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# The macroecology of phylogenetically structured hummingbird-plant networks

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44       122 biogeography, specialization, macroecology, modularity, phylogenetic signal, pollination,  
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46          quantitative networks.  
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1      124 **ABSTRACT**

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3      **Aim** To investigate the association between species richness, species' phylogenetic signal,  
4      insularity and historical and current climate with hummingbird-plant network structure.  
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7      **Location** 54 communities along a c. 10,000 kilometer latitudinal gradient across the  
8      Americas (39°N - 32°S), ranging from sea level to c. 3700 m asl, located on the mainland and  
9      on islands, and covering a wide range of climate regimes.  
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12      **Methods** We measured null-modeled corrected complementary specialization and bipartite  
13      modularity (compartmentalization) in networks of quantitative interactions between  
14      hummingbird and plant species. Using an ordinary least squares multi-model approach, we  
15      examined the influence of species richness, phylogenetic signal, insularity, and current and  
16      historical climate conditions on network structure.  
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19      **Results** Phylogenetically-related species, especially plants, showed a tendency to interact  
20      with a similar array of partners. The spatial variation in network structure exhibited a constant  
21      association with species' phylogeny ( $R^2=0.18-0.19$ ). Species richness and environmental  
22      factors showed the strongest associations with network structure ( $R^2=0.20-0.44$ ;  $R^2=0.32-$   
23      0.45, respectively). Specifically, higher levels of complementary specialization and modularity  
24      were associated to species-rich communities and communities in which closely-related  
25      hummingbirds visited distinct sets of flowering species. On the mainland, warmer  
26      temperatures and higher historical temperature stability associated to higher levels of  
27      complementary specialization.  
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30      **Main conclusions** Previous macroecological studies of interaction networks have  
31      highlighted the importance of environment and species richness in determining network  
32      structure. Here, for the first time, we report an association between species phylogenetic  
33      signal and network structure at macroecological scale. Specifically, null model corrected  
34      complementary specialization and modularity exhibited a positive association with species  
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richness and a negative association with hummingbird phylogenetic signal, indicating that  
both high richness and high inter-specific competition among closely-related hummingbirds  
exhibit important relationships with specialization in hummingbird-plant networks. Our results  
document how species richness, phylogenetic signal and climate associate with network  
structure in complex ways at macroecological scale.

For Peer Review

1       154 **INTRODUCTION**

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3           An urgent challenge facing contemporary ecologists is to understand how ecological,  
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5       156 evolutionary and environmental mechanisms affect the structure and function of ecological  
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7           communities. Better understanding of how these processes form and maintain communities  
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9       158 should help us to construct more robust theories and models of community dynamics that  
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11           better predict how species and communities may respond to disturbance (Vázquez *et al.*,  
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13       160 2009a; Woodward *et al.*, 2010). For this purpose, the use of network approaches to study  
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15           complex communities of interacting species, i.e. networks representing species as nodes and  
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17           interactions as links between species, has significantly advanced our understanding of  
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19       162 ecological systems (Woodward *et al.*, 2010).

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21           For instance, research on mutualistic networks such as plant-pollinator or plant-  
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23           frugivore interactions, has revealed that these communities typically exhibit a number of  
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25           architectural attributes, such as complementary specialization and modularity, which differ  
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27           significantly from random interactions among species (Olesen *et al.*, 2007; Blüthgen 2010;  
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29       166 Dalsgaard *et al.*, 2011, 2013; Schleuning *et al.*, 2012, 2014; Trøjelsgaard & Olesen, 2013;  
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31           Dormann & Strauss, 2014; Sebastián González *et al.*, 2015). Complementary specialization  
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33       168 measures the exclusiveness in species' interactions. Specifically, it is a measure of the  
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35           deviation from a neutral scenario in which species interact solely according to their  
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37       170 availability, measuring species' availability either as its abundance or its interaction frequency  
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39           (Blüthgen *et al.*, 2006). Modularity quantifies whether species interact more frequently with  
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41           172 subsets of available species within a community, forming modules of densely interacting  
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43           plants and hummingbirds, with loose connections between them (Olesen *et al.*, 2007). Such  
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45           174 modules have been suggested to reflect co-evolutionary units (Olesen *et al.*, 2007), within-  
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47           network functional specialization (Maruyama *et al.*, 2014), and phenological units of species  
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49       176 in environments with strong climatic seasonalities (Martín González *et al.*, 2012; Schleuning  
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1 et al., 2014).

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3 180 These architectural attributes have a profound impact on the dynamics of ecological  
4 communities. For example, higher levels of complementary specialization have been linked to  
5 higher dependencies between species and hence, to a higher risk of secondary extinctions  
6 (Blüthgen, 2010), whereas high levels of modularity may at first provide higher network  
7 resilience to perturbations, as these are not typically spread across modules, but ultimately  
8 may result into unconnected modules, community fragmentation and rapid species loss  
9 (Thébault & Fontaine, 2010; Stouffer & Bascompte, 2011). Complementary specialization and  
10 modularity are typically related, as some degree of modularity necessarily implies a certain  
11 level of complementary specialization on a subset of species from the available species pool  
12 (Dormann & Strauss, 2014). Nevertheless, these metrics measure different aspects of  
13 interaction specialization and, hence, may show disparate associations with different  
14 ecological, evolutionary and environmental factors.

15 192 Thus, studies at macroecological scales, where differences among network structure  
16 can be associated to varying ecological, evolutionary and environmental factors, may provide  
17 valuable insights into the structure and dynamics of ecological networks (Dalsgaard et al.,  
18 2011, 2013; Schleuning et al., 2012, 2014; Trøjelsgaard & Olesen, 2013, Sebastián González  
19 et al., 2015). For instance, species richness may promote higher degrees of complementary  
20 specialization and modularity by increasing interspecific competition and providing more  
21 species to interact with, i.e. allowing a finer niche partitioning (e.g. Dalsgaard et al., 2011;  
22 Junker et al., 2013). Additionally, evolution may structure biotic interactions via niche  
23 conservatism, as closely-related species may display more similar phenotypes, spatial  
24 distributions, and ecological interactions than distantly-related species (Webb et al., 2002;  
25 Ives & Godfray, 2006; Rezende et al., 2007; Vázquez et al., 2009a). Alternatively, closely-  
26 related species experiencing strong interspecific competition may undergo niche partitioning,  
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1           204 competitive exclusion, or resource-use complementarity (Webb *et al.*, 2002; Rezende *et al.*,  
2           209; Krasnov *et al.*, 2012). Environmental setting may also affect biotic interactions.  
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4           206 Historical climatic stability may promote specialization and modularity by providing species  
5           209 more opportunity to co-evolve over longer periods of time compared to species from  
6  
7           208 climatically more unstable areas. For example, large late Quaternary glacial-interglacial  
8           211 climate change is coupled with a decreased modularity in pollination networks (Dalsgaard *et*  
9  
10           210 *al.*, 2013), and decreased specialization in hummingbird-plant networks (Dalsgaard *et al.*,  
11           213 2011). Likewise, current environmental conditions may influence the identity and strength of  
12           216 species interactions by altering species' spatial distribution, phenophases and foraging  
13           219 capability (Cruden, 1972; Martín González *et al.*, 2009; Dalsgaard *et al.*, 2011, 2013;  
14  
15           221 Schleuning *et al.*, 2012, 2014). For example, bird pollination has been reported to become  
16           224 increasingly important and specialized in areas of high precipitation and low temperatures,  
17           227 possibly because insect pollinator diversity and activity is lower in cold and wet environments  
18           230 due to thermoregulatory constraints (Cruden, 1972; Dalsgaard *et al.*, 2009, 2011; Martín  
19           233 González *et al.*, 2009). Finally, insularity may influence pollination network structure, with  
20           236 significantly lower levels of specialization and modularity on islands compared to the  
21           240 mainland (Olesen *et al.*, 2002; Dalsgaard *et al.*, 2013).

22           244 Despite the recently increasing interest in determining the influence of species  
23           247 richness, phylogenetic signal and past and current environmental conditions on network  
24           250 structure (Dalsgaard *et al.*, 2011, 2013; Schleuning *et al.*, 2012, 2014; Sebastián González *et*  
25           253 *al.*, 2015), very little is known about the relative importance of these factors. Only Schleuning  
26           256 *et al.* (2014) studied their combined influence on the modular patterns of mutualistic  
27           260 networks. Specifically, they used a dataset of 18 frugivore bird-plant networks, and showed  
28           263 that modularity was high in areas having low temperatures and high temperature seasonality,  
29           266 whereas species' phylogenetic history and historical climate exhibited no detectable  
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1 association to modularity. Here, we likewise examine how species richness, phylogenetic  
2 signal and environmental conditions associate to the structure of hummingbird-plant  
3 networks, another type of mutualistic assemblage. To do so, we compiled a data-set  
4 consisting of 54 high-resolution quantitative hummingbird-plant interaction networks, and  
5 combined them with state-of-the-art species phylogenies and simulations of paleo and  
6 current climate. In accordance with previous studies on pollination networks, we show that  
7 species richness and climatic conditions influence complementary specialization and  
8 modularity. Additionally, for the first time, we report an association between species  
9 phylogenetic signal and network structure at macroecological scale. Specifically, communities  
10 where closely-related hummingbirds interact with distinct sets of plant species exhibit higher  
11 levels of complementary specialization and modularity.

## 240 MATERIAL & METHODS

### 242 *Hummingbird-plant interaction networks*

32 We compiled a large data-set of 54 high-resolution and geographically widely distributed  
33 hummingbird-plant interaction networks from published and unpublished sources (Fig. 1; see  
34 Table S1 for full references, and Dalsgaard *et al.*, 2011 for a previous version of this dataset).

36 246 This dataset does not include networks which have not sampled the entire floral community,  
37 i.e. we discarded studies which had focused only on ornithophilous species or on specific  
38 floral or hummingbird groups. We also discarded illegitimate interactions, i.e. our database  
39 focuses on interactions with potential for pollination. These networks describe the interaction  
40 frequency between plants and hummingbirds, recorded as number of observed visits. In total,  
41 our dataset is composed of 141 species of hummingbirds belonging to all major hummingbird  
42 lineages (about 42% of all described species; McGuire *et al.*, 2014) and 824 plant species  
43 belonging to 79 different plant families (Appendix S1).

1       254       Mutualistic hummingbird-plant interaction networks constitute a particularly suitable  
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3       model system to explore large-scale patterns in network structure as there is ample  
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5       knowledge of the historical biogeography of hummingbirds, these assemblages are widely  
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7       distributed in a variety of ecosystems [across the Americas](#), and they offer examples covering  
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9       the entire interaction specialization-generalization spectrum (Bleiweiss, 1998; Dalsgaard *et*  
10      *al.*, 2011). Furthermore, our data consists of a single and monophyletic pollinator group  
11  
12      (Brown & Bowers, 1985; McGuire *et al.*, 2014), allowing us to explore the [association](#)  
13  
14      [between](#) phylogeny on overall network structure more precisely.  
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17      262       Finally, the spatial distribution of these networks extend from sea level to [c.](#) 3700 m  
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19      a.s.l. and spans a [c.](#) 10,000 kilometer gradient from 39°N to 32°S, from [tropical and](#)  
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21      264       subtropical ecosystems in South America to temperate communities in North America. The  
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23      spatial distribution includes both mainland and insular communities (Fig. 1), and consists of  
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33      268       *Network composition and network metrics*

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35      Network interactions were quantified in terms of the number of interaction events between  
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37      270       hummingbird and plant species at particular locations. For each network we calculated two  
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39      complementary metrics that describe the degree of specialization or specificity of  
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41      272       hummingbird-plant interactions: [complementary specialization](#) (Blüthgen *et al.*, 2006) and  
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43      quantitative bipartite modularity (Dormann & Strauss, 2014). These metrics were correlated in  
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45      274       our dataset  $R^2=0.78$  (Table S2) [but](#) were chosen to more fully characterize [how](#) evolutionary  
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47      and environmental [factors associate to](#) specialized network structures. For the calculations  
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49      276       we used the R package *bipartite* 1.20 (Dormann *et al.*, 2009).

