resulting from pollen deposited by each pollinator species on the stigma of each plant species in one visit, relative to the pollen deposited by all pollinator species (Vázquez et al. 2012).

STATISTICAL ANALYSES

To assess the effects of trait matching and phenological overlap on interaction frequency, pollination success, and the temporal and spatial variability of interactions, after controlling for species abundance effects, we used generalised multilevel path models (Shipley 2009), a form of structural equation model (SEM) fitted through separate component models. Each SEM used to assess changes in interaction frequencies, or the temporal and spatial variability of interactions, consisted of three linear mixed effects models (LMMs). The first LMM consisted of plant and pollinator species abundances, trait matching and phenological overlap as predictor variables and interaction frequencies, temporal variability of interactions or spatial variability of interactions as response variables on each of the three SEMs respectively. The second LMM in each SEM included phenological overlap as a response variable and plant and pollinator species abundances as predictor variables. The third LMM included trait matching as a response variable and phenological overlap and mean specialisation of the interacting partners (d') as predictors. All LMMs included plant and pollinator identities as random factors to account for the non-independence of interactions sharing one partner. All SEMs included the relationship between mean specialisation of the interacting partners (d'), plant species abundances, pollinator species abundances and interaction frequency (or temporal or spatial variability of interactions) as correlated errors, as these relationships were not assumed to be causal and unidirectional (for visual interpretation see Figs. 2 and 3). See Supporting Information for a figure showing the model with justifications for each path.

Because the temporal and spatial variability of interactions was estimated as the coefficient of variation of interaction frequency, higher interaction frequencies could lead to higher mean interaction frequencies and hence lower coefficients of variation. Therefore, to assess whether the effects of trait matching and phenological overlap on the temporal and spatial variability of interactions were significant beyond differences in interaction frequencies, we conducted two additional LMMs. We used the temporal variability of interactions and the spatial variability of interactions as response variables (one on each model) and plant and pollinator species abundance, trait matching and phenological overlap as predictor variables in these two models. We included interaction frequency as a covariate in both models to account for differences in the frequency of interactions among different plant-pollinator interactions, which also represents a way to control for the potential effects of differences in sampling effort. Plant and pollinator identities were included in these models as random factors.

To determine the effects of trait matching and phenological overlap on pollination success, we used a SEM formed by three LMMs with the same random factors as the previous LMMs. The first LMM included interaction frequency as a response variable and trait matching and phenological overlap as predictor variables. The second LMM included the per capita pollinator's effect on plant reproduction (i.e. number of pollen tubes formed after a pollinator's visit, P_{ij}) as the response variable and trait matching as the predictor variable. The third LMM included the pollinators' total effect on plant reproduction (t_{ii}) as the response variable and interaction frequency and the per capita pollinator effect as predictor variables. We did not include abundance in this model because we had no reason to hypothesise that per capita effect would depend on abundance. Instead, we believe that abundance would only influence total pollination effect indirectly by influencing the frequency of interactions (which was included in our model). This SEM was estimated with a subset of our dataset that included interactions formed by pollinator species for which we had values of their impact on plant reproduction, and those plant species for which we had measures of reproductive success.

Variables used in all SEMs were scaled (to make their effect sizes comparable) by subtracting the mean and dividing by the s.d., and a Gaussian error distribution was used in all linear mixed effects models (normality and homoscedasticity assumptions were checked). All statistical analyses were performed in the R environment v 3.6.1 (R Core Team 2019). We used the lme4 v 1.1-21 package (Bates *et al.* 2015) for the LMMs and piecewiseSEM v 2.0.2 package (Lefcheck 2016) to fit the SEMs. We evaluated model fit by performing a d-separation test of each SEM (Shipley 2009). The d-separation test estimates a Fisher's *C* statistic and performs a chi-squared test

on it; when the P-value of the chi-squared test is greater than 0.05 it is considered that the hypothesised relationships are consistent with the data, that is the model is a good fit to the data. All models were good fits to the data.

