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Related plants tend to share pollinators and herbivores, but strength of phylogenetic signal varies among plant families

Alyssa R. Cirtwill^{†1,2}, Giulio V. Dalla Riva³, Nick J. Baker¹, Mikael Ohlsson⁴, Isabelle Norström⁴, Inger-Marie Wohlfarth⁴, Joshua A. Thia¹, & Daniel B. Stouffer¹

[†] Corresponding author: alyssa.cirtwill@gmail.com; +46 8 16 20 00 ¹Centre for Integrative Ecology, School of

Biological Sciences University of Canterbury Private Bag 4800 Christchurch 8140, New Zealand

²Present address: Department of Ecology, Environment, and Plant Sciences (DEEP) Stockholm University 114 19 Stockholm, Sweden ³Biomathematics Research Centre, School of Mathematics and Statistics University of Canterbury Private Bag 4800 Christchurch 8140, New Zealand

⁴Department of Physics, Chemistry, and Biology (IFM) Linköping University 581 83 Linköping, Sweden

ORCID IDs

Alyssa Cirtwill: https://orcid.org/0000-0002-1772-3868 Giulio V. Dalla Riva: https://orcid.org/0000-0002-3454-0633 Joshua A. Thia: https://orcid.org/0000-0001-9084-0959 Daniel B. Stouffer: https://orcid.org/0000-0001-9436-9674

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Summary

- Related plants are often hypothesised to interact with similar sets of pollinators and herbivores, but this idea has only mixed empirical support. This may be because plant families vary in their tendency to share interaction partners.
- We quantify overlap of interaction partners for all pairs of plants in 59 pollination and 11 herbivory networks based on the numbers of shared and unshared interaction partners (thereby capturing both proportional and absolute overlap). We test 1) for relationships between phylogenetic distance and partner overlap within each network, 2) whether these relationships varied with the composition of the plant community, and 3) whether well-represented plant families showed different relationships.
- Across all networks, more closely-related plants tended to have greater overlap. The strength of this relationship within a network was unrelated to the composition of the network's plant component, but, when considered separately, different plant families showed different relationships between phylogenetic distance and overlap of interaction partners.
- The variety of relationships between phylogenetic distance and partner overlap in different plant families likely reflects a comparable variety of ecological and evolutionary processes. Considering factors affecting particular species-rich groups within a community may be the key to understanding the distribution of interactions at the network level.

Keywords

ecological networks, herbivory, niche overlap, phylogenetic signal, pollination, specialisation

Author

Introduction

Interactions with animals affect plants' life cycles in several critical ways (Mayr, 2001). On one hand, pollination and other mutualistic interactions contribute to the reproductive success of many angiosperms (Ollerton et al., 2011). On the other, herbivores consume plant tissues (McCall & Irwin, 2006) which costs plants energy and likely lowers their fitness (Strauss et al., 2002). In both cases, these interactions do not occur randomly but are strongly influenced by plants' phenotypes (Fontaine & Thébault, 2015). For example, plants that produce abundant or high-quality nectar may receive more visits from pollinators (Robertson et al., 1999) whereas plants that produce noxious secondary metabolites may have fewer herbivores (Johnson *et al.*, 2014). Plant traits are also likely 9 to determine *which* specific pollinators and herbivores interact with a particular plant. 10 Plants with different defences (e.g., thorns vs. chemical defences) may deter different 11 groups of herbivores (Ehrlich & Raven, 1964; Johnson et al., 2014), and pollinators with 12 similar traits are often expected to attract similar sets of pollinators (Waser et al., 1996; 13 Fenster et al., 2004; Ollerton et al., 2009). 14

If attractive and/or defensive traits are heritable, then we can reasonably expect that 15 related plants will have similar patterns of interactions with animals, especially if there is 16 some selection in either group to avoid competition or the number of potential partners 17 is limited (Schemske & Bradshaw, 1999; Ponisio *et al.*, 2017). That is, there may be *phy*-18 logenetic signal in plants' interactions such that closely-related plants may tend to have 19 similar interaction partners. Recent studies that have investigated this question at the 20 level of whole communities, however, have yielded mixed results. In particular, significant 21 phylogenetic signal in plants' sets of interaction partners tends to be rare in empirical 22 networks (Rezende et al., 2007b; Lind et al., 2015; Ibanez et al., 2016; but see Elias et al., 23 2013; Fontaine & Thébault, 2015; Hutchinson et al., 2017). Moreover, statistically sig-24 nificant degrees of phylogenetic signal or coevolution may only result in small differences 25 in network structure, adding to the difficulty of understanding patterns in species' in-26

teraction partners (Ponisio et al., 2017). Further, the plant and animal components of 27 networks can show different degrees of phylogenetic conservation of interaction partners. 28 In mutualistic networks, animals often show a stronger phylogenetic signal in their part-29 ners than do plants (Rezende et al., 2007b; Chamberlain et al., 2014; Rohr et al., 2014; 30 Vamosi et al., 2014; Lind et al., 2015; Fontaine & Thébault, 2015) (but see Rafferty & 31 Ives (2013) for a counterexample). In antagonistic networks, however, actively-foraging 32 consumers tend to show less phylogenetic signal than their prey (Ives & Godfray, 2006; 33 Cagnolo et al., 2011; Naisbit et al., 2011; Fontaine & Thébault, 2015). In part, this 34 may be related to different degrees of interaction intimacy (dependence of one partner 35 on another), which appears to contribute to network structure in mutualistic, but not 36 antagonistic, networks (Guimarães et al., 2007; Ponisio et al., 2017). In any case, it is not 37 straightforward to assume that interactions will always be similar among related species. 38 There are several mechanisms that might weaken the conservation of interaction part-30 ners. Pollination and herbivory may be affected by a wide variety of traits, and not all of 40 these are likely to be phylogenetically conserved (Rezende *et al.*, 2007a; Kursar *et al.*, 2009; 41 Ibanez et al., 2016). If, for example, floral displays are strongly affected by environmental 42 conditions (Canto et al., 2004), then plant phylogeny may not strongly predict pollination. 43 Even if the traits affecting pollination and herbivory are heritable, plants may experience 44 conflicting selection pressures that weaken the overall association between plant phylogeny 45 and interaction partners (Armbruster, 1997; Lankau, 2007; Siepielski et al., 2010; Wise 46 & Rausher, 2013; Kariñho-Betancourt et al., 2015). For instance, floral traits that are 47 attractive to pollinators can also increase herbivory (Strauss et al., 2002; Adler & Bron-48 stein, 2004; Strauss & Whittall, 2006; Theis, 2006). Conversely, herbivory can reduce 49 pollination by inducing chemical defences (Adler *et al.*, 2006) or altering floral display or 50 nectar availability (Strauss, 1997). There may also be trade-offs between chemical and 51 physical defences, or defences at different life stages, that weaken the overall heritability 52 of plants' sets of herbivores (Kariñho-Betancourt et al., 2015; Endara et al., 2017). A 53

⁵⁴ plant's set of interaction partners therefore reflects a mixture of different environmental ⁵⁵ effects and different selection pressures, as well as shared phylogenetic history. If these ⁵⁶ factors affect closely-related plants differently, then closely-related species may not have ⁵⁷ more similar interaction partners than distantly-related species.