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51      (1)       [Complementary specialization \( \$H\_2'\$ \)](#) describes how species restrict their interactions  
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53      278       [relative to random expectations based on species' abundances/interaction frequency.](#)  
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2 Complementary specialization is calculated as  
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$$H_2 = - \sum_{i=1}^r \sum_{j=1}^c (p_{ij} \ln p_{ij})$$

, where  $p_{ij}$  reflects the

5 proportional number of interactions of each species relative to their availability, i.e., for their  
6 respective marginal total (in our case total interaction frequency) for  $r$  plant and  $c$  animal  
7 species. We illustrate this in Figure 1; for instance, from the example community A from Fig.  
8  
9  
10 12,  $p_{1A}=0.99$  (which results from an interaction frequency of 155 between plant 1 and  
11  
12 13,  $p_{A1}=1$  (155 over 155).  
14  
15  
16 Hence the interaction A-1 exhibits a high complementary specialization, as hummingbird A  
17  
18 visits exclusively plant 1 and at a high frequency, while plant 1 is only visited, and at a very  
19  
20 low frequency, by another hummingbird species. If a community is composed majoritarily by  
21  
22 interactions with such high complementarity, the community will exhibit high values of  
23  
24 complementary specialization (Blüthgen et al., 2006). On the other hand, species C and 4  
25  
26 also interact 155 times, but in this case hummingbird C interacts with many other plant  
27  
28 species and has a higher marginal total, lowering the complementary specialization of this  
29  
30 pair ( $p_{C4}=0.36$ ,  $p_{4C}=1$ ).  
31  
32

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34 We standardized complementary specialization as  $H_2' = \frac{H_{2\max} - H_2}{H_{2\max} - H_{2\min}}$ , so that  $H_2'$   
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36

37 ranges from minimum ( $H_2'=0$ ) to maximum ( $H_2'=1$ ) link selectiveness where species establish  
38 distinct and highly specific interactions far different than expected interactions (Blüthgen et  
39  
40 al., 2006). Thus,  $H_2'$  quantifies the deviation of the observed interactions from those expected  
41 under a neutral assumption that species' interactions are entirely determined by partner  
42 availability. This assumption minimizes the influence of rare interactions by causing frequent  
43 interactions to dominate  $H_2'$ .  
44  
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46 300 (2) Quantitative bipartite modularity (QuanBiMo) is an algorithm which places species  
47 among an *a priori* unespecified number of modules, such that species interact at high  
48 frequencies within their module, and show few links and/or low frequency links with species  
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1 outside their module (Dormann & Strauss, 2014; Schleuning *et al.*, 2014; Fig. 1). Such  
2  
3 304 partition is based on a hierarchical representation of interaction frequencies and optimal  
4 allocation of species into modules. Specifically, the algorithm maximizes the bipartite version  
5  
6 306 of Newman's modularity ( $Q$ ), so that 
$$Q = \frac{1}{2N} \sum (A_{ij} - K_{ij}) \delta(m_i, m_j)$$
, where  $N$  reflects the total  
7 number of interactions,  $A_{ij}$  the normalized number of interactions between species  $i$  and  $j$ ,  $K_{ij}$   
8  
9 308 the expected interaction probability between species  $i$  and  $j$  drawn from a neutral model of  
10 interactions, and the indicator function  $\delta(m_i, m_j)$  equals 1 when species  $i$  and  $j$  are placed in  
11 the same module and 0 otherwise. Modularity ranges from no ( $Q=0$ ) to maximum ( $Q=1$ ).  
12  
13 310 We ran the QuanBiMo algorithm following the methodology established by  
14 Schleuning *et al.* (2014) and the default specifications of the `computeModules` function in  
15 bipartite, that is, for each network we chose the partition showing highest modularity from five  
16 independent runs of the algorithm (Dormann & Strauss, 2014; Schleuning *et al.*, 2014).  
17  
18 312 Variations in the likelihood values of modularity were negligible (all  $SD < 0.05$ ).  
19  
20 314  
21  
22 316 As raw values for network metrics may be affected by species frequencies and  
23 network connectance, network estimates for complementary specialization and modularity  
24 were corrected using null models (Schleuning *et al.*, 2012, 2014; Dormann & Strauss, 2014).  
25  
26 318 Null models simulated matrices with the same number of species and interactions as the  
27 empirical network, and with a species' interaction probability distribution drawn from observed  
28 species' connectivity (vaznull model in bipartite; Vázquez *et al.*, 2007). Corrected metrics  
29 were then calculated as the difference between the value of the empirical network and the  
30 mean value obtained from 1000 and 100 null models for  $H_2'$  and QuanBiMo, respectively (as  
31 in e.g. Schleuning *et al.*, 2012, 2014). As for the calculation of empirical QuanBiMo values,  
32 for each of the 100 null matrices we used the maximum value of five independent runs of the  
33 QuanBiMo algorithm (Schleuning *et al.*, 2014). By extracting the network structure achieved  
34 under null conditions, corrected metrics quantify how much an empirical community departs  
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$$Q = \frac{1}{2N} \sum (A_{ij} - K_{ij}) \delta(m_i, m_j)$$

1 328 from an average random one with an equivalent set of species, interactions and interaction  
2 probability distribution. Notice that for instance, as Figure 1 illustrates, a corrected value of  
3  
4 330 0.3 gives no information on the uncorrected values, only that there is a 0.3 difference  
5 between observed and null values, e.g. this value can result both from a raw observed value  
6  
7 332 of 0.4 and a mean null value of 0.1 (Fig. 1 community A), and from a raw observed value of  
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9 0.5 and a mean null value of 0.2 (Fig. 1 community B), etc.

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17 *Analysis of phylogenetic signal in bipartite interaction networks*  
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19 336 We measured the phylogenetic signal exhibited by hummingbird-plant networks by  
20 quantifying the degree to which closely-related species share more interaction partners than  
21  
22 338 distantly-related species (Ives & Godfray, 2006). Species' interactions are considered to  
23 exhibit a higher phylogenetic signal, when closely-related species share relatively more  
24  
25 340 interaction partners than distantly-related species (Ives & Godfray, 2006; Vázquez *et al.*,  
26  
27 2009b).

28  
29 342 We used state-of-the-art phylogenies of plants and hummingbirds to create variance-  
30 covariance matrices, which quantify the phylogenetic relatedness of plants or hummingbirds  
31  
32 344 in each community, using the "vcv" function from the R package *ape* (Paradis *et al.*, 2004).

33  
34 346 We fitted these vcv matrices to each observed bipartite interaction matrix through a linear  
35 model using the "pblm" function from the R package *picante* (Kembel *et al.*, 2010). This  
36  
37 348 analysis results in two independent measures of the strength of the phylogenetic signal, one  
38 for plants ( $d_{plants}$ ) and another for hummingbirds ( $d_{hummingbirds}$ ), together with an overall  
39  
40 348 measure of strength of the model fits for the entire community (measured as mean squared  
41 error of the model, MSE). We evaluated three different models, one assuming no  
42 phylogenetic signal ( $d_{plants}=d_{hummingbirds}=0$ ; Star model), one assuming a maximum  
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44 350 phylogenetic signal ( $d_{plants}=d_{hummingbirds}=1$ ; Brownian model), and a final one which

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1 incorporated the observed phylogenetic signals combined (estimated  $d_{plants}$  and  $d_{hummingbirds}$ ;

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4 354 **Data model).** We used the bootstrapping option to calculate confidence intervals for  $d_{plants}$

5 and  $d_{hummingbirds}$ . Networks where the these confidence intervals did not overlap zero or when

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7  
8 356 the  $MSE_{Data} < MSE_{Star}$  were considered to exhibit a significant phylogenetic signal (Ives &

9 Godfray, 2006; Vázquez *et al.*, 2009b). For a list of all species included in this study and a

10  
11 358 detailed explanation of the phylogenetic analysis, refer to Appendices S1 and S2,

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14  
15 respectively.

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17 360  
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19 *Environmental variables*

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21 362 We analyzed six variables describing contemporary and historical temperature and

22  
23 precipitation known or hypothesized to affect the structure of pollination networks. Four of the

24  
25 364 six variables describe contemporary climatic conditions: *mean annual temperature* (MAT, °C),

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27  
28 temperature seasonality (TS, standard deviation × 100), *mean annual precipitation* (MAP,

29  
30 366 mm) and *precipitation seasonality* (PS, coefficient of variation). We obtained these

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32 measurements from the WorldClim data set with spatial resolutions of 1 x 1 km

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34 368 (<http://www.worldclim.org>; see also Hijmans *et al.*, 2005). The historical climate variables, the

35  
36 velocity of temperature change and the velocity of precipitation change, reflect the speed of

37  
38 370 temperature and precipitation change between the Last Glacial Maximum (LGM) and pre-

39  
40 industrial times (VT, m/yr; VP, m/yr), following the definition of Loarie *et al.* (2009). Projections

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42 372 of the global climate during the LGM and pre-industry were generated by Hadley Centre

43  
44 Coupled Model Version 3 (HadCM3) with a resolution of 3.75 × 2.5 arc degrees (Singarayer

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46 374 & Valdes, 2010), and were downscaled to 0.1 × 0.1 arc degrees. For each study site, climate

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48 estimates were calculated as the average values of all 1 x 1 km grid cells (0.1 × 0.1 arc

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50 376 degrees for paleoclimate data) within a concentric distance of 10 km from the sampling

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52 location. Additionally, we scored whether a network was from the mainland (0) or an island

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1 378 (1).

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6 380 *Macroecological models*

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8 We used a multi-model approach based on information theory as outlined in Diniz-Filho *et al.*  
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10 382 (2008) to simultaneously evaluate the relationships between species richness, phylogenetic  
11 signal and environment and hummingbird-plant network structure. First, for each network  
12  
13 384 metric, we calculated full ordinary least squares (OLS) regression models which included the  
14 following ten predictor variables: 1) species richness (network size), 2) the phylogenetic  
15 signal in the interaction pattern of plants' ( $d_{plants}$ ) and 3) hummingbirds' ( $d_{hummingbirds}$ ), 4)  
16 annual average temperature, 5) temperature seasonality, 6) annual average precipitation, 7)  
17 precipitation seasonality, 8) temperature-change velocity, 9) precipitation-change velocity and  
18 10) insularity. Visually, OLS regression models minimize the distance between observed data  
19 points and those predicted by a linear approximation of the data. Second, for each network  
20 metric, we used the Akaike Information Criterion (AIC<sub>C</sub>) aiming at identifying minimum  
21 adequate models (MAMs) among all possible model combinations of our ten predictor  
22 variables. MAMs were defined as models exhibiting a differences in AIC<sub>C</sub> of at least two  
23 points lower than other models (i.e. a  $\Delta AIC_C < 2$ ; Burnham & Anderson, 2002; Diniz-Filho *et*  
24 *al.*, 2008). As no single MAM was identified, often around 10 models had  $\Delta AIC_C < 2$ , we  
25 instead used a multi-model approach. Specifically, instead of calculating regression  
26 coefficients in a single best model (MAM), we calculated the overall importance of each  
27 model ( $w_i$ ) as the relative likelihood of any given model  $i$  over the sum of the likelihoods of the  
28

$$w_i = \frac{\exp\left(-\frac{1}{2}\Delta_i\right)}{\sum_{r=1}^R \exp\left(-\frac{1}{2}\Delta_r\right)}$$

29 entire dataset of models, such that  $\Delta_i$  being  $\Delta_i$  the differences in AIC

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31 400 between the set of  $R$  models, so that the relative strength of each model depends on the  
32 entire set of models. We report the standardized regression coefficients and the overall  
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1       402 importance ( $\sum w_i$ ) of each variable for ordinary least squares OLS regression for an averaged  
2       model based on weighted  $w_i$  (Burnham & Anderson, 2002; Diniz-Filho *et al.*, 2008), for which  
3         
4       404 we adopted an importance cut-off value of  $\geq 0.750$ . The standardized regression coefficients  
5         
6       were also reported for the OLS regression model including all ten predictor variables (“full”  
7         
8       406 model). For each network metric, we used partial regressions to separate the total, unique  
9         
10      and shared variation explained by species richness, phylogenetic signal and environmental  
11        
12      408 factors in the “full” models.  
13        
14

15                  Previous studies suggest that the structure of mainland and island hummingbird-plant  
16                  and pollination networks may differ (Olesen *et al.*, 2002; Dalsgaard *et al.*, 2009, 2013), and  
17                  the colonization of the Caribbean by hummingbirds has been considerably more recent than  
18                  mainland America (c. 5 million years ago versus 12-22 million years ago for North and South  
19                  America, respectively). Consequently insular hummingbirds have had less time for  
20                  specialization and coevolution with their nectar plants than their mainland counterparts  
21                  (Bleiweiss, 1998; McGuire *et al.*, 2014). Together with the higher levels of generalization and  
22                  less modularity predicted for insular pollinator communities (Olesen *et al.*, 2002; Dalsgaard *et*  
23                  al., 2013), and a high level of periodic disturbances characteristic of Caribbean communities  
24                  which may hinder high levels of specialization on islands (Graves & Olsen, 1987; Rivera-  
25                  Marchand & Ackerman, 2006), the relationship between species richness, phylogenetic  
26                  signal and environmental factors and network structure may differ between mainland and  
27                  insular communities. For instance, the influence of historical climate change may be weaker  
28                  on islands than on the mainland (Dalsgaard *et al.*, 2013, 2014). Our dataset contains too few  
29                  island networks (n=9) to allow for a separate analysis for insular networks, hence, we  
30                  explored putative differences in mainland and island networks by analyzing a subset of the  
31                  dataset composed exclusively by networks from the mainland (Mainland, n=45 networks) and  
32                  comparing these results to those of the entire dataset (Global dataset, which includes both  
33                  Mainland and Island networks, n=54 networks). We also explored the effect of including or  
34                  excluding the Global dataset on the results of the regression analysis. We found no significant  
35                  differences in the results obtained with the Global dataset and the Mainland dataset, and  
36                  therefore we present the results obtained with the Global dataset. We also explored the effect of  
37                  including or excluding the island networks on the results of the regression analysis. We found no  
38                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
39                  and therefore we present the results obtained with the Global dataset. We also explored the effect of  
40                  including or excluding the island networks on the results of the regression analysis. We found no  
41                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
42                  and therefore we present the results obtained with the Global dataset. We also explored the effect of  
43                  including or excluding the island networks on the results of the regression analysis. We found no  
44                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
45                  and therefore we present the results obtained with the Global dataset. We also explored the effect of  
46                  including or excluding the island networks on the results of the regression analysis. We found no  
47                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
48                  and therefore we present the results obtained with the Global dataset. We also explored the effect of  
49                  including or excluding the island networks on the results of the regression analysis. We found no  
50                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
51                  and therefore we present the results obtained with the Global dataset. We also explored the effect of  
52                  including or excluding the island networks on the results of the regression analysis. We found no  
53                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
54                  and therefore we present the results obtained with the Global dataset. We also explored the effect of  
55                  including or excluding the island networks on the results of the regression analysis. We found no  
56                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
57                  and therefore we present the results obtained with the Global dataset. We also explored the effect of  
58                  including or excluding the island networks on the results of the regression analysis. We found no  
59                  significant differences in the results obtained with the Global dataset and the Mainland dataset,  
60                  and therefore we present the results obtained with the Global dataset.