SPECIES SPECIALISATION AND PHENOLOGICAL OVERLAP

We calculated plant and pollinator species specialisation from the quantitative interaction matrix (i.e. data from all sites and years combined) as d' (Blüthgen et al. 2006), with the specieslevel function in the bipartite v 2.13 R package (Dormann et al. 2008). This specialisation metric can be interpreted as a measure of selectivity relative to available resources, as it incorporates the difference in availability of the different partners and estimates how strongly a species deviates from a random sampling of available interaction partners. It ranges from 0 (no selectivity, i.e. most generalist) to 1 (high selectivity, i.e. most specialist). Low levels of specialisation suggest species with opportunistic partner selection. When alternative specialisation metrics (normalised degree, i.e. the normalised number of interacting partners, and Paired Difference Index, which contrasts a species' strongest link to a resource with those over all remaining resources, Poisot et al. 2012) were used instead of d', the results did not change qualitatively (Tables S1-S6, S8-S10), although our structural equation models (described below) fit better when using d'.

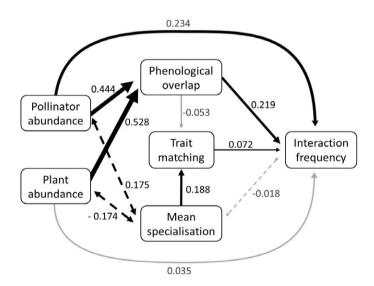


Figure 2 Structural equation model evaluating the effects of trait matching and phenological overlap on plant–pollinator interaction frequency. Specialisation was measured as d' (although using different measurements of specialisation such as normalised degree and PDI did not affect unidirectional relationships, see Supporting Information). Solid black arrows represent unidirectional statistically significant relationships (P < 0.05), grey arrows represent non-significant relationships and dashed arrows represent correlated errors. The thickness of the arrows is scaled to standardised coefficients from SEM and illustrates the relative effect size. Standardised path coefficients are shown next to each arrow. The strength of the indirect effects is calculated by the product of the coefficients along the path. For example pollinator abundance has a direct effect of 0.234 and an indirect effect, through phenological overlap, of 0.097 (0.444 \times 0.219) on interaction frequency. The model was a good fit to the data (Fisher C statistic = 4.644, d.f. = 6, P = 0.59).

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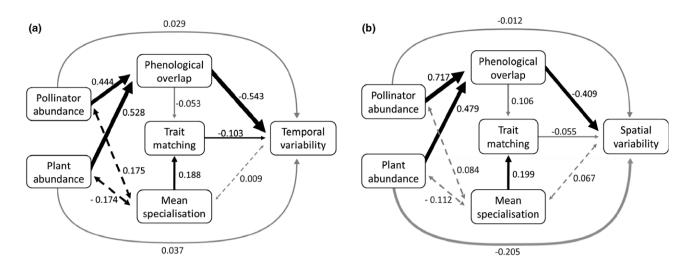


Figure 3 Structural equation models evaluating the effects of trait matching and phenological overlap on plant–pollinator interaction temporal (a) and spatial (b) variability. Specialisation was measured as d'. Other conventions as in Fig. 2. Both models were a good fit to the data (temporal variability model: Fisher C statistic = 4.644, d.f. = 6, P = 0.59; spatial variability model: Fisher C statistic = 5.714, d.f. = 6, P = 0.456)

We calculated the mean specialisation of each pair of interacting plant and pollinator species and refer, hereafter, to this variable as mean specialisation of the interacting partners. We also estimated phenological overlap between any pair of interacting plant and pollinator species as the number of weeks (across the entire 6-year sampling period) when these species co-occur.

SPATIO-TEMPORAL STABILITY OF PLANT-POLLINATOR INTERACTIONS

As a measure of the spatial and temporal stability of plant-pollinator interactions, we used the coefficient of variation of interaction frequency (lower coefficients of variation indicate higher stability of interactions). The temporal coefficient of variation of interactions was obtained by dividing the standard deviation (SD) of the frequency of each interaction sampled across the 6 years by its mean. Similarly, the spatial coefficient of variation of interactions was obtained by dividing the SD of the frequency of each interaction sampled across the four sites sampled in 2006 (year in which the highest number of sites were sampled) by the mean interaction frequency of those interactions sampled on those sites that year. The spatial stability was estimated only for those interactions that occurred at least once in 2006 (i.e. 150 links).

RESULTS

High trait matching and phenological overlap made interactions more frequent and stable, and indirectly improved pollination success. Interaction frequency was positively and directly affected by trait matching, phenological overlap and pollinator abundance, and indirectly by the mean specialisation of the interacting partners (d') and the interacting partners' abundances (Fig. 2, Table S1). The importance of abundance was emphasised by the direct effect of pollinator abundance on interaction frequency and of plant and pollinator abundance

on phenological overlap (which, in turn, influenced interaction frequency). Nevertheless, because the method we used to estimate pollinator abundance involved interaction frequencies, there could be some correlation between pollinator abundance and interaction frequency and hence we remain cautious about this particular result (i.e. the direct effect of pollinator abundance on interaction frequency).