This variety of different pressures makes it likely that the relationship between plants' 58 relatedness and the similarity of their interaction partners is not constant across plant 59 Closely-related plants in one clade might be under strong selection to favour clades. 60 dissimilar sets of pollinators to avoid exchanging pollen with other species (Levin & An-61 derson, 1970; Bell et al., 2005; Mitchell et al., 2009), while plants in other clades may be 62 under strong pressure to continue interacting with a common set of partners. Similarly, 63 plants may experience disruptive selection on defences against herbivores if congeners 64 tend to grow in the same places such that herbivore able to consume one species could 65 easily spread to close relatives (Kursar et al., 2009; Yguel et al., 2014). On the other 66 hand, unrelated plants might converge upon similar phenotypes which attract particu-67 larly efficient or abundant pollinators (Ollerton, 1996; Wilson et al., 2007; Ollerton et al., 68 2009; Ibanez et al., 2016). Likewise, unrelated plants may converge upon similar defences, 69 leading them to share those herbivores which can overcome these defences (Pichersky & 70 Gang, 2000). In either case, dissimilarity of interactions among related species or simi-71 larity of interactions among unrelated species could result in weaker phylogenetic signal 72 across an entire plant community. Moreover, all of the aforementioned hypotheses are 73 non-exclusive; different processes likely affect different clades, and these processes might 74 be associated with different pressures imposed by pollination and herbivory (Fontaine & 75 Thébault, 2015). 76

Here we investigate how overlap in interaction partners between pairs of plants (henceforth "niche overlap") varies over phylogenetic distance. Whereas previous studies have focused on the presence or absence of phylogenetic signal across entire networks, we take a pairwise perspective in order to obtain a more detailed picture of how plant phylogeny

relates to network structure. As different plant families (which represent tractable clades 81 for analysis) may have experienced different degrees of coevolution, convergence, etc., 82 we also complement analyses with entire networks with comparisons among plants in the 83 same family within a network. This novel perspective allows us to investigate the relation-84 ship between phylogenetic distance and partner overlap at different scales. Specifically, 85 we test 1) whether niche overlap decreases over increasing phylogenetic distance in a large 86 dataset of pollination and herbivory networks, 2) whether the plant family composition 87 of a community affects the relationship between niche overlap and phylogenetic distance 88 in that community, and 3) whether the relationship between niche overlap and phylo-89 genetic distance differs systematically across plant families. This fine-grained approach 90 gives more detailed information than previous studies. 91

⁹² Materials and Methods

93 Network data

We tested for phylogenetic signal in niche overlap within a set of 59 pollination and 11 94 herbivory networks. These networks span a range of biomes (desert to grassland to tundra) 95 and countries (Sweden to New Zealand). The herbivory networks included a variety of 96 types of herbivores but were dominated by leaf-chewing insects. Leaf-chewing and other 97 types of herbivory might be affected by different plant traits and cannot be expected to 98 show the same trends with respect to phylogeny. We therefore restricted our networks to 99 leaf-chewing insects by removing any non-leaf chewing insects and any plants which had 100 no interaction partners after removing other types of herbivores. The adjusted networks 101 range in size between 19 and 997 total species (mean=162, median=97) with between 8 102 and 132 plant species (mean=39.1, median=29.5). See Table S1, Notes S1 for details on 103 the original sources of all networks. All networks were qualitative and did not include 104 interaction strengths. 105

¹⁰⁶ Phylogenetic data

In order to fit the plant species in all networks to a common phylogeny, we first compared 107 all species and genus names with the National Center for Biotechnology Information and 108 Taxonomic Name Resolution Service databases to ensure correctness. This was done 109 using the function 'get_tsn' in the R (R Core Team, 2016) package taxize (Chamberlain 110 & Szocs, 2013: Chamberlain et al., 2019). Species which could not be assigned to an 111 accepted taxonomic name (e.g., 'Unknown Forb') were discarded, as were those with 112 binomial names that could not be definitively linked to higher taxa (e.g., 'Salpiqlossus 113 sp. '). We were left with 2341 unique species in 1027 genera and 195 families. On average, 114 11.43% of plants were removed from each network (median 4.60%, range 0.55.10%). 115

We then estimated phylogenetic distances between the remaining species. To ac-116 complish this, we constructed a phylogenetic tree based on a dated 'mega-tree' of an-117 giosperms (Zanne et al., 2014). Some species in our dataset were not included in the 118 angiosperm mega-tree. For angiosperms, a sister taxon was identified using Stevens (2001) 119 onwards) and the species added manually. Ferns, tree ferns, and a single club moss were 120 added to the base of the tree. This means that closely-related non-angiosperm species 121 appear to have very long phylogenetic distances between them. We therefore excluded 122 comparisons between pairs of non-angiosperms from our analyses. As only two networks 123 (both herbivory networks) included more than one such species and non-angiosperms were 124 always a small minority of any network, we do not believe that omitting these comparisons 125 has greatly affected our results. To obtain trees for each network, we pruned the dated 126 mega-tree to include only species in that network. 127

¹²⁸ Calculating niche overlap

We calculated niche overlap for each pair of plants i and j based on the number of shared and unshared interaction partners (M_{ij} , U_{ij} , respectively). The number of unshared interaction partners gives valuable information about cases where, for example, closely-

related plants may have experienced disruptive selection, leading to weaker phylogenetic signal. The sum $M_{ij} + U_{ij}$ indicates the amount of information provided by each pair of plants: a pair of generalists which share most of their interaction partners gives a stronger indication of phylogenetic signal than a pair of extreme specialists with one common interaction partner.