1 mainland and insular communities, n=54 networks).

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3  
4 428 For all macroecological models, we log<sub>10</sub>-transformed species richness, temperature  
5 seasonality, temperature velocity, and precipitation velocity; we squared temperature, and  
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7  
8 430 square-root transformed precipitation. All other variables were left untransformed. In all  
9 spatial models we tested whether significant positive spatial autocorrelation remained in  
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11  
12 432 model residuals of the “full” models (i.e., whether p<0.05 in all distance classes, tested using  
13  
14 10 equally-spaced distance classes and applying a permutation test with 10,000 iterations).  
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17 434 As no positive spatial autocorrelation was observed we did not build more sophisticated  
18 spatial models. All regression analyses were conducted using the software Spatial Analysis in  
19  
20  
21 436 Macroecology, SAM 4.0 (Rangel et al., 2010).

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25 To better support our findings, we performed a number of complementary  
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28 438 macroecological models which included sampling effort as a predictor variable (Appendix  
29  
30 S3), observed (uncorrected) metrics of network structure instead of null model corrected  
31  
32 440 metrics (Appendix S4), and climate anomaly as a measure of historical climate stability  
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34 instead of climate change velocity (Appendix S5).

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36 442  
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39 **RESULTS**

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41 444 *Phylogenetic signal on species interaction patterns*

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43 Values for the independent phylogenetic signals of plants ( $d_{plants}$ ) and hummingbirds

44 446 ( $d_{hummingbirds}$ ) were low but above zero in 85 and 65% of the networks, respectively,

45  
46 suggesting a significant relationship between phylogenetic signal and species interaction

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48 448 patterns. The confidence intervals (CI) of  $d_{hummingbirds}$  overlapped with zero in a higher number  
49  
50 of networks than did CIs of  $d_{plants}$  (82 and 57%; Table in Appendix S2), indicating that the  
51  
52 significant association between phylogeny and interaction patterns was exhibited mostly by  
53  
54 plant species.

1       452       We examined the overall association between phylogeny and the structure of bipartite  
2                   interaction networks by comparing the mean squared error (MSE) of the model fit among  
3  
4       454       models adjusted after observed phylogenetic signal ( $MSE_{data}$ ), models which assumed no  
5                   phylogenetic signal ( $MSE_{star}$ ), and models which assume a maximum signal ( $MSE_{Brownian}$ ).  
6  
7

8       456       Results showed that most networks exhibited a significant phylogenetic signal (in 52  
9                   networks  $MSE_{data} < MSE_{star}$ , in 1 networks  $MSE_{data} = MSE_{star}$ , and in 1  $MSE_{data} > MSE_{star}$ ; in all  
10                  cases  $MSE_{Brownian}$  had clearly the highest values; Table in Appendix S2), e.g. in most  
11                  communities the model which fitted best to the data (the one having the lowest error) was  
12                  achieved when adjusting the model with observed phylogenetic signals of plants and  
13                  hummingbirds ( $MSE_{data}$ ). Nevertheless, most differences between  $MSE_{data}$  and  $MSE_{star}$  were  
14                  small, and much smaller than between  $MSE_{data}$  and  $MSE_{Brownian}$ , indicating that in general, the  
15                  association between phylogenetic signal and species interaction pattern was weak.  
16  
17

### 28       464 29 30                  *Macroecological models*

31       466       Full OLS models on corrected network metrics were able to account for 54%-62% of the  
32                  observed variation in network metrics across the sampled communities (Table 1, Table S3,  
33  
34       468       Fig. 3). Species richness and environmental factors showed the strongest association with  
35                  network structure, although with varying strengths across the "Global" and "Only Mainland"  
36  
37       470       datasets and network metrics (Table 1, Table S3, Figure S1). On the contrary, the  
38                  associations between phylogenetic signal and network structure remained constant,  
39  
40       472       exhibiting low regression coefficients in both metrics and datasets (Table 1, Table S3).  
41  
42                  Collectively, complementary specialization exhibited a stronger association with current  
43  
44       474       temperature and historical temperature stability, and to a lesser extent with species richness;  
45  
46                  whereas species richness had the strongest association with modularity (Fig. 3; Table S3).  
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49       476       When the ten predictor variables included in the full OLS models were examined in  
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1 detail, species richness and hummingbird phylogenetic signal were the only variables  
2  
3 478 showing an important, spatially consistent and widespread association to network structure,  
4 emerging as highly important for both network metrics and datasets. Species richness  
5  
6 480 showed a strong positive relationship with complementary specialization and modularity  
7 (Table 1). On the contrary, a higher phylogenetic signal among hummingbirds was related  
8 with a lower complementary specialization and level of modularity. In other words, despite the  
9 association between phylogenetic signal and species' interaction pattern was weak and the  
10 weighted regression coefficients of  $d_{hummingbirds}$  were low, complementary specialization and  
11 modularity consistently increased when closely-related hummingbirds visited distinct arrays  
12  
13 482 with a lower complementary specialization and level of modularity. In other words, despite the  
14 association between phylogenetic signal and species' interaction pattern was weak and the  
15 weighted regression coefficients of  $d_{hummingbirds}$  were low, complementary specialization and  
16 modularity consistently increased when closely-related hummingbirds visited distinct arrays  
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18 484 with a lower complementary specialization and level of modularity. In other words, despite the  
19 association between phylogenetic signal and species' interaction pattern was weak and the  
20 weighted regression coefficients of  $d_{hummingbirds}$  were low, complementary specialization and  
21 modularity consistently increased when closely-related hummingbirds visited distinct arrays  
22  
23 486 of plant species (Table 1, Table S3, Fig. 3).

24 Complementary specialization was also highly associated with the environmental  
25

26 488 conditions, as the total variation explained by environmental factors was consistently higher  
27 than the variation attributed to species richness and phylogenetic signal in both the "Global"  
28 and "Only Mainland" datasets, and approximated to the variation explained by the full models  
29 (Table S3, Fig. 3). The amount of unique variation (i.e. variation explained exclusively by the  
30 predictor-type) explained by environmental factors was similar to the unique variation  
31 explained by species richness on the Global dataset (Table S3, Fig. 3). However, when  
32 examining only mainland networks, the unique variation explained by environmental factors  
33 was two times greater than the unique variation explained by the other two predictor-types  
34 (Table S3, Fig. 3). Among the seven environmental variables examined, only current  
35 temperature and historical temperature stability showed a strong relationship with  
36 complementary specialization, with increasing complementary specialization in communities  
37 with current warmer temperatures and with a higher historical temperature stability (Table 1).

38  
39 496 Quantitative bipartite modularity showed the strongest association with species  
40

41 498 richness, particularly in the global dataset, with environmental conditions having a similar  
42

1 502 strength association among mainland communities (Table S3, Fig. 3).

2  
3 Shared variability among species richness and phylogenetic signal (i.e. variability  
4  
5 explained by these two predictor types) was non-existent, i.e. richness and phylogenetic  
6 signal were distinctly associated to specialization and modularity. Shared variability between  
7 environmental factors and species richness or phylogenetic signal was low (Table S3).

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10  
11 The remaining five environmental variables (insularity, temperature seasonality, mean  
12 annual precipitation, precipitation seasonality, and precipitation change velocity) as well as  
13 phylogenetic signal in the interaction pattern of plants, showed no important associations with  
14  
15 null model corrected network structure (Table 1).

## 21 22 23 512 DISCUSSION

24  
25 As previous macroecological studies of interaction networks, we found complementary  
26 specialization and modularity to vary along with environmental conditions and species  
27 richness. Additionally, for the first time, we report an association between phylogenetic signal  
28 and network structure at macroecological scale. Specifically, species richness and  
29 phylogenetic signal in hummingbird interaction patterns were the two predictor variables that  
30  
31 associated most consistently to network structure, with an association between  
32 complementary specialization and current and historical temperature conditions limited to  
33 mainland networks.

34  
35 Hummingbird phylogenetic signal had a constant association with both complementary  
36 specialization and modularity, and on both the Global and Only Mainland datasets, although  
37 the variability associated with phylogenetic signal was the lowest of all predictor types. A  
38 weak but significant relationship between phylogenetic signal and species' interaction pattern  
39 within ecological networks has also been reported by previous studies on food webs, host-  
40 parasite networks or plant-pollinator networks (Ives & Godfray, 2006; Vázquez *et al.*, 2009b;

1 Krasnov *et al.*, 2012; Rafferty & Ives, 2013). These studies also identified asymmetries in the  
2 phylogenetic signal between trophic levels, with the association between phylogenetic  
3 relatedness and species interaction pattern being stronger at lower trophic levels, i.e. plants  
4 in our system (but see Rezende *et al.*, 2007). According to these studies, such asymmetries  
5 may stem from a differential relationship between phylogeny and the interaction pattern of  
6 each trophic group, as species from the higher trophic group (“consumer” species) are more  
7 likely to adjust their feeding behavior according to local conditions. Our results corroborate  
8 this hypothesis: hummingbird phylogenetic signal showed a weaker relationship with their  
9 interaction pattern than plant species ( $d_{hummingbirds}$  tends to be lower than  $d_{plants}$ , and CI  
10  $d_{hummingbirds}$  overlap with zero in more networks). Nevertheless, and unlike for plants which  
11 showed no association to the spatial variation in network structure, hummingbird phylogenetic  
12 signal was associated to network structure, with higher levels of complementary  
13 specialization and modularity consistently achieved when closely-related hummingbird  
14 species visited distinct sets of flowering plant species. This suggests that resource  
15 partitioning and inter-specific competition among closely-related hummingbirds might play an  
16 important role in structuring interactions in hummingbird-plant networks. Indeed, inter-specific  
17 competition is known to be strong among hummingbirds, and has been noted as a potential  
18 driver of patterns of hummingbird biodiversity (Brown & Bowers, 1985; Bleiweiss, 1998;  
19 Cotton, 1998).

20 546 Species richness had the strongest association with both metrics and in both the  
21 Global and Only Mainland datasets except for complementary specialization in the mainland.  
22

23 548 Higher species richness may result in a higher niche availability, thus providing ample  
24 opportunities for biotic specialization. At the same time, higher species richness may increase  
25 inter-specific competition, which may explain why high species richness lead to higher levels  
26 of complementary specialization and modularity (Rezende *et al.*, 2009; Dalsgaard *et al.*,  
27 550 2011; Dalsgaard *et al.*, 2013).

1 552 2011; Krasnov *et al.*, 2012; Junker *et al.*, 2013). Moreover, a higher plant richness may also  
2 translate into a temporally-stable availability of floral resources (or the establishment of a  
3 constant minimum local flower supply), enabling a locally constant hummingbird population  
4 and, hence, potential for biotic specialization (Montgomerie & Gass, 1981; Stiles, 1985;  
5 Araujo & Sazima, 2003; Cotton, 2007; Abrahamczyk *et al.*, 2011).