Trait matching and phenological overlap independently stabilised mutualistic interactions in time (i.e. they decreased the temporal variability of interactions; Fig. 3a, Table S4), even after controlling for the effects of interaction frequency on interaction variability (Table S7). The temporal variability of interactions was indirectly negatively affected by the abundance of plants and pollinators and the mean specialisation (d') of the interacting partners (Fig. 3a, Table S4). On the other hand, the spatial variability of interactions was only directly affected by phenological overlap, and indirectly negatively affected by the interacting species' abundance (Fig. 3b, Table S8).

The impact of pollinator species on plant reproduction was directly affected by both the *per capita* pollinator effect and interaction frequency, because both variables were used to calculate pollinator impact. More importantly, interactions in which pollinators and plants had matching traits and overlapping phenologies were more frequent, and therefore associated with stronger pollinator impacts (t_{ij}) , even though the pollinator's *per capita* effect was not significantly affected by trait matching (Fig. 4, Table S11). Combined with the previous SEM showing the effects of abundance and mean specialisation on interaction frequency, we can conclude that these variables also indirectly influence pollination impact via interaction frequency.

DISCUSSION

Morphology, phenology and abundance have been identified as the main drivers of plant-pollinator interaction network structure (Vázquez et al. 2009b; Olesen et al. 2011; Encinas-

Viso et al. 2012; Morente-López et al. 2018), but their effects on functionality and stability of interactions (measured as interaction variability) were previously unresolved. Here, we demonstrate that plant-pollinator interactions among species with well-matched traits and higher phenological overlap are stable in time and increase pollination success by affecting interaction frequency, independent of the effect of plant or pollinator abundance. In addition, interactions among specialised species had higher trait matching, such that specialisation indirectly favours interaction stability. These results suggest that the non-random extinction of interactions among specialists (Aizen et al. 2012) and biotic invasion (with invaders typically being generalists, Aizen et al. 2008, Frost et al. 2019, and promoters of interaction rewiring, Arroyo-Correa et al. 2020), can disproportionately increase the variability of interaction frequencies within communities, increasing ecological network variability and making more difficult the prediction of interaction networks. Furthermore, we show that the strong positive effects of species abundance on the spatio-temporal stability of interactions are mediated by phenological overlap, such that temporal co-occurrence, rather than abundance itself, appears to be stabilising.

The relevance of phenological overlap for species coexistence (Rudolf 2019) and for species interaction variability, as shown here, reflects the importance of phenologies for community dynamics and for determining the robustness of communities to environmental change (Encinas-Viso et al. 2012; Ramos-Jiliberto et al. 2018). Furthermore, phenological shifts (mismatches) induced by global environmental changes have been identified as potential drivers of the extinction of plantpollinator interactions (Memmott et al. 2007), which could be initiated by a decrease in interaction frequency, as shown here. Phenological overlap and abundance explain a substantial fraction of species interactions in the system we consider (Vázquez et al. 2009b). Furthermore, as we show here, species abundance effects on the spatio-temporal stability of interactions were completely mediated by phenological overlap. Therefore, the quantification of phenological overlap among interacting species could be even more effective than species abundance data for predicting changes in species interactions. The key role of phenological overlap suggests that measuring phenological overlap, that is species temporal co-occurrence,

could be an effective and simpler alternative to measuring the abundance of each species in the community.

Although phenological overlap had direct effects on interaction frequency and the stability of interactions, it did not favour trait matching among interacting partners as we had expected. Despite using a 6-year dataset to estimate phenological overlap, it is likely that a longer time span needs to be considered in the estimation of phenological overlap to capture coevolutionary changes in species morphology (Thompson 1998).

Morphological traits and rewards influence the array of flower visitors a plant receives (Bosch et al. 1997; Stang et al. 2006), making trait matching between plants and pollinators a key driver of the occurrence and frequency of interactions among these partners. Even though trait matching is an important determinant of plant-pollinator interactions, to our knowledge, its contribution to the stability of interactions has been overlooked. Previous studies have explained spatial and temporal variability of plant-pollinator interactions in terms of changes in species composition (species turnover) and interaction rewiring (Carstensen et al. 2014; Trøjelsgaard et al. 2015; CaraDonna et al. 2017), but without including species morphological dependencies. Here, we took a step forward and showed that trait matching influences not only the temporal stability (i.e. inverse of temporal variability) of interaction frequency, but also the stability of pollination success. Although trait matching had no significant effect on the spatial variability of plant-pollinator interactions, it is possible that the smaller sample size used in this particular analysis (we only used interactions sampled in 2006 for which we had spatial variability data) did not allow us to detect such an effect. It is also possible that, because of the limited spatial extent of our study, not enough variability in interactions was captured across space (spatial variability was lower than temporal variability of interactions) to detect the effects of trait matching on interaction spatial variability.