Together, M_{ij} and U_{ij} give a Jaccard index (J_{ij}) describing the proportion of shared interactions. J_{ij} is defined:

$$J_{ij} = \frac{M_{ij}}{U_{ij} + M_{ij}},\tag{1}$$

where M_{ij} is the set of *mutual* (shared) interaction partners and U_{ij} the set of unshared interaction partners for plants *i* and *j*. In our statistical analyses (see below), we used the tuple (M_{ij}, U_{ij}) as the dependent variable rather than the single value J_{ij} . This allows us to preserve information about the amount of information provided by each pair of plants and weight the observations accordingly. Note that species sharing a large *number* of interaction partners may not share a large *proportion* of interaction partners if the number of interaction partners that are not shared is also large.

¹⁴⁶ Testing conservation of niche overlap within networks

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We modelled the relationship between niche overlap and phylogenetic distance using a 147 logistic regression. We used the numbers of shared (M_{ij}) and non-shared (U_{ij}) partners as 148 dependent variables and centred, scaled phylogenetic distance as the independent variable. 149 This approach is conceptually similar to modelling successes and failures in a binomial-150 distributed process. Accordingly, we assumed a binomially-distributed error structure and 151 used a logit link function to model the dissimilarity in interaction partners J_{ij} of plants i 152 and j. Regressions of niche overlap and phylogenetic distance within each network were 153 fit using the R (R Core Team, 2016) base function "glm" and took the form 154

$$\operatorname{logit}(J_{ij}) \propto \beta_{distance} \delta_{ij},\tag{2}$$

where δ_{ij} is the phylogenetic distance between plants *i* and *j* and J_{ij} is defined by the tuple (M_{ij}, U_{ij}) (see Notes S2 for R implementation). The fixed effect of distance in this regression, $\beta_{distance}$, can be understood as the change in log odds of sharing an interaction partner per million-year change in phylogenetic distance.

These separate regressions avoid the potential for confounding the effects of differ-159 ent relationships in different networks. As we also wished to evaluate the overall trend 160 across networks, we fit an additional regression of niche overlap and phylogenetic dis-161 tance across all network types. As well as the fixed effect of phylogenetic distance, this 162 regression included fixed effects of network type (pollination or herbivory) and the inter-163 action between phylogenetic network type and random intercepts and slopes per network. 164 This expanded regression was fit using the R (R Core Team, 2016) function 'glmer' from 165 package lme4 (Bates *et al.*, 2015) and took the form 166

$$logit (J_{ij}) \propto \beta_{distance} \delta_{ij} + \beta_{pollination} I_{ij} + \beta_{distance:pollination} \delta_{ij} I_{ij}, \qquad (3)$$

where $I_{ij} = 1$ when plants *i* and *j* are drawn from a pollination network and $I_{ij} = 0$ when 167 i and j are drawn from a herbivory network, and all other symbols are as above. Note 168 that we only compared pairs of plants taken from the same network. The fixed effects 169 $\beta_{pollination}$ and $\beta_{distance:pollination}$ are the change in intercept and slope of the log odds of 170 sharing an interaction partner, respectively, relative to the baseline of herbivory networks. 171 To demonstrate the power of defining J_{ij} as a tuple of M_{ij} and U_{ij} rather than a 172 single value, we repeated the above analyses using a Jaccard index based only on the 173 proportion of interaction partners that are shared (i.e., $J_{ij} = M_{ij}/[M_{ij} + U_{ij}]$). Note that 174 while the proportion of shared interaction partners is the same in both cases, the tuple 175 formulation gives more weight to plants with many interaction partners as these provide 176 more information. When comparing the two approaches we observed similar trends but, 177 notably, the tuple definition of J_{ij} had greater power to detect weak relationships (Notes 178 S3). We therefore show only the results when defining J_{ij} as a tuple in the main text. 179

To test whether the relationship between phylogenetic distance and niche overlap depended on network size, we fit a general linear model for the slope of this relationship inferred from the glm models against the number of plant pairs for which distances could be calculated (hereafter "network size"), network type (again using herbivory networks as a baseline), and their interaction:

$$\beta_{distance} \propto \beta_{size} \eta_N + \beta_{pollination} I_N + \beta_{size:pollination} \eta_N I_N, \tag{4}$$

where η_N is the number of plant pairs in network N for which distances could be calcu-185 lated, I_N is an indicator equal to 1 if network N is a pollination network and 0 otherwise. 186 As the interaction between network type and network size was strong and opposite to 187 the direction of the main effect of network size, we fit an additional general linear model 188 using only data from pollination networks and including only the effect of network size 189 (herbivory networks were the baseline in the full glm). Both models were fit using the 190 R (R Core Team, 2016) base function "glm". A similar model relating the strength of the 191 relationship between phylogenetic distance and niche overlap to connectance showed no 192 significant trends (*Notes* S_4). 193

194 Accounting for non-independence

Note that pairs of plants are not independent: the same plant will appear in many pairs, 195 and interactions may be influenced by the overall structure of the community. This vio-196 lates the assumptions used when calculating the significance of logistic regressions within 197 the R (R Core Team, 2016) base package or the package *lme4* (Bates *et al.*, 2015). To 198 fairly estimate the significance of our regressions, it was therefore necessary to compare 199 the observed relationships to those in a suite of appropriately permuted networks. To 200 create these networks, we shuffled interactions among species while preserving row and 201 column totals. Each species retained the same number of interaction partners as in the 202 observed network but the exact set of partners (and therefore niche overlaps with all 203

other species) varied across permuted networks. We preserved the observed phylogenetic relationships between species in all cases. For each observed network, we created 999 such permuted networks and calculated the relationship between niche overlap and phylogenetic distance. This gave us a null distribution for each observed network with which to determine the significance of the observed relationship.

This permutation approach also allows us to estimate type I and type II error for 209 our analysis. Because the permuted networks should not demonstrate any particular 210 relationship between phylogenetic distance and partner overlap, these slopes should be 211 similar to those obtained after permuting these networks a second time. To estimate type 212 I and type II errors, we created 500 permutations of each permuted network and, again 213 keeping the observed phylogenetic distances between plant species, repeated our analyses. 214 We then determined the number of permuted networks which appear to have significant 215 overlap-phylogenetic distance relationships relative to the permutations of these permuted 216 networks (type I error). Type II error can be determined from the distribution of p-217 values obtained when comparing the permuted networks to permutations of the permuted 218 networks. Although calculating the exact type II error requires a specific alternative 219 hypothesis, the uniform distribution of *p*-values we obtained after permuting the permuted 220 networks means that the type II error would increase linearly as the alternative hypothesis 221 was set farther from zero (*Notes* S5). 222