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In the mainland, higher levels of complementary specialization were also found in warmer areas and in areas with higher historical temperature stability. The former may again be linked to higher inter-specific competition in warmer areas, and particularly in mainland settings, where hummingbird species and phylogenetic richness is highest (Bleiweiss, 1998; McGuire *et al.*, 2014). This relationship might be stronger on the mainland as insular pollinator faunas are typically depauperate (Olesen *et al.*, 2002). Moreover, in historically stable climates, hummingbirds have been able to establish long-term associations with plants, which may lead to higher levels of local adaptation, interaction specificity and specialization (see Dalsgaard *et al.*, 2011 for a study on a subset of the networks from the dataset used here). Such long associations may lead to higher complementary specialization in areas with low temperature-change velocity, such as montane habitats and tropical environments, leaving more generalized interaction patterns to areas with low topography and especially at higher latitudes where changes in temperature velocity have been greatest (Sandel *et al.*, 2011). Moreover, a higher complementary specialization in areas of high historical temperature stability is consistent with general hummingbird historical biogeography and speciation patterns, as the core area of hummingbird speciation and diversity occurs in Andean highland and in tropical lowland South-America (Brown & Bowers, 1985; Bleiweiss, 1998; McGuire *et al.*, 2014). We note that complementary macroecological analysis presented in Appendices S3-S5 generally confirmed the importance of species richness and hummingbird phylogenetic signal, and, additionally, showed an association between

1 observed (uncorrected) complementary specialization/modularity and insularity and  
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3 578 precipitation, with areas of high precipitation and mainland communities showing higher  
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5 observed complementary specialization and modularity. Precipitation has previously been  
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7 580 shown to influence both phenotypcal and network specialization of hummingbird-plant  
8 interactions and of entire pollination communities (Cruden, 1972; Dalsgaard *et al.*, 2009;  
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10 582 Martín González *et al.*, 2009). The fact that precipitation and insularity only associated with  
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12 observed complementary specialization and observed modularity and not with null model  
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14 584 corrected values of these network metrics, may indicate that insularity and precipitation do  
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16 not directly influence the interaction pattern of species, but rather associate indirectly with  
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18 586 network structure through species richness and/or interaction probability distributions.  
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26 588 **CONCLUSIONS**  
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29 Characterizing potential ecological, evolutionary and environmental mechanisms **associated**  
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31 590 **with** the structure of ecological communities is a critical first step towards understanding the  
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33 determinants of community assembly and how climate-change may affect biodiversity  
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35 (Woodward *et al.*, 2010; Schleuning *et al.*, 2014). By examining null model corrected network  
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37 metrics, we are actively investigating how much observed communities depart from random  
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39 594 ones with an equivalent set of species and interactions. Hence, we were able to investigate  
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41 not only whether a community presents a structure **which differs significantly from random**,  
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43 596 but also to associate the difference in complementary specialization and modularity to  
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45 species richness, evolutionary history, **and environmental conditions**.  
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48 598 We have shown that **complementary specialization and modularity** in hummingbird-  
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50 plant **networks associate to** species richness, hummingbird phylogenetic signal and  
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52 600 environmental factors acting at varying spatio-temporal scales and in different aspects of  
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54 network structure. Notably, species richness and hummingbird phylogenetic signal showed a  
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1 602 consistent association with network structure, with a more restricted but still important role of  
2 contemporary temperature and historical temperature stability, which were only important  
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4 604 among mainland communities. These results are markedly different to the ones for avian  
5 seed-dispersal networks for which species' phylogeny and historical climatic stability were  
6  
7 606 unrelated to modularity and complementary specialization (Schleuning *et al.*, 2012, 2014;  
8 Sebastián González *et al.*, 2015). Such difference suggests that hummingbirds and flowers  
9 engage in tighter co-evolutionary associations than frugivore birds and their plants, and  
10 hence, historical and evolutionary factors may have a stronger role in hummingbird-plant  
11 assemblages than for frugivore birds-plant assemblages.  
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1 the Center for Macroecology, Evolution and Climate.  
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19 **Biosketch:** This work results from an ongoing collaboration among numerous researchers  
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21 762 from different scientific backgrounds and institutions, who have gathered an extensive  
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23 dataset of hummingbird plant interaction networks and complementary data on species traits,  
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25 764 phylogeny, climate, etc. Such multidisciplinary information is allowing us to study in depth  
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27 different aspects of hummingbird-plant interaction patterns at macroecological scales, and to  
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29 766 identify potential mechanisms responsible for them. Our ultimate goal is to describe general  
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31 macroecological patterns in biotic interactions using hummingbird-plant interactions as a  
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33 768 model system.  
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770    **Table 1.** Relationship between species richness, phylogenetic signal and environment, and complementary specialization ( $H_2'$ ) and modularity (QuanBiMo) across the Americas (Global dataset) and the mainland, i.e. when excluding  
4 insular communities (Only Mainland). Complementary specialization measures the exclusiveness of the interactions in the community, whereas modularity quantifies whether species interact more frequently with subsets of available  
5 species within a community. For each network metric, the standardized regression coefficients are reported for ordinary least squares regression for a model including all variables ("Full") and for a model averaged across all possible  
6 models using Akaike Weights (AICc  $w_i$  - "Averaged"). For the averaged models, the relative importance of each predictor variable (" $\Sigma w_i$ ") is given by the Akaike Weights (AICc  $w_i$ ). Predictors with high importance are marked in bold.  
7 Phylogenetic signal among plants and hummingbirds,  $d_p$  and  $d_h$ , respectively; Mean annual temperature, MAT; Temperature seasonality, TS; Mean annual precipitation, MAP; Precipitation seasonality, PS; Velocity of temperature change  
8 since the last Glacial Maximum, MAT velocity; Velocity of precipitation change since the last Glacial Maximum, MAP velocity. The significance of Moran's  $I$  was tested with 10 distance classes and a permutation test with 10,000  
9 iterations. In all models multicollinearity was not an issue, i.e. VIF  $\leq 3.4$ , CN  $\leq 4.0$ , and there was no positive spatial autocorrelation, i.e. Moran's  $I$  was non-significant.  
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			Species richness			Phylogenetic signal			Environmental factors							
			Model	AICc	R <sup>2</sup>	Network size	d <sub>p</sub>	d <sub>h</sub>	Insularity	MAT	TS	MAP	PS	MAT velocity	MAP velocity	
Global dataset (54 networks)	$H_2'$	Full	-54.368	0.543	<b>0.461</b>		-0.087	<b>-0.303</b>	-0.222	0.284	0.176	0.045	-0.209	-0.475	0.190	
		Averaged			<b>0.520</b>		-0.143	<b>-0.361</b>	-0.121	0.137	0.121	0.085	-0.217	-0.270	0.106	
		$\Sigma w_i$			<b>0.999</b>		0.380	<b>0.962</b>	0.283	0.308	0.311	0.259	0.603	<b>0.671</b>	0.290	
	QuanBiMo	Full	-99.867	0.638	<b>0.638</b>		-0.145	<b>-0.232</b>	-0.187	0.188	0.081	-0.003	-0.204	-0.132	-0.098	
		Averaged			<b>0.652</b>		-0.183	<b>-0.283</b>	-0.137	0.080	0.058	0.097	-0.184	-0.098	-0.134	
		$\Sigma w_i$			<b>1.000</b>		0.619	<b>0.915</b>	0.341	0.257	0.237	0.286	0.567	0.288	0.392	
	Only Mainland (45 networks)	$H_2'$	Full	-55.244	0.581	<b>0.326</b>		-0.042	<b>-0.332</b>	na	<b>0.477</b>	0.125	-0.085	-0.261	<b>-0.714</b>	0.307
		Averaged			<b>0.324</b>		-0.048	<b>-0.358</b>	na	<b>0.451</b>	0.110	-0.044	-0.241	<b>-0.653</b>	0.359	
		$\Sigma w_i$			<b>0.807</b>		0.198	<b>0.919</b>	na	<b>0.854</b>	0.258	0.227	0.606	<b>0.931</b>	0.776	
		QuanBiMo	Full	-76.885	0.567	<b>0.539</b>		-0.174	<b>-0.283</b>	na	0.289	0.046	-0.079	-0.248	-0.289	-0.093
		Averaged			<b>0.544</b>		-0.184	<b>-0.300</b>	na	0.231	0.021	0.096	-0.232	-0.237	-0.127	
		$\Sigma w_i$			<b>0.999</b>		0.476	<b>0.805</b>	na	0.414	0.201	0.257	0.592	<b>0.427</b>	0.304	

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4 776 **Figure 1.** Cartoon illustrations depicting two different communities and the characteristics  
5 and relationship between complementary specialization ( $H_2'$ ) and quantitative bipartite  
6 modularity (QuanBiMo). Both communities depict 700 interaction events between 10 plants  
7 and 5 pollinators. Plants are labeled as numbers 1-10 in the matrix and as yellow nodes in  
8 the network representation, and pollinators by letters A-E in the matrix and as blue nodes in  
9 the network. Differences in interaction frequencies between plants and pollinators are  
10 illustrated as numbers of interaction events in the matrix format and as varying line widths in  
11 the network illustration. Complementary specialization measures the exclusiveness in  
12 species' interactions, whereas modularity quantifies whether species interact more frequently  
13 with subsets of available species within a community. Community A exhibits a moderate  
14 complementary specialization, with different species-pairs exhibiting various degrees of  
15 complementary specialization. For instance, species-pair A-1 shows a high complementary  
16 specialization, as hummingbird A visits exclusively and with high frequency (155 times) plant  
17 1, which in turn is only visited once by another hummingbird. On the other hand, the  
18 complementary specialization or pair E-4 is lower than for the previous pair despite these  
19 species interact with the same frequency because hummingbird E also interacts with other  
20 plant species, i.e. the interaction E-4 is less exclusive. Interactions in community B are also  
21 somewhat specialized (species favored interactions with few of the available partners), but  
22 the exclusiveness of these interactions is lower than for community A. Both communities  
23 show very similar values for corrected modularity and can be divided into three distinct  
24 modules, although the modules are composed of a different array of species. By having  
25 different degree of complementary specialization but similar values of modularity, these  
26 communities show that although these metrics are positively related, measure  
27 complementary but different aspects of specialization. Network drawings were created using  
28 Network3D and energized with the 3D Force-directed algorithm to enhance visualization of  
29 modularity patterns (Yoon et al. 2004, Williams 2010).

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4 802 **Figure 2.** Map of the American continent showing the location of the 54 study sites and a  
5 number of example networks located along a species richness gradient. Localities with  
6 darker shades of green denote networks with a higher richness. Some networks have been  
7 slightly moved horizontally to maximize visualization (exact coordinates of the localities on  
8 Table S1). For each illustrated network, the reference number and a concise description of  
9 the vegetation type is given, along with a network drawing. For the network drawings, blue  
10 and yellow nodes depict hummingbird and plant species, respectively, while line width  
11 depicts log+1 frequency of interaction among species. Notice that species-rich networks in  
12 general present more complex structures, with networks 11, 21 and 50 exhibiting the lowest  
13 corrected complementary specialization; networks 50, 11 and 21 the lowest corrected  
14 modularity; networks 53, 12 and 4 the highest corrected complementary specialization; and  
15 networks 10, 53 and 4 the highest corrected modularity, respectively. Complementary  
16 specialization measures the exclusiveness in species' interactions, whereas modularity  
17 quantifies whether species interact more frequently with subsets of available species within a  
18 community. Network drawings were created using Network3D and energized with the 3D  
19 Force-directed algorithm to enhance visualization of modularity patterns (Yoon et al. 2004,  
20 Williams 2010). The grey shading of the background illustrates altitude, with darker shades  
21 depicting higher altitudes.  
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802 **Figure 3.** Coefficients of determination ( $R^2$ ) for complementary specialization and modularity  
804 obtained from partial regression of full models, i.e. in models including all ten predictor  
806 variables (see Table 1 for standardized coefficients of each variable and more details of  
808 model fit, and Table S3 for the  $R^2$  values used in this figure). Complementary specialization  
810 measures the exclusiveness in species' interactions, whereas modularity quantifies whether  
812 species interact more frequently with subsets of available species within a community. We  
814 represent values for all networks in the study (Global dataset; n=54) and excluding insular  
816 communities (Only Mainland; n=45). Bars illustrate the association between the different

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3 828 "predictor-types" and network structure. Predictor-types refers to (a) species richness (one  
4 variable), (b) phylogenetic signal (two variables: phylogenetic signal in the interaction pattern  
5  
6 830 of plants and hummingbirds), and (c) environmental factors (eight variables: insularity,  
7 average annual temperature, temperature seasonality, total annual precipitation, precipitation  
8  
9 832 seasonality, temperature and precipitation-change velocity between the Last Glacial  
10 Maximum and the present). The amount of variation explained by each pooled predictor-type  
11  
12 834 is color-coded with different shades of blue: bars colored in darkest blue depict the overall  
13 variation explained by all factors together, whereas medium-blue illustrate the total variation  
14  
15 836 explained by that predictor-type, and light-blue the unique variation explained by each  
16 predictor-type and not shared by other variable-types.  $H_2'$ , complementary specialization;  
17  
18 838 QuanBiMo, quantitative bipartite modularity.

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Figure 1.