Specialisation of both interacting partners indirectly increased the temporal stability of interactions by promoting trait matching. The idea that interactions among specialist species are more temporally stable, at least in a 6-year period, could be somewhat counter-intuitive, as previous studies have shown that specialised species (Burkle *et al.* 2013) and their interactions (Aizen *et al.* 2012) are the most sensitive to

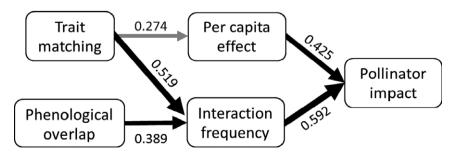


Figure 4 Structural equation model evaluating the effects of trait matching and phenological overlap on each pollinator's impact on plant reproduction (t_{ij}) , as an estimate of pollination success. Pollination success is hence determined by *per capita* pollinator species effects and interaction frequency effects. *Per capita* (per visit) pollinator effects were estimated by measuring plant reproductive success (number of pollen tubes growing from pollen deposited by each pollinator species on the stigma of each plant species in one visit, relative to the pollen deposited by all pollinator species). Other conventions as in Fig. 2. The model was a good fit to the data (Fisher C statistic = 11.875, d.f. = 8, P = 0.157)

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habitat changes, and to the loss of climatic suitability (Schleuning et al. 2016). However, specialised interactions (i.e. those among species with few interacting partners) could be stable and fragile (i.e. vulnerable to rapid change) at the same time. In part this discrepancy may occur because the characteristics of the species that succeed within a given environment could differ from those that succeed in a changing environment. Moreover, studies forecasting the susceptibility of specialised interactions do not consider the effects of species abundance, which suggests that the survival of the most abundant species could confound the effects of species specialisation (Winfree et al. 2014). This difficulty in teasing apart the effects of abundance vs. interacting partner specialisation raises the question of whether the persistent core of plant-pollinator networks is formed by interactions among generalist species (Bascompte et al. 2003; Bascompte & Jordano 2007) or by the most abundant species that tend to show lower specialisation (Fort et al. 2016). For instance Winfree et al. (2014) showed that once the positive effect of abundance is accounted for, specialised species are more persistent than generalist species. This trend is clearly observed in our results, where the indirect effects of plant and pollinator abundance on the temporal stability of interactions is more than 13-times larger than the indirect effects of specialisation on temporal stability. Therefore, once the effects of abundance are separated from specialisation, the positive indirect effect of specialisation on interaction stability becomes apparent.

Previous studies have shown that trait matching can influence the accuracy of pollen deposition by pollinators (Poblete Palacios et al. 2019) or the amount of pollen deposited (Nagano et al. 2014; Solís-Montero & Vallejo-Marín 2017), potentially favouring plant reproduction. However, our results show that interactions among species with matching traits had higher pollination success because of an increase in interaction frequency, but not due to an increase in the pollinator's per capita effect on plant reproduction. It is possible that only some of the traits considered here would be extremely important for influencing the pollinator's per capita effect on plant reproduction, and using a multivariate metric of trait matching (i.e. simultaneously considering the multiple traits that determine interactions) dilutes the positive relationship between those specific traits and pollinator's per capita effect. Another potential explanation for the lack of relationship between trait matching and a pollinator's per capita effect is that we assessed pollination success only for generalist plant species. It is possible that including specialised interactions (i.e. interactions among specialised plants and specialised pollinators, with a stronger trait matching signal) could strengthen the trait matching - pollinator's per capita effect relationship. Finally, pollination success increased with phenological overlap between interacting species, which supports the idea that phenological mismatches affect species reproductive success and survival (Visser & Gienapp 2019).