²²³ Linking network-level trends and community composition

Next, we examined the connection between our network-level observations and the number of species in each plant family present in each community. Specifically, we tested the hypothesis that varying relationships between phylogenetic distance and pairwise niche overlap are due to the different distributions of families across networks. We defined the relationship between phylogenetic distance and niche overlap as the change in log odds of two plants in a given network sharing an interaction partner per million years

of divergence (i.e., the slope $\beta_{distance}$ from the regression of niche overlap against phylo-230 genetic distance within a single network). We then related differences in this relation-231 ship to differences in Bray-Curtis dissimilarity in the family-wise composition of the two 232 plant communities using a non-parametric permutational multi-variate analysis of vari-233 ance (PERMANOVA; Anderson, 2001). Bray-Curtis dissimilarity considers only those 234 plant families which appear in at least one of a pair of networks (Anderson, 2001; Cirtwill 235 & Stouffer, 2015), ensuring that the shared absence of rare plant families will not make 236 two networks appear more similar than they actually are. 237

Note that a PERMANOVA does not assume that the data are normally distributed, 238 but rather compares the pseudo-F statistic calculated from the observed data to a null 239 distribution obtained by permuting the raw data. As pollination and herbivory networks 240 might have different community composition, we stratified these permutations by network 241 type. That is, the response variable of change in log odds for a pollination network could 242 only be exchanged for that of another pollination network. This stratification procedure 243 ensures that the null distribution used to calculate the *P*-value is not biased by including 244 combinations of changes in log odds and community composition that would not occur 245 because of inherent differences in the two network types (e.g., *Pinaceae* only appeared in 246 herbivory networks and should not be assigned to pollination networks). We used 9999 247 such stratified permutations to obtain the null distribution and obtain a *P*-value. 248

249 Calculating niche overlap within families

Finally, we compared the breakdown of niche overlap in different plant families. Withinfamily genetic and trait diversity can be high due to adaptive radiations, heterogeneous selection, and other influences on different species. Plant families offer a reasonable balance between collecting enough species to identify meaningful trends and maintaining a tractable number of analyses. They are therefore the best taxonomic level to investigate phylogenetic conservation in more detail across our large dataset. To test whether different families show different conservation of interactions, we used the same definitions of overlap and phylogenetic distance as in the within-network analysis but restricted our regressions to pairs of plants from the same family and the same network. Unlike in our previous analysis, we analysed data from pollination and herbivory networks separately as most well-represented plant families appeared in only one network type. For those families which appeared in both network types, we ran separate analyses on each subset of data.

For each plant family, within each network type, we fit one of two similar sets of 263 models. If family f was found in several networks of the same type (e.g., several pollination 264 networks), we fit a mixed-effects logistic regression relating niche overlap to a fixed effect 265 of phylogenetic distance and a random effect for each network using the R (R Core Team, 266 2016) function "glmer" from package lme4 (Bates et al., 2015). If family f was found in 267 only one network, we omitted the network-level random effect and fit a logistic regression 268 using the R (R Core Team, 2016) base function "glm". These equations took the same 269 form as equation 2. 270

Models for two families did not converge. In the *Lauraceae*, (represented by four species in one pollination network) and the *Sapindaceae* (represented by five species in one herbivory network and five species in two pollination networks), only one pair of species per network type shared any interaction partners while all other pairs did not share any interaction partners.

By considering each family separately, we do risk obtaining some significant results purely by chance. The standard technique for addressing this type of multiple hypothesis testing, the Bonferroni correction, tends to be over-zealous and lead to a failure to reject the null hypothesis even when a large number of significant results before the correction supports the alternative hypothesis (Moran, 2003). To account for multiple testing while also allowing the number of families showing significant trends to carry some weight, we use the correlated Bonferroni test introduced in Drezner & Drezner (2016) (*Notes S6*).

$_{283}$ Results

²⁸⁴ Within-network conservation of niche overlap

Across all networks, more distantly-related plants were less likely to share interaction 285 partners ($\beta_{distance} = -6.82, p < 0.001$). Plants in pollination networks tended to share fewer 286 interaction partners overall, and the decrease in overlap with increasing phylogenetic 287 distance was steeper ($\beta_{pollination} = -1.44$, p < 0.001 and $\beta_{distance:pollination} = -18.5$, p < 0.001, re-288 spectively). That is, a pair of plants in the same genus was more likely to share interaction 289 partners than a pair of plants in the same family in both types of networks, but a pair of 290 congeners would be less likely to share pollinators than to share herbivores. Note that, as 291 our networks are qualitative, these results refer only to the number of shared interaction 292 partners rather than to the quantitative strength of competition. 293

As an illustration, a pair of plants which diverged 10 mya would have a probability of 294 0.202 of sharing a given herbivore and 0.094 of sharing a given pollinator, while a pair 295 of plants which diverged 750mya would have a probability of 0.121 of sharing a given 296 herbivore or 0.011 of sharing a given pollinator. These trends may be related to the 297 numbers of extreme specialists in each network. In our dataset, an average of 48% (+/-298 14) of pollinators in a given web were extreme specialists (i.e., visited only one plant 299 species) compared to 29% (+/- 29) of herbivores (z=5.62, df=68, P<0.001 for a binomial 300 regression of specialists and generalists over network type). 301

Despite these general trends, there was substantial variation between pollination networks, with overlap of interaction partners decreasing with increasing phylogenetic distance in some networks and increasing in others (Fig. 1). Overlap of interaction partners decreased significantly with increasing phylogenetic distance in 7/11 herbivory networks and 33/59 pollination networks. In the remaining four herbivory networks and 25 of the 26 remaining pollination networks, overlap of interaction partners was not related to phylogenetic distance. Overlap of interaction partners increased with increasing phylogenetic ³⁰⁹ distance in only a single pollination network.