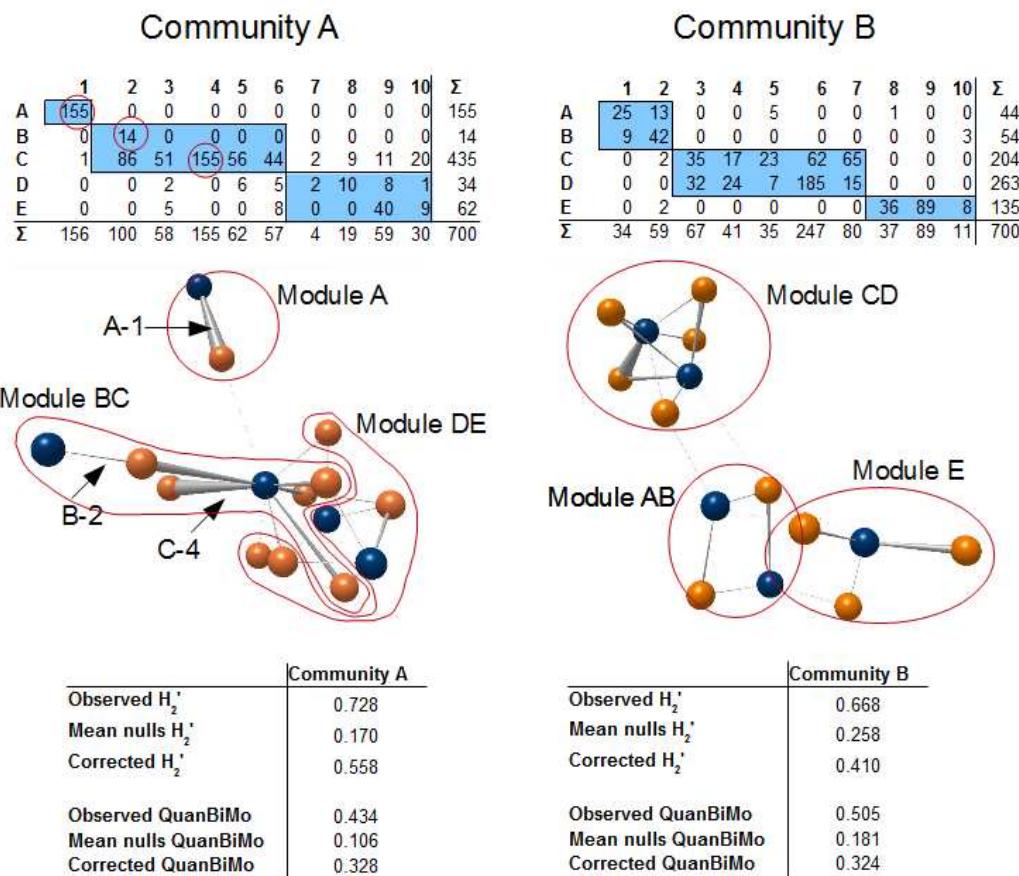


Figure 2.

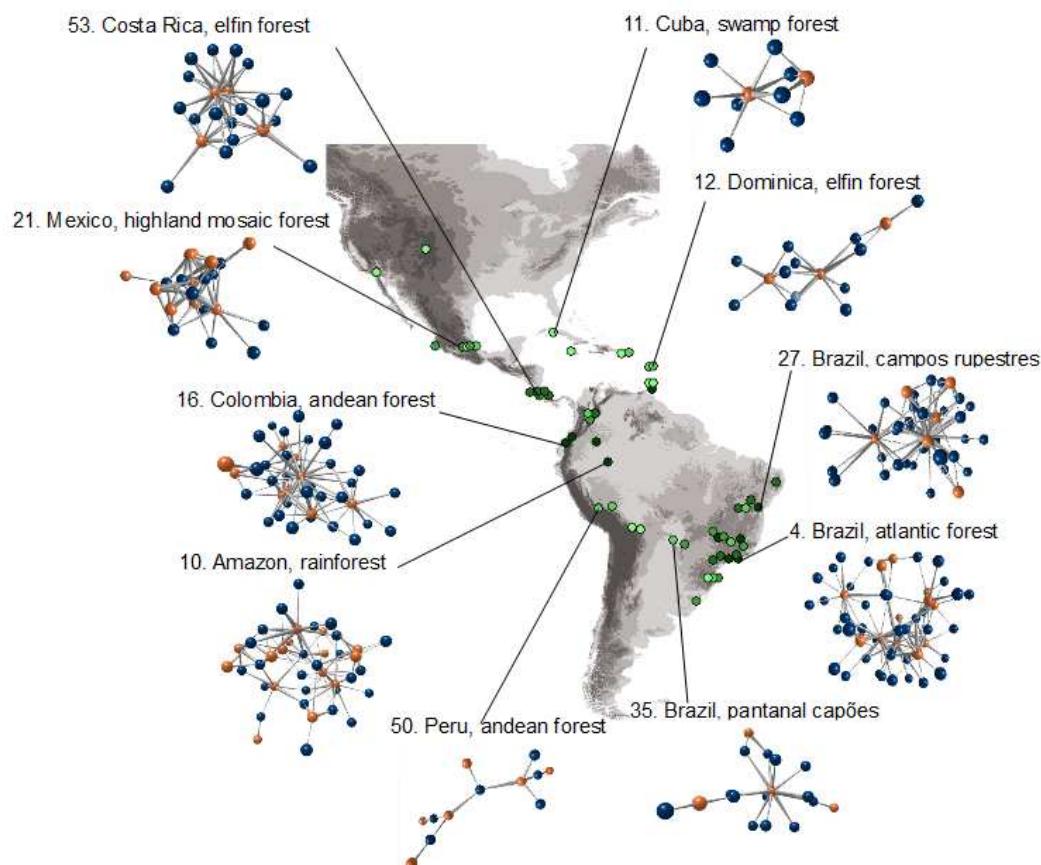
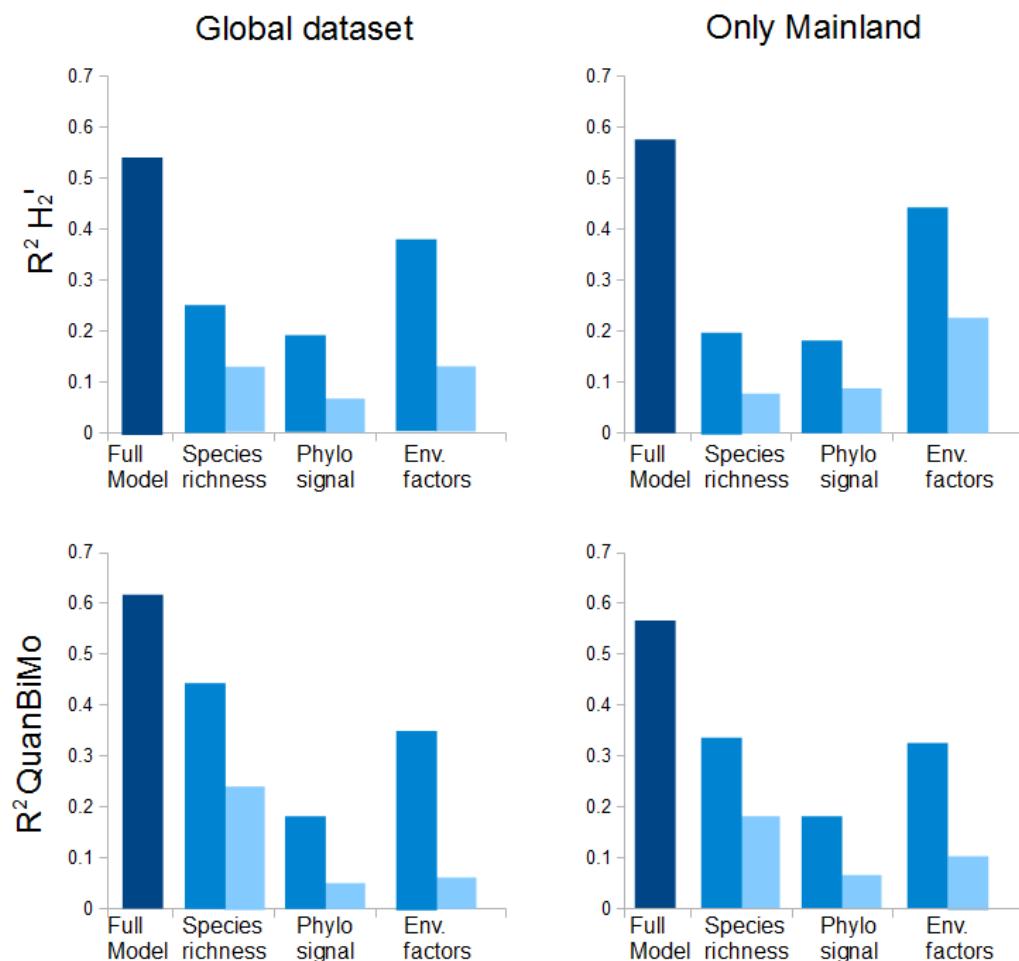


Figure 3.



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**Appendix S1.** List of all plant and hummingbird species present in the studied networks.**Plant species**

Family	Plant Species
Acanthaceae	<i>Anisacanthus brasiliensis</i> Lindau
Acanthaceae	<i>Aphelandra aurantiaca</i> (Scheidw.) Lindl.
Acanthaceae	<i>Aphelandra colorata</i> (Vell.) Wassh.
Acanthaceae	<i>Aphelandra macrostachya</i> Nees
Acanthaceae	<i>Aphelandra</i> sp. 1
Acanthaceae	<i>Aphelandra</i> sp. 2
Acanthaceae	<i>Aphelandra storkii</i> Leonard
Acanthaceae	<i>Dicliptera pohliana</i> Nees
Acanthaceae	<i>Dicliptera squarrosa</i> Nees
Acanthaceae	<i>Geissomeria longiflora</i> Lindl.
Acanthaceae	<i>Geissomeria schottiana</i> Nees
Acanthaceae	<i>Geissomeria</i> sp.
Acanthaceae	<i>Justicia brasiliiana</i> Roth
Acanthaceae	<i>Justicia candicans</i> (Nees) L.D.Benson
Acanthaceae	<i>Justicia carnea</i> Lindl.
Acanthaceae	<i>Justicia cuatrecasasii</i> Wassh.
Acanthaceae	<i>Justicia scheidweileri</i> V.A.W. Graham
Acanthaceae	<i>Justicia sebastianopolitanae</i> Profice
Acanthaceae	<i>Justicia secunda</i> Vahl
Acanthaceae	<i>Justicia</i> sp. 1
Acanthaceae	<i>Justicia</i> sp. 2
Acanthaceae	<i>Justicia tenuistachys</i> (Rusby) Wassh. & J.R.I. Wood
Acanthaceae	<i>Mendoncia</i> sp.
Acanthaceae	<i>Mendoncia coccinea</i> Ruiz & Pav.
Acanthaceae	<i>Odontonema barleroides</i> Kuntze
Acanthaceae	<i>Pachystachys coccinea</i> (Aubl.) Nees
Acanthaceae	<i>Ruellia asperula</i> (Mart. & Nees) Lindau
Acanthaceae	<i>Ruellia brevifolia</i> (Pohl) C.Ezcurra
Acanthaceae	<i>Ruellia inflata</i> Rich.
Acanthaceae	<i>Ruellia</i> sp.
Acanthaceae	<i>Ruellia subsessilis</i> Lindau
Acanthaceae	<i>Sanchezia munita</i> (Nees) Planch.
Acanthaceae	<i>Sanchezia oblonga</i> Ruiz & Pav.
Acanthaceae	<i>Sanchezia putumayensis</i> Leonard
Acanthaceae	<i>Sanchezia scandens</i> (Lindau) Leonard & L.B. Sm.
Acanthaceae	<i>Staurogyne itatiaiae</i> (Wawra) Leonard
Acanthaceae	<i>Suessenguthia multisetosa</i> (Rusby) Wassh. & J.R.I. Wood
Acanthaceae	<i>Sambucus</i> sp.
Alstroemeriaceae	<i>Alstroemeria inodora</i> Herb.
Alstroemeriaceae	<i>Alstroemeria isabellana</i> Herb.
Alstroemeriaceae	<i>Alstroemeria rupestris</i> M.C.Assis
Alstroemeriaceae	<i>Alstroemeria</i> sp.