Plant and pollinator morphology and phenology are important determinants of their interaction network structure and extinction risk (Stang *et al.* 2007; Vázquez *et al.* 2009b; Morente-López *et al.* 2018). In this study we demonstrate that interactions with well-matched traits and phenologies are temporally and spatially more stable. Moreover, the effects of

species abundance on interactions and their stability are largely mediated by phenological overlap, and the importance of phenological overlap for the stability of plant–pollinator interactions largely surpasses the effects of trait matching. Furthermore, our results reinforce the idea that trait matching (Garibaldi *et al.* 2015) and phenological overlap are extremely important for pollination success by increasing species interaction frequencies. Therefore, the systematic loss of well-matched interactions, alongside the proliferation of less well-matched interactions (e.g. invasive species), and the phenological shifts and mismatches induced by global environmental changes (Memmott *et al.* 2007), could alter interaction stability and ecosystem functions, and further impair our ability to predict changes in species interactions.

ACKNOWLEDGEMENTS

We thank P. Staniczenko for kindly sharing code for species abundance estimation; D.B. Stouffer, the Poisot lab group, R. Irwin and two anonymous reviewers for useful comments and M. Walters and L. Pereira Martins for assistance with Figure 1 (symbols courtesy of the Integration and Application Network, University of Maryland Center for Environmental Science, ian.umces.edu/symbols/). This work was funded by the Marsden Fund (grant number UOC1705), administered by the Royal Society of New Zealand.

COMPETING INTERESTS

The authors declare no competing interests.

AUTHORSHIP

JMT, DPV and GLWP conceived the idea and secured funding. GP and JMT further developed the idea. DPV, NPC and SBL collected data. GP performed analyses and wrote the first draft of the manuscript. All authors contributed substantially to revisions.

DATA AVAILABILITY STATEMENT

Chacoff, N.P., Vazquez, D.P, Lomáscolo, S.B. Data from: Trait matching and phenological overlap increase the spatio-temporal stability and functionality of plant-pollinator interactions. Dryad Digital Repository, https://datadryad.org/stash/dataset/doi:10.5061/dryad.8cz8w9gm1.

REFERENCES

Aizen, M., Morales, C. & Morales, J. (2008). Invasive mutualists erode native pollination webs. *PLoS Biol.*, 6(2), e31.

Aizen, M.A., Sabatino, M. & Tylianakis, J.M. (2012). Specialization and rarity predict nonrandom loss of interactions from mutualist networks. *Science*, 335, 1486–1489.

Anderson, B., Terblanche, J.S. & Ellis, A.G. (2010). Predictable patterns of trait mismatches between interacting plants and insects. *BMC Evol. Biol.*, 10, 204.

Arroyo-Correa, B., Burkle, L.A. & Emer, C. (2020). Alien plants and flower visitors disrupt the seasonal dynamics of mutualistic networks. *J. Ecol.*, in press (accepted).