The slope of the relationship between phylogenetic distance and overlap of interaction 310 partners was related to the number of plant pairs in herbivory, but not pollination, net-311 works. Larger herbivory networks had higher values of $\beta_{distance}$ ($\beta_{size} = 2.58 \times 10^{-4}$, p = 0.011312 for the full glm; herbivory networks are the baseline). Pollination networks had higher 313 (less negative) slopes overall ($\beta_{pollinator}=0.306$, p<0.001 compared to the intercept value 314 of -0.434 for herbivory networks). Pollination networks moreover showed a much weaker 315 relationship between network size and the strength of the overlap-distance relationship 316 $(\beta_{pollination:size} = -2.64 \times 10^{-4}, p = 0.009)$. After refitting the glm to the pollination networks 317 alone, there was no significant relationship between network size and the slope of the 318 overlap-distance relationship ($\beta_{size} = -5.91 \times 10^{-6}$, p = 0.572). 319

Comparing the results in the observed networks to those obtained after permuting 320 interactions, the observed slope of the relationship between phylogenetic distance and 321 interaction partner overlap was always more extreme (i.e., always more negative or always 322 more positive) than that obtained in the permuted networks (Fig. 2). Observed networks 323 with a negative relationship between phylogenetic distance and overlap always had a more 324 negative slope than that obtained from the permuted networks, while the 10 networks with 325 positive relationships between phylogenetic distance and overlap always had more positive 326 relationships than the permuted networks. This indicates that even in the networks 327 with non-significant relationships, the association between niche overlap and phylogenetic 328 distance was not random and confirms that the significant results we observe are not 329 due to non-independence of plants within a network. When the slopes of the permuted 330 networks were compared to those obtained from permutations of the permuted networks, 331 there was no relationship, which speaks to the robustness of our methodology (*Notes S5*). 332

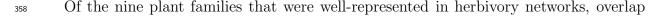
³³³ Linking network-level trends and community composition

We were interested in whether the slope of the relationship between phylogenetic distance 334 and niche overlap varied with community composition. In a PERMANOVA of slope 335 against community composition, stratified by network type, we did not find a significant 336 relationship between slope and community composition ($F_{1,68}=1.06$, p=0.493). Of the 337 200 families in our dataset, only 29 were represented by more than 20 species. Lumping 338 all other families into an "other" category and repeating the PERMANOVA, we still did 339 not find a significant relationship between slope and community composition ($F_{1,68}=1.12$, 340 p=0.409). 341

³⁴² Within-family conservation of niche overlap

Taking all families together, the probability of species in the same family sharing interaction partners was not significantly related to phylogenetic distance ($\beta_{distance}$ =-6.48, p=0.087). Pollination networks did not show a significantly different slope from the herbivory networks ($\beta_{distance:pollination}$ =1.73, p=0.681). Plants in pollination networks did, however, have a lower intercept probability of sharing interaction partners ($\beta_{pollination}$ =-0.776, p=0.007), similar to our within-network results above.

Considering each family separately, the relationship between within-family niche over-349 lap and phylogenetic distance varied widely in both pollination and herbivory networks. 350 In pollination networks, overlap decreased significantly with increasing phylogenetic dis-351 tance in 14 of the 48 well-represented families (Table 1; Fig. 3). If we apply the correlated 352 Bonferroni correction to account for multiple testing (Drezner & Drezner, 2016), all of 353 these slopes remain significant (*Notes S6*). There was no significant relationship between 354 overlap and phylogenetic distance in a further 34 plant families (see Notes S6 for further 355 details). Finally, the overlap between pairs of *Apiaceae* and *Poaceae* increased signifi-356 cantly with increasing phylogenetic distance. 357



decreased significantly with increasing phylogenetic distance in four (Table 2; Fig. 3). Four families did not show significant relationships between phylogenetic distance and overlap, and in one family, *Fabaceae*, overlap of interaction partners increased significantly with increasing phylogenetic distance. If we again apply the correlated Bonferroni correction, all five significant slopes remain significant (*Supportin information 6*).

364 Discussion

We found general support for the hypothesis that more closely-related pairs of plants 365 have a higher degree of niche overlap. Taking all networks together, the probability of 366 two plants sharing the same animal interaction partners decreased with increasing phy-367 logenetic distance. Considering networks separately, $\approx 56\%$ of pollination and $\approx 64\%$ of 368 herbivory networks showed the expected trend of decreasing overlap with increasing dis-369 tance. This variation between networks echoes earlier studies (e.g., Fontaine & Thébault, 370 2015; Hutchinson et al., 2017), which also found broad evidence for phylogenetic conser-371 vation of interaction partners despite variation between particular networks. The lack of 372 a significant relationship between phylogenetic distance and niche overlap in many net-373 works could be partly due to the large number of extreme specialist insects, especially 374 in the pollination networks. These species interact with only one plant and therefore 375 weaken any signal of niche overlap. The herbivory networks did not contain as many 376 obligate specialists, but we note that herbivores, like pollinators, often interact with only 377 a few closely-related plants (Novotny & Basset, 2005; Brändle & Brandl, 2006; Astegiano 378 et al., 2017). These oligotrophs may affect overall phylogenetic signal in the same way as 379 the strict specialists: in both cases plants that are not very closely related are unlikely 380 to share interaction partners. Note that some of the apparent specialists in our dataset 381 may actually be rare species involved in more interactions which have not yet been ob-382 served (Blüthgen et al., 2006; Poisot et al., 2015). Without information on the sampling 383 completeness of the networks in our dataset, it is difficult to estimate the size of this effect. 384

It is possible, however, that we might observe stronger relationships between phylogenetic
distance and niche overlap with more complete data on rare species.

In our dataset, the slope of the relationship between phylogenetic distance and niche 387 overlap was not related to the composition of the plant community in each network. Com-388 bined with the overall trend for conservation of interaction partners above, this suggests 389 that trends among closely-related plants (e.g., congeners or members of the same subfam-390 ilies) are more important than phylogenetic signal from deeper within the phylogenetic 391 tree. This echoes earlier results relating plant phylogeny to predation by particular insect 392 species (Novotny et al., 2002, 2004; Ødegaard et al., 2005) and in whole herbivory net-393 works (Volf *et al.*, 2017). As we did not find any relationship between the families present 394 in a network and the relationship between phylogenetic distance and niche overlap in 395 either pollination or herbivory networks, the greater importance of shallow phylogeny (as 396 reported for leaf miners and gallers in Volf *et al.* (2017) may be a general feature of 397 plant-insect interaction networks. This contrasts with Chamberlain et al. (2014), who 398 found that the shape of the phylogenetic tree had a larger effect on network structure 399 than the timing of speciation. As Chamberlain *et al.* (2014) were interested in overall 400 structural properties of networks rather than niche overlap, this discrepancy may indi-401 cate that different aspects of plant-insect interaction networks are influenced by different 402 aspects of plant phylogenies. 403

The variability of the strength of phylogenetic signal across networks and the lack of 404 influence of community composition on the strength of this signal could be partly due 405 to different trends within families. More than half of the plant families in each network 406 type behaved as we hypothesised, with more closely-related plants having greater niche 407 overlap than distantly related plants. This relationship between overlap and phyloge-408 netic distance is consistent with the idea that traits affecting interactions are heritable 409 and change gradually such that closely related plants resemble their common ancestor— 410 and each other — more than they do distantly related plants (Schemske & Bradshaw, 411