Family	Plant Species
Alstroemeriaceae	<i>Bomarea cf. carderi</i> Mast.
Alstroemeriaceae	<i>Bomarea costaricensis</i> Kraenzl.
Alstroemeriaceae	<i>Bomarea edulis</i> (Tussac) Herb.
Alstroemeriaceae	<i>Bomarea hirsuta</i> (Kunth) Herb.
Alstroemeriaceae	<i>Bomarea linifolia</i> (Kunth) Baker
Alstroemeriaceae	<i>Bomarea pardina</i> Herb.
Alstroemeriaceae	<i>Bomarea</i> sp. 1
Alstroemeriaceae	<i>Bomarea</i> sp. 2
Amaryllidaceae	<i>Hippeastrum aulicum</i> (Ker Gawl.) Herb.
Amaryllidaceae	<i>Hippeastrum aviflorum</i> (Ravenna) Dutilh
Amaryllidaceae	<i>Hippeastrum puniceum</i> (Lam.) Voss
Amaryllidaceae	<i>Hippeastrum cipoanum</i> (Ravenna) Meerow
Apocynaceae	<i>Allamanda blanchetii</i> A.DC.
Apocynaceae	<i>Blepharodon pictum</i> (Vahl) W.D.Stevens
Apocynaceae	<i>Ditassa retusa</i> Mart.
Apocynaceae	<i>Mandevilla aff. mollissima</i> (Kunth) K.Schum.
Apocynaceae	<i>Mandevilla funiformis</i> (Vell.) K.Schum.
Apocynaceae	<i>Mandevilla hirsuta</i> (Rich.) K.Schum.
Apocynaceae	<i>Mandevilla scabra</i> (Hoffmanns. ex Roem. & Schult.) K.Schum.
Apocynaceae	<i>Nephradenia asparagoides</i> (Decne.) E.Fourn.
Apocynaceae	<i>Pentalinon luteum</i> (L.) B.F.Hansen & Wunderlin
Apocynaceae	<i>Prestonia coalita</i> (Vell.) Woodson
Apocynaceae	<i>Stipecoma peltigera</i> (Stadelm.) Müll.Arg.
Apocynaceae	<i>Tabernaemontana alba</i> Mill.
Apocynaceae	<i>Tabernaemontana cymosa</i> Jacq.
Apocynaceae	<i>Tabernaemontana macrocalyx</i> Müll.Arg.
Araceae	<i>Philodendron fragrantissimum</i> (Hook.) G.Don
Asparagaceae	<i>Furcraea</i> sp.
Asteraceae	<i>Acritopappus longifolius</i> (Gardner) R.M.King & H.Rob.
Asteraceae	<i>Asteraceae</i> sp.
Asteraceae	<i>Baccharis salicina</i> Torr. & A.Gray
Asteraceae	<i>Barnadesia horrida</i> Muschl.
Asteraceae	<i>Barnadesia spinosa</i> L.f.
Asteraceae	<i>Chronopappus bifrons</i> (DC. ex Pers.) DC.
Asteraceae	<i>Cirsium subcoriaceum</i> (Less.) Sch.Bip. ex Sch.Bip.
Asteraceae	<i>Conocliniopsis prasiifolia</i> (DC.) R.M.King & H.Rob.
Asteraceae	<i>Dasyphyllum candolleanum</i> (Gardner) Cabrera
Asteraceae	<i>Diplostethium</i> sp.
Asteraceae	<i>Eremanthus crotonoides</i> (DC.) Sch.Bip.
Asteraceae	<i>Eremanthus erythropappus</i> (DC.) MacLeish
Asteraceae	<i>Espeletia</i> sp.
Asteraceae	<i>Gynoxys longifolia</i> Sch.Bip. ex Wedd.
Asteraceae	<i>Holelepis pedunculata</i> (DC. ex Pers.) DC.
Asteraceae	<i>Lepidaploa cotoneaster</i> (Willd. ex Spreng.) H.Rob.
Asteraceae	<i>Lepidaploa</i> sp.
Asteraceae	<i>Lessingianthus linearis</i> (Spreng.) H.Rob.
Asteraceae	<i>Moquinia kingii</i> (H.Rob.) Gamerro

1	2	3	Family	Plant Species
4	Asteraceae			<i>Mutisia speciosa</i> Aiton ex Hook.
5	Asteraceae			<i>Neomirandea eximia</i> (B.L.Rob.) R.M.King & H.Rob.
6	Asteraceae			<i>Peltophorum dubium</i> (Spreng.) Taub.
7	Asteraceae			<i>Piptocarpha notata</i> (Less.) Baker
8	Asteraceae			<i>Piptolepis leptospermoides</i> (Mart. ex DC.) Sch.Bip.
9	Asteraceae			<i>Pithecoseris pacourinoides</i> Mart. ex DC.
10	Asteraceae			<i>Trixis vauthieri</i> DC.
11	Asteraceae			<i>Vernonanthura phosphorica</i> (Vell.) H.Rob.
12	Balsaminaceae			<i>Impatiens</i> sp.
13	Balsaminaceae			<i>Impatiens walleriana</i> Hook.f.
14	Bignoniaceae			<i>Adenocalymma campicola</i> (Pilg.) L.G.Lohmann
15	Bignoniaceae			<i>Bignonia binata</i> Thunb.
16	Bignoniaceae			<i>Bignoniaceae</i> sp.
17	Bignoniaceae			<i>Campsis cf. Grandiflora</i> (Thunb.) K.Schum.
18	Bignoniaceae			<i>Cuspidaria inaequalis</i> (DC. ex Splitg.) L.G.Lohmann
19	Bignoniaceae			<i>Cuspidaria lateriflora</i> (Mart.) DC.
20	Bignoniaceae			<i>Cuspidaria simplicifolia</i> DC.
21	Bignoniaceae			<i>Dolichandra unguis-cati</i> (L.) L.G.Lohmann
22	Bignoniaceae			<i>Fridericia erubescens</i> (DC.) L.G.Lohmann
23	Bignoniaceae			<i>Fridericia florida</i> (DC.) L.G.Lohmann
24	Bignoniaceae			<i>Fridericia speciosa</i> Mart.
25	Bignoniaceae			<i>Fridericia triplinervia</i> (Mart. ex DC.) L.G.Lohmann
26	Bignoniaceae			<i>Handroanthus chrysotrichus</i> (Mart. ex DC.) Mattos
27	Bignoniaceae			<i>Handroanthus heptaphyllus</i> (Vell.) Mattos
28	Bignoniaceae			<i>Handroanthus impetiginosus</i> (Mart. ex DC.) Mattos
29	Bignoniaceae			<i>Handroanthus selachidentatus</i> (A.H.Gentry) S.O.Grose
30	Bignoniaceae			<i>Jacaranda irwinii</i> A.H.Gentry
31	Bignoniaceae			<i>Jacaranda mimosifolia</i> D.Don
32	Bignoniaceae			<i>Jacaranda puberula</i> Cham.
33	Bignoniaceae			<i>Jacaratia spinosa</i> (Aubl.) A.DC.
34	Bignoniaceae			<i>Lundia cordata</i> (Vell.) DC.
35	Bignoniaceae			<i>Pyrostegia venusta</i> (Ker Gawl.) Miers
36	Bignoniaceae			<i>Tabebuia aurea</i> (Silva Manso) Benth. & Hook.f. ex S.Moore
37	Bignoniaceae			<i>Tabebuia heterophylla</i> (DC.) Britton
38	Bignoniaceae			<i>Tabebuia orinocensis</i> (Sandwith) A.H.Gentry
39	Bignoniaceae			<i>Tabebuia</i> sp.
40	Bignoniaceae			<i>Tabebuia stenocalyx</i> Sprague & Stapf
41	Bignoniaceae			<i>Tecoma stans</i> (L.) Juss. ex Kunth
42	Bignoniaceae			<i>Zeyheria montana</i> Mart.
43	Boraginacea			<i>Cordia bullata</i> var. <i>globosa</i> (Jacq.) Govaerts
44	Boraginacea			<i>Cordia multispicata</i> Cham.
45	Boraginacea			<i>Cordia seleriana</i> Fernald
46	Boraginacea			<i>Cordia superba</i> Cham.
47	Boraginacea			<i>Cordia bicolor</i> A.DC.
48	Boraginacea			<i>Cordia curassavica</i> (Jacq.) Roem. & Schult.
49	Boraginacea			<i>Cordia glabrata</i> (Mart.) A.DC.
50	Boraginacea			<i>Aechmea aquilega</i> (Salisb.) Griseb.
51	Bromeliaceae			
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Family	Plant Species
Bromeliaceae	<i>Aechmea blumenavii</i> Reitz
Bromeliaceae	<i>Aechmea bromeliifolia</i> (Rudge) Baker
Bromeliaceae	<i>Aechmea chantinii</i> (Carrière) Baker
Bromeliaceae	<i>Aechmea contracta</i> (Mart. ex Schult. & Schult.f.) Baker
Bromeliaceae	<i>Aechmea corymbosa</i> (Mart. ex Schult. & Schult.f.) Mez
Bromeliaceae	<i>Aechmea cylindrata</i> Lindm.
Bromeliaceae	<i>Aechmea dichlamydea</i> Baker
Bromeliaceae	<i>Aechmea distichantha</i> Lem.
Bromeliaceae	<i>Aechmea fendleri</i> André ex Mez
Bromeliaceae	<i>Aechmea gamosepala</i> Wittm.
Bromeliaceae	<i>Aechmea longifolia</i> (Rudge) L.B.Sm. & M.A.Spencer
Bromeliaceae	<i>Aechmea mariae-reginae</i> H.Wendl.
Bromeliaceae	<i>Aechmea nudicaulis</i> (L.) Griseb.
Bromeliaceae	<i>Aechmea organensis</i> Wawra
Bromeliaceae	<i>Aechmea pectinata</i> Baker
Bromeliaceae	<i>Aechmea poitaei</i> (Baker) L.B.Sm. & M.A.Spencer
Bromeliaceae	<i>Aechmea recurvata</i> (Klotzsch) L.B.Sm.
Bromeliaceae	<i>Aechmea rubiginosa</i> Mez
Bromeliaceae	<i>Aechmea</i> sp.
Bromeliaceae	<i>Aechmea van-houtteana</i> (Van Houtte) Mez
Bromeliaceae	<i>Aechmea williamsii</i> (L.B.Sm.) L.B.Sm. & M.A.Spencer
Bromeliaceae	<i>Ananas ananassoides</i> (Baker) L.B.Sm.
Bromeliaceae	<i>Ananas comosus</i> (L.) Merr.
Bromeliaceae	<i>Billbergia amoena</i> (Lodd.) Lindl.
Bromeliaceae	<i>Billbergia decora</i> Poepp. & Endl.
Bromeliaceae	<i>Billbergia distachya</i> (Vell.) Mez
Bromeliaceae	<i>Billbergia pyramidalis</i> (Sims) Lindl.
Bromeliaceae	<i>Billbergia vittata</i> Brongn. ex C.Morel
Bromeliaceae	<i>Bromelia antiacantha</i> Bertol.
Bromeliaceae	<i>Bromelia balansae</i> Mez
Bromeliaceae	<i>Bromelia</i> sp.
Bromeliaceae	<i>Bromeliaceae</i> sp. 1
Bromeliaceae	<i>Bromeliaceae</i> sp. 2
Bromeliaceae	<i>Canistrum cyathiforme</i> (Vell.) Mez
Bromeliaceae	<i>Canistrum fragrans</i> (Linden) Mabb.
Bromeliaceae	<i>Canistrum giganteum</i> (Baker) L.B.Sm.
Bromeliaceae	<i>Canistrum perplexum</i> L.B.Sm.
Bromeliaceae	<i>Dyckia leptostachya</i> Baker
Bromeliaceae	<i>Dyckia</i> sp.
Bromeliaceae	<i>Encholirium brachypodium</i> L.B.Sm. & Read
Bromeliaceae	<i>Encholirium spectabile</i> Mart. ex Schult. & Schult.f.
Bromeliaceae	<i>Encholirium subsecundum</i> (Baker) Mez
Bromeliaceae	<i>Guzmania bakeri</i> (Wittm.) Mez
Bromeliaceae	<i>Guzmania berteroniana</i> (Schult. & Schult. f.) Mez
Bromeliaceae	<i>Guzmania candelabrum</i> (André) André ex Mez
Bromeliaceae	<i>Guzmania squarrosa</i> (Mez & Sodiro) L.B.Sm. & Pittendr.
Bromeliaceae	<i>Guzmania danielii</i> L.B.Sm.

1	2	Family	Plant Species
4	5	Bromeliaceae	<i>Guzmania jaramilloi</i> H.E.Luther
6	7	Bromeliaceae	<i>Guzmania monostachia</i> (L.) Rusby ex Mez
8	9	Bromeliaceae	<i>Guzmania nicaraguensis</i> Mez & C.F.Baker
10	11	Bromeliaceae	<i>Guzmania</i> sp. 1
12	13	Bromeliaceae	<i>Guzmania</i> sp. 2
14	15	Bromeliaceae	<i>Guzmania</i> sp. 3
16	17	Bromeliaceae	<i>Guzmania teuscheri</i> L.B.Sm.
18	19	Bromeliaceae	<i>Hohenbergia cf. catingae</i> Ule
20	21	Bromeliaceae	<i>Hohenbergia ridleyi</i> (Baker) Mez
22	23	Bromeliaceae	<i>Neoglaziovia variegata</i> (Arruda) Mez
24	25	Bromeliaceae	<i>Neoregelia bahiana</i> (Ule) L.B.Sm.
26	27	Bromeliaceae	<i>Nidularium rutilans</i> E. Morren
28	29	Bromeliaceae	<i>Nidularium innocentii</i> Lem.
30	31	Bromeliaceae	<i>Nidularium itatiaiae</i> L.B.Sm.
32	33	Bromeliaceae	<i>Nidularium marigoi</i> Leme
34	35	Bromeliaceae	<i>Nidularium procerum</i> Lindm.
36	37	Bromeliaceae	<i>Nidularium seidelii</i> L.B.Sm. & Reitz
38	39	Bromeliaceae	<i>Nidularium</i> sp.
40	41	Bromeliaceae	<i>Orthophytum albopictum</i> Philcox
42	43	Bromeliaceae	<i>Orthophytum lemei</i> E.Pereira & I.A.Penna
44	45	Bromeliaceae	<i>Pitcairnia brittoniana</i> (Mez) Mez
46	47	Bromeliaceae	<i>Pitcairnia flammea</i> Lindl.
48	49	Bromeliaceae	<i>Pitcairnia nigra</i> (Carrière) André
50	51	Bromeliaceae	<i>Pitcairnia</i> sp.
52	53	Bromeliaceae	<i>Pitcairnia sprucei</i> Baker
54	55	Bromeliaceae	<i>Puya clava-herculis</i> Mez & Sodiro
56	57	Bromeliaceae	<i>Puya ferruginea</i> (Ruiz & Pav.) L.B.Sm.
58	59	Bromeliaceae	<i>Quesnelia augusto-coburgii</i> Wawra
60		Bromeliaceae	<i>Quesnelia</i> sp.
		Bromeliaceae	<i>Tillandsia aeranthos</i> (Loisel.) L.B.Sm.
		Bromeliaceae	<i>Tillandsia aff. turneri</i> Baker
		Bromeliaceae	<i>Tillandsia bartramii</i> Elliott
		Bromeliaceae	<i>Tillandsia complanata</i> Benth.
		Bromeliaceae	<i>Tillandsia dasyliriifolia</i> Baker
		Bromeliaceae	<i>Tillandsia fasciculata</i> Sw.
		Bromeliaceae	<i>Tillandsia gardneri</i> Lindl.
		Bromeliaceae	<i>Tillandsia orbicularis</i> L.B.Sm.
		Bromeliaceae	<i>Tillandsia paucifolia</i> Baker
		Bromeliaceae	<i>Tillandsia</i> sp. 1
		Bromeliaceae	<i>Tillandsia</i> sp. 2
		Bromeliaceae	<i>Tillandsia stricta</i> Sol. ex Ker Gawl.
		Bromeliaceae	<i>Tillandsia tetrantha</i> Ruiz & Pav.
		Bromeliaceae	<i>Tillandsia utriculata</i> L.
		Bromeliaceae	<i>Vriesea altodaserrae</i> L.B.Sm.
		Bromeliaceae	<i>Vriesea carinata</i> Wawra
		Bromeliaceae	<i>Vriesea chrysostachys</i> E.Morren
		Bromeliaceae	<i>Vriesea ensiformis</i> (Vell.) Beer