- Bascompte, J. & Jordano, P. (2007). Plant-animal mutualistic networks: the architecture of biodiversity. Annu. Rev. Ecol. Evol. Syst., 38, 567–593.
- Bascompte, J., Jordano, P., Melian, C.J. & Olesen, J.M. (2003). The nested assembly of plant-animal mutualistic networks. *Proc. Natl Acad.* Sci. USA, 100, 9383–9387.
- Bates, D., Maechler, M., Bolker, B. & Walker, S. (2015). Fitting linear mixed-effects models using lme4. J. Stat. Softw., 67, 1–48.
- Blüthgen, N., Menzel, F. & Bluthgen, N. (2006). Measuring specialization in species interaction networks. *BMC Ecol.*, 6, 1-12.
- Bosch, J., Retana, J. & Cerdá, X. (1997). Flowering phenology, floral traits and pollinator composition in a herbaceous Mediterranean plant community. *Oecologia*, 109, 583–591.
- Burkle, L.A., Marlin, J.C. & Knight, T.M. (2013). Plant-pollinator interactions over 120 years: loss of species, co-occurrence, and function. *Science*, 339, 1611–1615.
- Burkle, L.A., Myers, J.A. & Belote, R.T. (2016). The beta-diversity of species interactions: Untangling the drivers of geographic variation in plant-pollinator diversity and function across scales. *Am. J. Bot.*, 103, 118–128.
- Canard, E., Mouquet, N., Marescot, L., Gaston, K.J., Gravel, D. & Mouillot, D. (2012). Emergence of structural patterns in neutral trophic networks. *PLoS ONE*, 7, e38295.
- Canard, E.F., Mouquet, N., Mouillot, D., Stanko, M., Miklisova, D. & Gravel, D. (2014). Empirical evaluation of neutral interactions in hostparasite networks. Am. Nat., 183, 468–479.
- CaraDonna, P.J., Petry, W.K., Brennan, R.M., Cunningham, J.L., Bronstein, J.L., Waser, N.M. et al. (2017). Interaction rewiring and the rapid turnover of plant-pollinator networks. Ecol. Lett., 20, 385–394.
- Cariveau, D.P., Nayak, G.K., Bartomeus, I., Zientek, J., Ascher, J.S., Gibbs, J. et al. (2016). The allometry of bee proboscis length and its uses in ecology. PLoS ONE, 11, e0151482.
- Carstensen, D.W., Sabatino, M., Trøjelsgaard, K. & Morellato, L.P.C. (2014). Beta diversity of plant-pollinator networks and the spatial turnover of pairwise interactions. *PLoS ONE*, 9, e112903.
- Chacoff, N.P., Vázquez, D.P., Lomáscolo, S.B., Stevani, E.L., Dorado, J. & Padrón, B. (2012). Evaluating sampling completeness in a desert plant-pollinator network. J. Anim. Ecol., 81, 190–200.
- Chacoff, N.P., Resasco, J. & Vázquez, D.P. (2018). Interaction frequency, network position, and the temporal persistence of interactions in a plant-pollinator network. *Ecology*, 99, 21–28.
- Coux, C., Rader, R., Bartomeus, I. & Tylianakis, J.M. (2016). Linking species functional roles to their network roles. *Ecol. Lett.*, 19, 762–770.
- Dorado, J., Vázquez, D.P., Stevani, E.L. & Chacoff, N.P. (2011). Rareness and specialization in plant–pollinator networks. *Ecology*, 92, 19–25.
- Dormann, C.F., Gruber, B. & Fründ, J. (2008). Introducing the bipartite package: analysingecological networks. *R news*, 8, 8-11.
- Eklöf, A., Jacob, U., Kopp, J., Bosch, J., Castro-Urgal, R., Chacoff, N.P. et al. (2013). The dimensionality of ecological networks. Ecol. Lett., 16, 577–583.
- Encinas-Viso, F., Revilla, T.A. & Etienne, R.S. (2012). Phenology drives mutualistic network structure and diversity: effects on the dynamics of mutualistic networks. *Ecol. Lett.*, 15, 198–208.
- Fort, H., Vázquez, D.P. & Lan, B.L. (2016). Abundance and generalisation in mutualistic networks: solving the chicken-and-egg dilemma. *Ecol. Lett.*, 19, 4–11.
- Frost, C.M., Allen, W.J., Courchamp, F., Jeschke, J.M., Saul, W.-C. & Wardle, D.A. (2019). Using network theory to understand and predict biological invasions. *Trends Ecol. Evol.*, 34(9), 831–843.
- Garibaldi, L.A., Bartomeus, I., Bommarco, R., Klein, A.M., Cunningham, S.A., Aizen, M.A. et al. (2015). Trait matching of flower visitors and crops predicts fruit set better than trait diversity. J. Appl. Ecol., 52, 1436–1444.
- Gillespie, M.A.K., Birkemoe, T. & Sverdrup-Thygeson, A. (2017). Interactions between body size, abundance, seasonality, and phenology in forest beetles. *Ecol. Evol.*, 7, 1091–1100.