1999; Gilbert et al., 2015; Ponisio et al., 2017). The degree of heritability of key traits 412 may, however, differ between families. In some families, such as Asteraceae in pollination 413 networks, the positive slope of this relationship was very shallow while in others, such 414 as *Melastomataceae* in herbivory networks, the positive slope was extremely steep. This 415 could indicate different rates of phenotypic drift or evolution in different families (or their 416 interaction partners). In other families, there was no significant relationship between 417 phylogenetic distance and niche overlap. In these cases, key traits affecting plant-insect 418 interactions may be highly labile or plastic (environmentally determined). These possi-419 bilities are supported by several studies showing a stronger relationship between niche 420 overlap and trait similarity than niche overlap and phylogenetic similarity (Junker et al., 421 2015; Ibanez et al., 2016; Endara et al., 2017). 422

While the majority of plant families in our dataset showed the expected trend, two 423 (Polygonaceae in pollination networks and Fabaceae in herbivory networks) showed the 424 opposite pattern. In these families, closely-related plants had *lower* overlap than more 425 distantly-related pairs of plants. There are several possible explanations for this pattern. 426 First, part of the family may have recently undergone a period of rapid diversification 427 with closely-related species developing novel phenotypes and attracting different inter-428 action partners (Linder, 2008; Breitkopf et al., 2015). Likewise, the animals may have 429 undergone an adaptive radiation to specialise on their most profitable partner (Janz et al., 430 2006). Alternatively, plants in these families could have undergone convergent evolution 431 or ancestral traits could be strongly preserved. Either case would allow distantly-related 432 Polygonaceae and Fabaceae to interact with the same insects. Finally, this pattern could 433 be the result of ecological or environmental filtering (Ackerly, 2003; Mayfield et al., 2009). 434 Closely-related species with strong niche overlap might compete too severely to coex-435 ist. This is especially likely for plants sharing pollinators, where the loss of pollen to 436 related species might severely limit reproductive success (Levin & Anderson, 1970; Bell 437 et al., 2005; Mitchell et al., 2009). Indeed, animal pollination and seed dispersal have 438

been shown to act as filters for several plant clades (Mayfield *et al.*, 2009). Selection to avoid competition and restrict numbers of interaction partners may lead to more intimate or specialised interactions (Ponisio *et al.*, 2017). This is particularly the case in highly intimate interactions, where both partners may specialise (Hembry *et al.*, 2018). Past selection to avoid competition is consistent with the relatively high proportion of extreme specialists we observed in the pollination networks. As described above, these specialists likely weaken the relationship between phylogenetic distance and niche overlap.

The remaining families did not show significant relationships in either direction. That 446 is, the niche overlap between two plants did not vary linearly over phylogenetic distance. 447 Once again, there are several possible explanations for this result. These plants might be 448 highly specialised on different interaction partners and therefore have low overlap at all 449 levels of relatedness. In other plant families with more moderate levels of specialisation, it 450 is possible that pollination and/or herbivory do not exert large selection pressures on the 451 plants. If traits affecting pollination or herbivory are not heritable in these groups [Kursar 452 et al., 2009 or their phenotypes are constrained by other factors (e.g., environmental con-453 ditions, trade-offs with other traits, ontogenic change [Kariñho-Betancourt et al., 2015]), 454 then we should not expect a relationship between phylogenetic distance and overlap of 455 interaction partners. Alternatively, pollination and/or herbivory might exert large pres-456 sures that maintain the clade within a single pollination or defensive syndrome. These 457 syndromes are commonly believed to predict the pollinators or herbivores with which a 458 plant will interact (Waser et al., 1996; Fenster et al., 2004; Ollerton et al., 2009; Johnson 459 et al., 2014). As some recent studies have suggested that pollination syndromes do not 460 accurately predict plants' visitors in all plant families (Ollerton et al., 2009), it may be of 461 interest for future researchers to test whether syndromes are better predictors in families 462 with weak relationships between overlap and phylogenetic distance. 463

For those few families which were well-represented in *both* pollination and herbivory networks, we can also contrast the trends in the two network types. Notably, all families

except Asteraceae showed different trends in different network types. This could be be-466 cause of conflicting selection from pollinators and herbivores, with one type of selection 467 placing greater constraints on plant traits than the other. Multiple types of interactions 468 (e.g., pollination, herbivory, nectar robbing) and even environmental factors can influ-469 ence traits such as flower colour, nectar abundance, and flowering phenology (Strauss & 470 Whittall, 2006). These influences can act in the same or different directions (Strauss 471 & Whittall, 2006). Plant phenotypes in turn affect which species participate in both 472 pollination and herbivory (Strauss, 1997; Strauss et al., 2002; Adler & Bronstein, 2004; 473 Adler et al., 2006; Theis, 2006). The interplay between these different selective pres-474 sures may mean that plants cannot evolve to respond optimally to both pollinators and 475 herbivores. Put another way, stronger selective pressure from herbivores might cause 476 phenotypic changes that disrupt phylogenetic signal in pollinators, or vice versa. This 477 could result from asymmetric degree distributions: within a single system, most plants 478 tend to interact with many pollinators or many herbivores but not both (Melián et al., 479 2009; Pocock et al., 2012; Astegiano et al., 2017). These asymmetric interactions may also 480 affect higher-order network structures such as modularity or nestedness (Astegiano et al., 481 2017). The nature of the effects of multiple interaction types on both phylogenetic signal 482 in interactions and overall network structure is, however, still an open question deserving 483 of much more research. 484

Altogether, our study has revealed general trends for conservation of interaction part-485 ners between closely-related species, with some networks and plant families showing dif-486 ferent trends. This overall similarity between closely-related species has a potential appli-487 cation in ecological restoration. Close relatives could be used interchangeably to restore 488 missing interactions and fill ecosystem functions. This may be advantageous when a tar-489 get plant is more difficult to establish than its relatives, or if the restoration site is not 490 large enough to support viable populations of many species. We should urge caution, 491 however, since plants which support the same pollinators may also support similar sets 492

of herbivores. To avoid unwanted indirect effects, all interactions involving the target 493 species should be considered. Although here we considered only the presence or absence 494 of interactions, (i.e., qualitative networks) recent work also suggests that the phylogenetic 495 composition of a plant community can also affect the strength of interactions, and that the 496 spatial arrangement of plants within a community may be particularly important (Yguel 497 et al., 2011; Castagneyrol et al., 2014). These further nuances in the relationship between 498 phylogenetic distance and niche overlap could also strongly affect the ability of closely-499 related species to fill the same functions in restoration efforts. This is clearly a topic with 500 many unresolved questions, deserving of further study. 501

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510 Author Contribution

ARC, DBS, GVDR, and NJB designed the research. ARC, MO, IN, IMW, and JAT collected published data. ARC and GVDR performed the analyses. All authors contributed to the manuscript.