Family	Plant Species
Bromeliaceae	<i>Vriesea erythrodactylon</i> (E.Morren) E.Morren ex Mez
Bromeliaceae	<i>Vriesea incurvata</i> Gaudich.
Bromeliaceae	<i>Vriesea jonghei</i> (K.Koch) E.Morren
Bromeliaceae	<i>Vriesea longicaulis</i> (Baker) Mez
Bromeliaceae	<i>Vriesea medusa</i> Versieux
Bromeliaceae	<i>Vriesea ororiensis</i> (Mez) L.B.Sm. & Pittendr.
Bromeliaceae	<i>Vriesea philippocoburgii</i> Wawra
Bromeliaceae	<i>Vriesea procera</i> (Mart. ex Schult. & Schult.f.) Wittm.
Bromeliaceae	<i>Vriesea procera</i> var. <i>tenuis</i> L.B.Sm.
Bromeliaceae	<i>Vriesea rodigasiana</i> E.Morren
Bromeliaceae	<i>Vriesea sceptrum</i> Mez
Bromeliaceae	<i>Vriesea simplex</i> (Vell.) Beer
Bromeliaceae	<i>Vriesea</i> sp. 1
Bromeliaceae	<i>Vriesea</i> sp. 2
Bromeliaceae	<i>Vriesea</i> sp. 3
Bromeliaceae	<i>Vriesea vagans</i> (L.B.Sm.) L.B.Sm.
Burseraceae	<i>Protium</i> sp.
Cactaceae	<i>Cipocereus minensis</i> (Werderm.) F.Ritter
Cactaceae	<i>Consolea rubescens</i> (Salm-Dyck ex DC.) Lem.
Cactaceae	<i>Melocactus concinnus</i> Buining & Brederoo
Cactaceae	<i>Melocactus ernestii</i> Vaupel
Cactaceae	<i>Melocactus intortus</i> (Mill.) Urb.
Cactaceae	<i>Melocactus zehntneri</i> (Britton & Rose) Luetzelb.
Cactaceae	<i>Micranthocereus flaviflorus</i> Buining & Brederoo
Cactaceae	<i>Micranthocereus purpureus</i> (Gürke) F. Ritter
Cactaceae	<i>Opuntia excelsa</i> Sánchez-Mej.
Cactaceae	<i>Opuntia karwinskiana</i> Salm-Dyck
Cactaceae	<i>Opuntia wernerii</i> Eggli
Cactaceae	<i>Pilosocereus aurisetus</i> (Werderm.) Byles & G.D. Rowley
Cactaceae	<i>Pilosocereus gounellei</i> (F.A.C.Weber ex K.Schum.) Byles & G.D.Rowley
Cactaceae	<i>Pilosocereus pachycladus</i> F. Ritter
Cactaceae	<i>Pilosocereus royenii</i> (L.) Byles & G.D.Rowley
Cactaceae	<i>Pilosocereus</i> sp.
Cactaceae	<i>Tacinga inamoena</i> (K.Schum.) N.P.Taylor & Stuppy
Cactaceae	<i>Tacinga palmadora</i> (Britton & Rose) N.P.Taylor & Stuppy
Campanulaceae	<i>Burmeistera cyclostigmata</i> Donn.Sm.
Campanulaceae	<i>Burmeistera globosa</i> E.Wimm.
Campanulaceae	<i>Burmeistera parviflora</i> E.Wimm. ex Standl.
Campanulaceae	<i>Burmeistera</i> sp.
Campanulaceae	<i>Centropogon cornutus</i> (L.) Druce
Campanulaceae	<i>Centropogon costaricae</i> (Vatke) McVaugh
Campanulaceae	<i>Centropogon granulosus</i> C.Presl
Campanulaceae	<i>Centropogon solanifolius</i> Benth.
Campanulaceae	<i>Centropogon</i> sp. 1
Campanulaceae	<i>Centropogon</i> sp. 2
Campanulaceae	<i>Centropogon talamancensis</i> Wilbur
Campanulaceae	<i>Centropogon valerii</i> Standl.

Family	Plant Species
Campanulaceae	<i>Lobelia cirsifolia</i> Lam.
Campanulaceae	<i>Lobelia fistulosa</i> Vell.
Campanulaceae	<i>Lobelia stricta</i> Sw.
Campanulaceae	<i>Siphocampylus bogotensis</i> E.Wimm.
Campanulaceae	<i>Siphocampylus cf. convolvulaceous</i> (Cham.) G.Don
Campanulaceae	<i>Siphocampylus fimbriatus</i> Regel
Campanulaceae	<i>Siphocampylus giganteus</i> (Cav.) G.Don
Campanulaceae	<i>Siphocampylus imbricatus</i> (Cham.) G.Don
Campanulaceae	<i>Siphocampylus longipedunculatus</i> Pohl
Campanulaceae	<i>Siphocampylus paramicola</i> McVaugh
Campanulaceae	<i>Siphocampylus</i> sp.
Campanulaceae	<i>Siphocampylus sulfureus</i> E.Wimm.
Campanulaceae	<i>Siphocampylus westianus</i> (Thunb.) Pohl
Cannaceae	<i>Canna indica</i> L.
Cannaceae	<i>Canna paniculata</i> Ruiz & Pav.
Cannaceae	<i>Canna</i> sp.
Capparaceae	<i>Crateva tapia</i> L.
Caprifoliaceae	<i>Lonicera japonica</i> Thunb.
Caryocaraceae	<i>Caryocar brasiliense</i> A.St.-Hil.
Chrysobalanaceae	<i>Couepia schottii</i> Fritsch
Clusiaceae	<i>Clusia mangle</i> Rich. ex Planch. & Triana
Clusiaceae	<i>Clusia melchiorii</i> Gleason
Clusiaceae	<i>Clusia</i> sp. 1
Clusiaceae	<i>Clusia</i> sp. 2
Clusiaceae	<i>Symphonia globulifera</i> L.f.
Columelliaceae	<i>Desfontainia spinosa</i> Ruiz & Pav.
Combretaceae	<i>Combretum farinosum</i> Kunth
Combretaceae	<i>Combretum llewelynii</i> J.F.Macbr.
Convolvulaceae	<i>Aniseia martinicensis</i> (Jacq.) Choisy
Convolvulaceae	<i>Ipomoea aquatica</i> Forssk.
Convolvulaceae	<i>Ipomoea bracteata</i> Cav.
Convolvulaceae	<i>Ipomoea quamoclit</i> L.
Convolvulaceae	<i>Ipomoea</i> sp. 1
Convolvulaceae	<i>Ipomoea</i> sp. 2
Convolvulaceae	<i>Ipomoea</i> sp. 3
Convolvulaceae	<i>Ipomoea subincana</i> Meisn.
Convolvulaceae	<i>Jacquemontia</i> sp. 1
Convolvulaceae	<i>Jacquemontia</i> sp. 2
Convolvulaceae	<i>Merremia macrocalyx</i> (Ruiz & Pav.) O'Donell
Costaceae	<i>Costus curvibracteatus</i> Maas
Costaceae	<i>Costus lasius</i> Loes.
Costaceae	<i>Costus pulverulentus</i> C.Presl
Costaceae	<i>Costus scaber</i> Ruiz & Pav.
Costaceae	<i>Costus spiralis</i> (Jacq.) Roscoe
Costaceae	<i>Dimerocostus strobilaceus</i> Kuntze
Crassulaceae	<i>Kalanchoe</i> sp.
Cucurbitaceae	<i>Apodanthera villosa</i> C.Jeffrey

Family	Plant Species
Cucurbitaceae	<i>Gurania acuminata</i> Cogn.
Cucurbitaceae	<i>Gurania coccinea</i> Cogn.
Cucurbitaceae	<i>Gurania lobata</i> (L.) Pruski
Cucurbitaceae	<i>Gurania rhizantha</i> (Poep. & Endl.) C.Jeffrey
Cucurbitaceae	<i>Gurania rufipila</i> Cogn.
Cucurbitaceae	<i>Psiguria ternata</i> (M.Roem.) C.Jeffrey
Ericaceae	<i>Agarista coriifolia</i> (Thunb.) Hook. ex Nied.
Ericaceae	<i>Agarista sp.</i>
Ericaceae	<i>Bejaria aestuans</i> Mutis ex L.
Ericaceae	<i>Cavendishia bracteata</i> (Ruiz & Pav. ex J.St.Hil.) Hoerold
Ericaceae	<i>Cavendishia callista</i> Donn.Sm.
Ericaceae	<i>Cavendishia complectens</i> Hemsl.
Ericaceae	<i>Cavendishia grandifolia</i> Herold
Ericaceae	<i>Cavendishia guatapeensis</i> Mansf.
Ericaceae	<i>Cavendishia pubescens</i> (Kunth) Hemsl.
Ericaceae	<i>Cavendishia quereme</i> (Kunth) Benth. & Hook.f.
Ericaceae	<i>Cavendishia sp.</i>
Ericaceae	<i>Cavendishia tarapotana</i> Benth. & Hook.f.
Ericaceae	<i>Disterigma codonanthum</i> S.F.Blake
Ericaceae	<i>Disterigma humboldtii</i> (Klotzsch) Nied.
Ericaceae	<i>Disterigma sp. 1</i>
Ericaceae	<i>Disterigma sp. 2</i>
Ericaceae	<i>Ericaceae sp. 1</i>
Ericaceae	<i>Ericaceae sp. 2</i>
Ericaceae	<i>Gaultheria erecta</i> Vent.
Ericaceae	<i>Gaultheria gracilis</i> Small
Ericaceae	<i>Gaultheria insipida</i> Benth.
Ericaceae	<i>Gaultheria myrsinoides</i> Kunth
Ericaceae	<i>Gaultheria sp.</i>
Ericaceae	<i>Gaylussacia brasiliensis</i> (Spreng.) Meisn.
Ericaceae	<i>Gaylussacia montana</i> (Pohl) Sleumer
Ericaceae	<i>Gaylussacia oleifolia</i> Dunal
Ericaceae	<i>Gaylussacia virgata</i> Mart. ex Meisn.
Ericaceae	<i>Gonocalyx pterocarpus</i> (Donn.Sm.) Luteyn
Ericaceae	<i>Macleania pentaptera</i> Hoerold
Ericaceae	<i>Macleania recumbens</i> A.C. Sm.
Ericaceae	<i>Macleania rupestris</i> (Kunth) A.C.Sm.
Ericaceae	<i>Psammisia aberrans</i> A.C. Sm.
Ericaceae	<i>Psammisia ecuadorensis</i> Hoerold
Ericaceae	<i>Psammisia falcata</i> Klotzsch
Ericaceae	<i>Psammisia oreogenes</i> Sleumer
Ericaceae	<i>Psammisia pauciflora</i> Griseb. ex A.C. Sm.
Ericaceae	<i>Psammisia penduliflora</i> (Dunal) Klotzsch
Ericaceae	<i>Psammisia ramiflora</i> Klotzsch
Ericaceae	<i>Psammisia sodiroi</i> Hoerold
Ericaceae	<i>Psammisia ulbrichiana</i> Hoerold
Ericaceae	<i>Satyria meiantha</i> Donn.Sm.