- Gower, J.C. (1971). A general coefficient of similarity and some of its properties. *Biometrics*, 27, 857–871.
- Hoehn, P., Tscharntke, T., Tylianakis, J.M. & Steffan-Dewenter, I. (2008). Functional group diversity of bee pollinators increases crop yield. Proc. R. Soc. B., 275, 2283–2291.
- Johnson, S.D. & Raguso, R.A. (2016). The long-tongued hawkmoth pollinator niche for native and invasive plants in Africa. Ann. Bot., 117, 25–36.
- Junker, R.R., Blüthgen, N., Brehm, T., Binkenstein, J., Paulus, J., Martin Schaefer, H. et al. (2013). Specialization on traits as basis for the nichebreadth of flower visitors and as structuring mechanism of ecological networks. Funct. Ecol., 27, 329–341.
- Klumpers, S.G.T., Stang, M. & Klinkhamer, P.G.L. (2019). Foraging efficiency and size matching in a plant-pollinator community: the importance of sugar content and tongue length. *Ecol. Lett.*, 22, 469–479.
- Krishna, A., Guimarães, P.R. Jr, Jordano, P. & Bascompte, J. (2008). A neutral-niche theory of nestedness in mutualistic networks. *Oikos*, 117, 1609–1618.
- Lefcheck, J.S. (2016). piecewiseSEM: piecewise structural equation modelling in r for ecology, evolution, and systematics. *Methods Ecol. Evol.*, 7, 573–579.
- Legendre, P., Desdevises, Y. & Bazin, E. (2002). A statistical test for host-parasite coevolution. Syst. Biol., 51, 217–234.
- Lomáscolo, S.B., Giannini, N., Chacoff, N.P., Castro-Urgal, R. & Vázquez, D.P. (2019). Inferring coevolution in a plant–pollinator network. *Oikos*, 128, 775–789.
- Melin, A., Krenn, H.W., Bowie, R.C.K., Beale, C.M., Manning, J.C. & Colville, J.F. (2019). The allometry of proboscis length in Melittidae (Hymenoptera: Apoidae) and an estimate of their foraging distance using museum collections. *PLoS ONE*, 14, e0217839.
- Memmott, J., Waser, N.M. & Price, M.V. (2004). Tolerance of pollination networks to species extinctions. *Proc. R. Soc. Lond. Ser. B-Biol. Sci.*, 271, 2605–2611.
- Memmott, J., Craze, P.G., Waser, N.M. & Price, M.V. (2007). Global warming and the disruption of plant-pollinator interactions. *Ecol. Lett.*, 10, 710–717
- Morente-López, J., Lara-Romero, C., Ornosa, C. & Iriondo, J.M. (2018). Phenology drives species interactions and modularity in a plant-flower visitor network. *Sci. Rep.*, 8, 9386.
- Nagano, Y., Abe, K., Kitazawa, T., Hattori, M., Hirao, A.S. & Itino, T. (2014). Changes in pollinator fauna affect altitudinal variation of floral size in a bumblebee-pollinated herb. *Ecol. Evol.*, 4, 3395–3407.
- Oksanen, J., Blanchet, F.G., Friendly, M., Kindt, R., Legendre, P., McGlinn, D. *et al.* (2019). vegan. Community Ecology Package.
- Olesen, J.M., Bascompte, J., Dupont, Y.L., Elberling, H., Rasmussen, C. & Jordano, P. (2011). Missing and forbidden links in mutualistic networks. *Proc. R. Soc. B Biol. Sci.*, 278, 725–732.
- Paradis, E., Claude, J. & Strimmer, K. (2004). APE: analyses of phylogenetics and evolution in R language. *Bioinformatics*, 20, 289–290.
- Petanidou, T., Kallimanis, A.S., Tzanopoulos, J., Sgardelis, S.P. & Pantis, J.D. (2008). Long-term observation of a pollination network: fluctuation in species and interactions, relative invariance of network structure and implications for estimates of specialization. *Ecol. Lett.*, 11, 564–575.
- Pleasants, J.M. (1990). Null-model tests for competitive displacement: the fallacy of not focusing on the whole community. *Ecology*, 71, 1078– 1084.
- Poblete Palacios, J.A., Soteras, F. & Cocucci, A.A. (2019). Mechanical fit between flower and pollinators in relation to realized precision and accuracy in the hummingbird-pollinated Dolichandra cynanchoides. *Biol. J. Linn. Soc.*, 126, 655–665.
- Poisot, T., Canard, E., Mouquet, N. & Hochberg, M.E. (2012). A comparative study of ecological specialization estimators. *Methods Ecol. Evol.*, 3, 537–544.
- R Core Team. (2019). R: a language and environment for statistical computing. R Found. Stat. Comput, Vienna Austria.
- Rabeling, S.C., Lim, J.L., Tidon, R., Neff, J.L., Simpson, B.B. & Pawar, S. (2019). Seasonal variation of a plant-pollinator network in the

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Brazilian Cerrado: Implications for community structure and robustness. *PLoS ONE*, 14, e0224997.