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Notes legends

Notes 1: Sources for networks. Original sources for all networks used in this study.

Notes 2: R implementation of tuple-form regression. Further details and R code for our statistical model.

Notes 3: Repeating our analyses with proportion of shared partners. Results when niche overlap is defined as the proportion of shared interaction partners rather than by a tuple of shared and unshared partners.

Notes 4: A test of connectance. Methods and results for a test as to whether the connectance of a network affects the relationship between phylogenetic distance and niche overlap.

Notes 5: Distributions of *p*-values for permuted networks. Figures showing the distributions of *p*-values in permuted networks referenced in Figs. 1-2.

Notes 6: Details of within-family regressions. Supplemental results for within-family regressions, including a comparison of *p*-values and critical values for the sequential correlated Bonferroni test.

Tables

Table 1: Change (Δ) in log odds (per million years of phylogenetic distance) of a pair of plants in the same family sharing a pollinator.

Family	$\Delta \log odds$	P-value	Family	$\Delta \log odds$	P-value
Adoxaceae	-65.8	0.163	Malvaceae	-5.56	0.363
A maryllidace a e	-17.9	0.015	$Melastomataceae^*$	5.19	0.577
Apiaceae	10.9	0.006	Montiaceae	-1.12	0.87
Apocynaceae	-6.96	0.037	Myrtaceae	8.55	0.071
A sparagaceae	-6.23	0.189	Oleaceae	0.995	0.855
Asteraceae*	-1.47	< 0.001	Onagraceae	-556	>0.999
Berberidaceae	-1.48×10^{3}	>0.999	Orchidaceae	-14.5	0.145
Boraginaceae	-5.15	< 0.001	Orobanchaceae	24.2	0.326
Brassicaceae	-11.2	0.072	Papaveraceae	-11.2	0.511
Calceolariaceae	156	0.998	Phyllanthaceae	9.99	0.433
Campanulaceae	334	0.999	Plantaginaceae	-8.48	0.001
Caprifoliaceae	0.31	0.959	$Poaceae^*$	69.2	0.003
Cary ophyllaceae	2.09	0.644	Polygonaceae	-14.8	< 0.001
Cistaceae	-11.4	< 0.001	Primulaceae	14.9	0.343
Convolvulaceae	-1.84	0.837	Ranunculaceae	-38	< 0.001
Ericaceae	4.61	0.116	Rosaceae	0.759	0.735
Fabaceae*	-12.9	< 0.001	$Rubiaceae^*$	-13	0.026
Gerania ceae	-3.31	0.624	Salicaceae	-1.9	0.545
Hydrangeaceae	0.057	0.982	Sapindaceae	821	0.999
Iridaceae	-27.9	0.078	Saxifragaceae	-0.092	0.992
Lamiaceae	-5.01	< 0.001	Solanaceae	-21.9	0.189
Lauraceae	-79.9	< 0.001	Tropaeolaceae	192	0.997
Loasaceae	-865	>0.999	Verbenaceae	-9.03	0.627
Malpighiaceae	2.8	0.168	Violaceae	-0.487	0.974

We were able to fit these models to 48 plant families (see *Materials and Methods* for details). Families marked with an asterisk were also sufficiently diverse to model in herbivory networks. Statistically significant values are indicated in bold.

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Family	$\Delta \log odds$	P-value
Asteraceae	-1.73	0.550
Euphorbiaceae	-19.2	< 0.001
Fabaceae	18.7	0.046
Melastomataceae	-13.2	0.022
Moraceae	-2.13	0.092
Nothof a gaceae	-595	>0.999
Pinaceae	-25.8	0.733
Poaceae	-4.50	0.020
Rubiaceae	-8.16	0.006

Table 2: Change (Δ) in log odds (per million years of phylogenetic distance) of a pair of plants in the same family sharing a herbivore.

Nine plant families were sufficiently diverse in our dataset to permit this analysis (see *Materials* and *Methods* for details). For each pattern of overlap, we show the change in log odds per million years and the associated *P*-value. Statistically significant values are indicated in bold.

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Figures

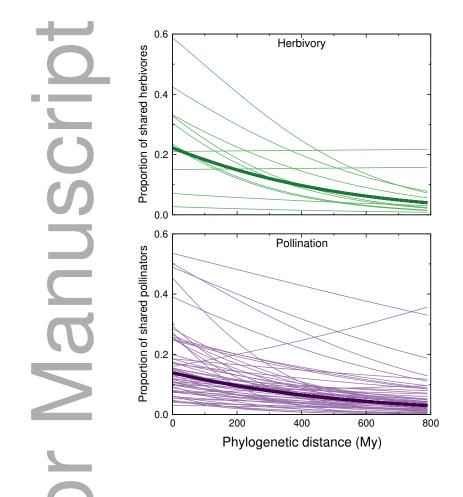


Figure 1: Results of a mixed-effects logistic regression of pairwise niche overlap against phylogenetic distance for plants in 11 herbivory networks (top; green) and 59 pollination networks (bottom; purple). In both network types, the probability of a pair of plants sharing an interaction partner decreased with increasing phylogenetic distance (thick, dark lines). There was substantial variation among networks (thin, pale lines) of both types. The slope of the regression for each network was significantly more extreme than that obtained from 999 permutatations of that network (slopes obtained from the permuted networks ranged between -1.34×10^{-12} and 9.19×10^{-13}).

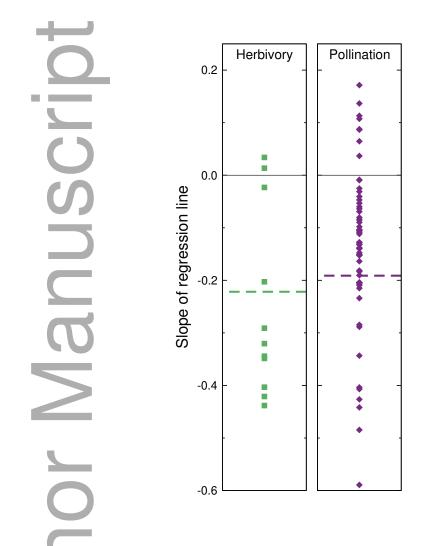


Figure 2: The slopes of the mixed-effect logistic regression of pairwise niche overlap against phylogenetic distance (representing the change in log odds of a pair of plants sharing an interaction partner) was significantly different from 0 for each network. Here we show the observed slopes for herbivory (green squares) and pollination (purple diamonds) networks. Thick, dashed lines represent the mean slopes across all networks of each type. The maximum and minimum slopes obtained from 999 permutations of each network are depicted by thin, black lines. For both network types, the slopes obtained from permuted networks were always very close to 0 (range -1.34×10^{-12} to 9.19×10^{-13}).