- Ramos-Jiliberto, R., Moisset de Espanés, P., Franco-Cisterna, M., Petanidou, T. & Vázquez, D.P. (2018). Phenology determines the robustness of plant-pollinator networks. Sci. Rep., 8, 14873.
- Rudolf, V.H.W. (2019). The role of seasonal timing and phenological shifts for species coexistence. *Ecol. Lett.*, 22, 1324-1338.
- Santamaría, L. & Rodríguez-Gironés, M.A. (2007). Linkage rules for plant–pollinator networks: trait complementarity or exploitation barriers? PLOS Biol., 5, e31.
- Schleuning, M., Fründ, J., Schweiger, O., Welk, E., Albrecht, J., Albrecht, M. et al. (2016). Ecological networks are more sensitive to plant than to animal extinction under climate change. Nat. Commun., 7, 13965.
- Shipley, B. (2009). Confirmatory path analysis in a generalized multilevel context. *Ecology*, 90, 363–368.
- Solís-Montero, L. & Vallejo-Marín, M. (2017). Does the morphological fit between flowers and pollinators affect pollen deposition? An experimental test in a buzz-pollinated species with anther dimorphism. *Ecol. Evol.*, 7, 2706–2715.
- Stang, M., Klinkhamer, P.G.L. & van der Meijden, E. (2006). Size constraints and flower abundance determine the number of interactions in a plant-flower visitor web. *Oikos*, 112, 111–121.
- Stang, M., Klinkhamer, P.G.L. & van der Meijden, E. (2007). Asymmetric specialization and extinction risk in plant–flower visitor webs: a matter of morphology or abundance? *Oecologia*, 151, 442–453.
- Stang, M., Klinkhamer, P.G.L., Waser, N.M., Stang, I. & van der Meijden, E. (2009). Size-specific interaction patterns and size matching in a plant–pollinator interaction web. *Ann. Bot.*, 103, 1459–1469.
- Staniczenko, P.P.A., Kopp, J.C. & Allesina, S. (2013). The ghost of nestedness in ecological networks. *Nat. Commun.*, 4, 1391.
- Staniczenko, P.P.A., Lewis, O.T., Tylianakis, J.M., Albrecht, M., Coudrain, V., Klein, A.M. et al. (2017). Predicting the effect of habitat modification on networks of interacting species. Nat. Commun., 8, 792.
- Thompson, J.N. (1998). Rapid evolution as an ecological process. *Trends Ecol. Evol.*, 13, 329–332.
- Trøjelsgaard, K. & Olesen, J.M. (2016). Ecological networks in motion: micro- and macroscopic variability across scales. *Funct. Ecol.*, 30, 1926–1935.
- Trøjelsgaard, K., Jordano, P., Carstensen, D.W. & Olesen, J.M. (2015). Geographical variation in mutualistic networks: similarity, turnover and partner fidelity. Proc. R. Soc. B Biol. Sci., 282, 20142925.

- Tylianakis, J.M. & Morris, R.J. (2017). Ecological networks across environmental gradients. *Annu. Rev. Ecol. Evol. Syst.*, 48, 25–48.
- Vázquez, D.P., Morris, W.F. & Jordano, P. (2005). Interaction frequency as a surrogate for the total effect of animal mutualists on plants. *Ecol. Lett.*, 8, 1088–1094.
- Vázquez, D.P., Blüthgen, N., Cagnolo, L. & Chacoff, N.P. (2009a). Uniting pattern and process in plant-animal mutualistic networks: a review. Ann. Bot., 103, 1445–1457.
- Vázquez, D.P., Chacoff, N.P. & Cagnolo, L. (2009b). Evaluating multiple determinants of the structure of plant-animal mutualistic networks. *Ecology*, 90, 2039–2046.
- Vázquez, D.P., Lomáscolo, S.B., Maldonado, M.B., Chacoff, N.P., Dorado, J., Stevani, E.L. et al. (2012). The strength of plant–pollinator interactions. *Ecology*, 93, 719–725.
- Visser, M.E. & Gienapp, P. (2019). Evolutionary and demographic consequences of phenological mismatches. *Nat. Ecol. Evol.*, 3, 879– 885
- Winfree, R., Williams, N.M., Dushoff, J. & Kremen, C. (2014). Species abundance, not diet breadth, drives the persistence of the most linked pollinators as plant-pollinator networks disassemble. *Am. Nat.*, 183, 600–611.
- Wolf, A.A., Zavaleta, E.S. & Selmants, P.C. (2017). Flowering phenology shifts in response to biodiversity loss. *Proc. Natl Acad. Sci.*, 114, 3463– 3468.
- Wootton, J.T. & Emmerson, M. (2005). Measurement of interaction strength in nature. *Annu. Rev. Ecol. Evol. Syst.*, 36, 419–444.

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Editor, Rebecca Irwin Manuscript received 4 December 2019 First decision made 11 January 2020 Second decision made 5 March 2020 Manuscript accepted 21 March 2020