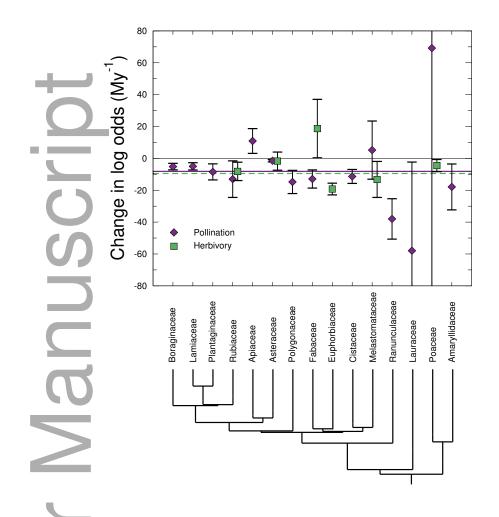
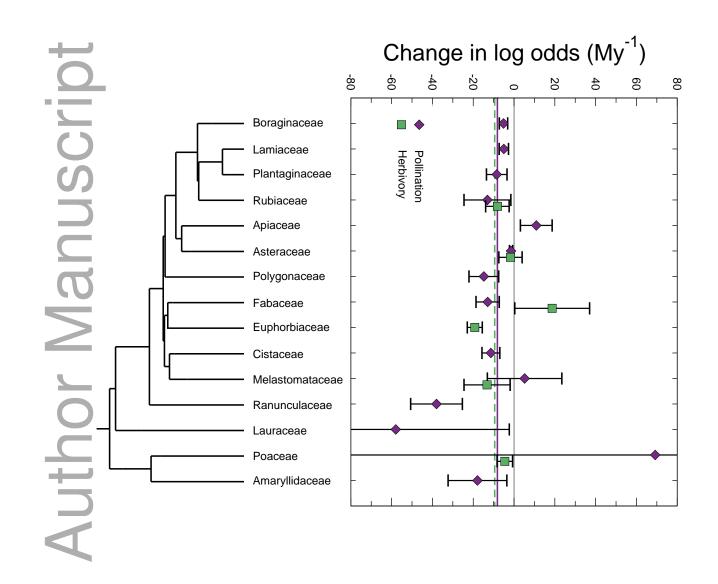
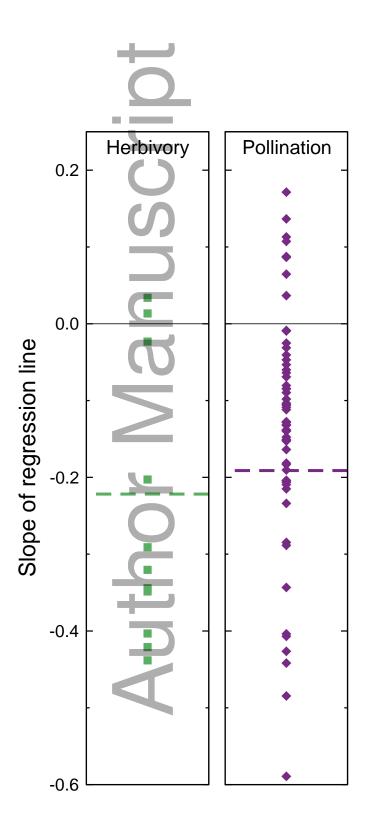
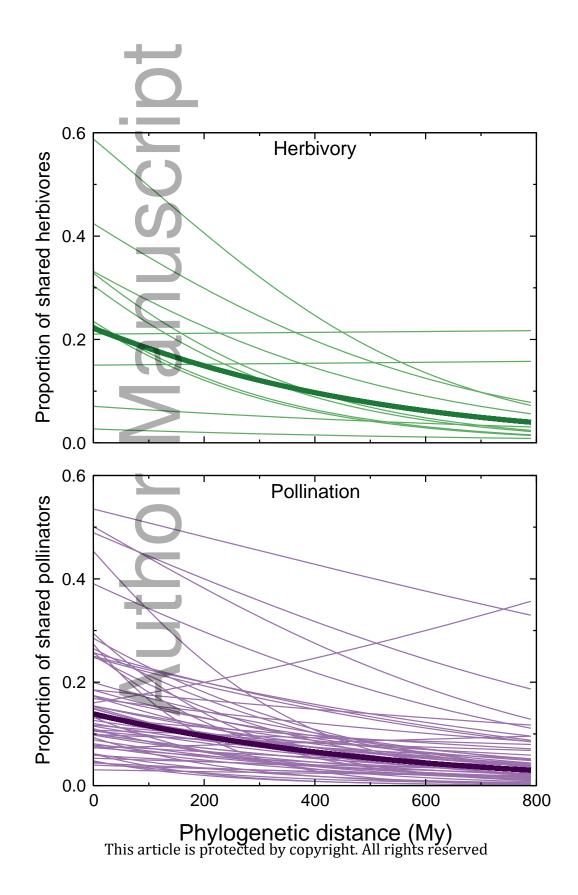


Figure 3: Change in the log odds of a pair of plants sharing a pollinator or herbivore (i.e., the slopes of the mixed-effect logistic regressions) as phylogenetic distance between the plants increases. These values are analogous to the slopes of the regression lines from Eq. 2-3 and represent the change in the probability of observing shared interaction partners per million years of divergence time. For clarity, we show only the 15 plant families for which the slope of the regression of the proportion of shared interaction partners against phylogenetic distance was significant in at least one network type. Note that the change in log odds for Asteraceae in herbivory networks and Melastomataceae in pollination networks are not significantly different from zero; we present these values only for comparison across network types. All other plant families were well-represented in only one network type. Families in pollination networks are indicated by dark purple diamonds while families in herbivory networks are indicated by pale green circles. We also show the slope of the relationship between the log-odds of observing each overlap pattern and phylogenetic distance across all plant families in herbivory (pale, green horizontal line) and pollination (dark, purple horizontal line) networks. The phylogenetic tree below the plots indicates the relatedness between these plant families. Error bars represent 95%confidence intervals. 38

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Author/s:

Cirtwill, AR; Dalla Riva, GV; Baker, NJ; Ohlsson, M; Norstrom, I; Wohlfarth, I-M; Thia, JA; Stouffer, DB

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