1	2	3	Family	Plant Species
4	Ericaceae			<i>Satyria panurensis</i> (Benth. ex Meisn.) Benth. & Hook.f. ex Nied.
5	Ericaceae			<i>Thibaudia costaricensis</i> Hoerold
6	Ericaceae			<i>Thibaudia rigidiflora</i> A.C.Sm.
7	Ericaceae			<i>Vaccinium consanguineum</i> Klotzsch
8	Ericaceae			<i>Vaccinium floribundum</i> Kunth
9	Ericaceae			<i>Vaccinium racemosum</i> (Vahl) Wilbur & Lutelyn
10	Ericaceae			<i>Vaccinium poasanum</i> Donn.Sm.
11	Erythroxylaceae			<i>Erythroxylum vaccinifolium</i> Mart.
12	Euphorbiaceae			<i>Cnidoscolus halteris</i> Fern.Casas
13	Euphorbiaceae			<i>Cnidoscolus pubescens</i> Pohl
14	Euphorbiaceae			<i>Croton blanchetianus</i> Baill.
15	Euphorbiaceae			<i>Jatropha mollissima</i> (Pohl) Baill.
16	Euphorbiaceae			<i>Jatropha mutabilis</i> (Pohl) Baill.
17	Euphorbiaceae			<i>Manihot jacobinensis</i> Müll.Arg.
18	Euphorbiaceae			<i>Manihot reniformis</i> Pohl
19	Fabaceae			<i>Abarema brachystachya</i> (DC.) Barneby & J.W.Grimes
20	Fabaceae			<i>Albizia saman</i> (Jacq.) Merr.
21	Fabaceae			<i>Anadenanthera colubrina</i> (Vell.) Brenan
22	Fabaceae			<i>Bauhinia brevipes</i> Vogel
23	Fabaceae			<i>Bauhinia cheilantha</i> (Bong.) Steud.
24	Fabaceae			<i>Bauhinia forficata</i> Link
25	Fabaceae			<i>Bauhinia longifolia</i> (Bong.) Steud.
26	Fabaceae			<i>Bauhinia rufa</i> (Bong.) Steud.
27	Fabaceae			<i>Bauhinia ungulata</i> L.
28	Fabaceae			<i>Bauhinia variegata</i> L.
29	Fabaceae			<i>Bowdichia virgilioides</i> Kunth
30	Fabaceae			<i>Brownea coccinea</i> subsp. <i>capitella</i> (Jacq.) D.Velasquez & Agostini
31	Fabaceae			<i>Caesalpinia pyramidalis</i> Tul.
32	Fabaceae			<i>Calliandra brevipes</i> Benth.
33	Fabaceae			<i>Calliandra carbonaria</i> Benth.
34	Fabaceae			<i>Calliandra guildingii</i> Benth.
35	Fabaceae			<i>Calliandra hygrophila</i> Mackinder & G.P.Lewis
36	Fabaceae			<i>Calliandra mucugeana</i> Renvoize
37	Fabaceae			<i>Calliandra purdiaei</i> Benth.
38	Fabaceae			<i>Calliandra tweedii</i> Benth.
39	Fabaceae			<i>Calliandra viscidula</i> Benth.
40	Fabaceae			<i>Camptosema coriaceum</i> Benth.
41	Fabaceae			<i>Camptosema ellipticum</i> (Desv.) Burkart
42	Fabaceae			<i>Camptosema scarlatinum</i> (Benth.) Burkart
43	Fabaceae			<i>Centrosema brasiliianum</i> (L.) Benth.
44	Fabaceae			<i>Centrosema coriaceum</i> Benth.
45	Fabaceae			<i>Chaetocalyx subulatus</i> Mackinder
46	Fabaceae			<i>Clathrotropis brachypetala</i> (Tul.) Kleinhoonte
47	Fabaceae			<i>Collaea speciosa</i> (Loisel.) DC.
48	Fabaceae			<i>Dahlstedtia pinnata</i> (Benth.) Malme
49	Fabaceae			<i>Dioclea grandiflora</i> Benth.
50	Fabaceae			<i>Dioclea sp.</i>
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Family	Plant Species
Fabaceae	<i>Erythrina corallodendron</i> L.
Fabaceae	<i>Erythrina crista-galli</i> L.
Fabaceae	<i>Erythrina fusca</i> Lour.
Fabaceae	<i>Erythrina lanata</i> Rose
Fabaceae	<i>Erythrina poeppigiana</i> (Walp.) O.F.Cook
Fabaceae	<i>Erythrina</i> sp.
Fabaceae	<i>Erythrina speciosa</i> Andrews
Fabaceae	<i>Fabaceae</i> sp.
Fabaceae	<i>Inga edulis</i> Mart.
Fabaceae	<i>Inga ingoides</i> (Rich.) Willd.
Fabaceae	<i>Inga leiocalycina</i> Benth.
Fabaceae	<i>Inga luschnathiana</i> Benth.
Fabaceae	<i>Inga sessilis</i> (Vell.) Mart.
Fabaceae	<i>Inga</i> sp. 1
Fabaceae	<i>Inga</i> sp. 2
Fabaceae	<i>Inga</i> sp. 3
Fabaceae	<i>Inga venosa</i> Griseb.
Fabaceae	<i>Inga vera</i> Willd.
Fabaceae	<i>Lonchocarpus benthamianus</i> Pittier
Fabaceae	<i>Lysiloma latisiliquum</i> (L.) Benth.
Fabaceae	<i>Mucuna rostrata</i> Benth.
Fabaceae	<i>Mysanthus uleanus</i> (Harms) G.P.Lewis & A.Delgado
Fabaceae	<i>Neorudolphia volubilis</i> (Willd.) Britton
Fabaceae	<i>Periandra coccinea</i> (Schrad.) Benth.
Fabaceae	<i>Periandra mediterranea</i> (Vell.) Taub.
Fabaceae	<i>Phaseolus coccineus</i> L.
Fabaceae	<i>Abarema jupunba</i> (Willd.) Britton & Killip
Fabaceae	<i>Senna reticulata</i> (Willd.) H.S.Irwin & Barneby
Fabaceae	<i>Tachigali paniculata</i> Aubl.
Fabaceae	<i>Tephrosia noctiflora</i> Baker
Fabaceae	<i>Tipuana tipu</i> (Benth.) Kuntze
Fabaceae	<i>Viguieranthus alternans</i> (Benth.) Villiers
Gentianaceae	<i>Zygia lathetica</i> Barneby & J.W. Grimes
Gentianaceae	<i>Chelonanthus alatus</i> (Aubl.) Pulle
Gentianaceae	<i>Calolisianthus pendulus</i> (Mart.) Gilg
Gentianaceae	<i>Lisianthus roseus</i> Sessé & Moc.
Gentianaceae	<i>Macrocarpaea cf. rubra</i> Malme
Gentianaceae	<i>Macrocarpaea</i> sp.
Gentianaceae	<i>Macrocarpaea valerioi</i> Standl.
Gentianaceae	<i>Symbolanthus tricolor</i> Gilg
Gentianaceae	<i>Tachia occidentalis</i> Maguire & Weaver
Gesneriaceae	<i>Alloplectus</i> sp. 1
Gesneriaceae	<i>Alloplectus</i> sp. 2
Gesneriaceae	<i>Besleria cf. solanoides</i> Kunth
Gesneriaceae	<i>Besleria columneoides</i> Hanst.
Gesneriaceae	<i>Besleria longimucronata</i> Hoehne
Gesneriaceae	<i>Besleria notabilis</i> C.V.Morton

1	2	3	Family	Plant Species
4	Gesneriaceae			<i>Besleria rotundifolia</i> Rusby
5	Gesneriaceae			<i>Besleria solanoides</i> Kunth
6	Gesneriaceae			<i>Besleria petiolaris</i> Griseb.
7	Gesneriaceae			<i>Columnea ciliata</i> (Wiehler) L.P.Kvist & L.E.Skog
8	Gesneriaceae			<i>Columnea ericae</i> Mansf.
9	Gesneriaceae			<i>Columnea magnifica</i> Klotzsch ex Oerst.
10	Gesneriaceae			<i>Columnea medicinalis</i> (Wiehler) L.E.Skog & L.P.Kvist
11	Gesneriaceae			<i>Columnea microcalyx</i> Hanst.
12	Gesneriaceae			<i>Columnea purpurata</i> Hanst.
13	Gesneriaceae			<i>Columnea querceti</i> Oerst.
14	Gesneriaceae			<i>Columnea sp.</i>
15	Gesneriaceae			<i>Columnea strigosa</i> Benth.
16	Gesneriaceae			<i>Drymonia coccinea</i> (Aubl.) Wiehler
17	Gesneriaceae			<i>Drymonia conchocalyx</i> Hanst.
18	Gesneriaceae			<i>Drymonia semicordata</i> (Poepp.) Wiehler
19	Gesneriaceae			<i>Gasteranthus sp.</i>
20	Gesneriaceae			<i>Gesneriaceae sp. 1</i>
21	Gesneriaceae			<i>Gesneriaceae sp. 2</i>
22	Gesneriaceae			<i>Gesneriaceae sp. 3</i>
23	Gesneriaceae			<i>Gesneriaceae sp. 4</i>
24	Gesneriaceae			<i>Glossoloma bolivianum</i> (Britton ex Rusby) J.L.Clark
25	Gesneriaceae			<i>Hypocyrta fritschii</i> (Hoehne) Handro
26	Gesneriaceae			<i>Kohleria spicata</i> (Kunth) Oerst.
27	Gesneriaceae			<i>Kohleria tigridia</i> (Ohlend.) Roalson & Boggan
28	Gesneriaceae			<i>Nematanthus aff. fritschii</i> Hoehne
29	Gesneriaceae			<i>Nematanthus australis</i> Chautems
30	Gesneriaceae			<i>Nematanthus crassifolius</i> (Schott) Wiehler
31	Gesneriaceae			<i>Nematanthus fluminensis</i> (Vell.) Fritsch
32	Gesneriaceae			<i>Nematanthus formix</i> (Vell.) Chautems
33	Gesneriaceae			<i>Nematanthus gregarius</i> D.L. Denham
34	Gesneriaceae			<i>Nematanthus lanceolatus</i> (Poir.) Chautems
35	Gesneriaceae			<i>Nematanthus savannarum</i> (C.V.Morton) J.L.Clark
36	Gesneriaceae			<i>Nematanthus strigillosus</i> (Mart.) H.E. Moore
37	Gesneriaceae			<i>Paliavana sericiflora</i> Benth.
38	Gesneriaceae			<i>Sinningia cooperi</i> (J. Paxton) Wiehler
39	Gesneriaceae			<i>Sinningia douglasii</i> (Lindl.) Chautems
40	Gesneriaceae			<i>Sinningia elatior</i> (Kunth) Chautems
41	Gesneriaceae			<i>Sinningia gigantifolia</i> Chautems
42	Gesneriaceae			<i>Sinningia glazioviana</i> (Fritsch) Chautems
43	Gesneriaceae			<i>Sinningia rupicola</i> (Mart.) Wiehler
44	Heliconiaceae			<i>Heliconia acuminata</i> A.Rich.
45	Heliconiaceae			<i>Heliconia angusta</i> Vell.
46	Heliconiaceae			<i>Heliconia atropurpurea</i> G.S.Daniels & F.G.Stiles
47	Heliconiaceae			<i>Heliconia bihai</i> (L.) L.
48	Heliconiaceae			<i>Heliconia burleana</i> Abalo & G.Morales
49	Heliconiaceae			<i>Heliconia subulata</i> Ruiz & Pav. <i>subsp. subulata</i>
50	Heliconiaceae			<i>Heliconia episcopal</i> Vell.
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Family	Plant Species
Heliconiaceae	<i>Heliconia farinosa</i> Raddi
Heliconiaceae	<i>Heliconia hirsuta</i> L.f.
Heliconiaceae	<i>Heliconia imbricata</i> (Kuntze) Baker
Heliconiaceae	<i>Heliconia juruana</i> Loes.
Heliconiaceae	<i>Heliconia lankesteri</i> Standl.
Heliconiaceae	<i>Heliconia latispatha</i> Benth.
Heliconiaceae	<i>Heliconia marginata</i> (Griggs) Pittier
Heliconiaceae	<i>Heliconia mariae</i> Hook.f.
Heliconiaceae	<i>Heliconia mathiasiae</i> G.S.Daniels & F.G.Stiles
Heliconiaceae	<i>Heliconia metallica</i> Planch. & Linden ex Hook.
Heliconiaceae	<i>Heliconia pogonantha</i> Cufod.
Heliconiaceae	<i>Heliconia psittacorum</i> L.f.
Heliconiaceae	<i>Heliconia rostrata</i> Ruiz & Pav.
Heliconiaceae	<i>Heliconia schumanniana</i> Loes.
Heliconiaceae	<i>Heliconia</i> sp. 1
Heliconiaceae	<i>Heliconia</i> sp. 2
Heliconiaceae	<i>Heliconia</i> sp. 3
Heliconiaceae	<i>Heliconia</i> sp. 4
Heliconiaceae	<i>Heliconia spathocircinada</i> Aristeg.
Heliconiaceae	<i>Heliconia stricta</i> Huber
Heliconiaceae	<i>Heliconia subulata</i> Ruiz & Pav.
Heliconiaceae	<i>Heliconia vaginalis</i> Benth.
Heliconiaceae	<i>Heliconia wagneriana</i> Petersen
Humiriaceae	<i>Humiria balsamifera</i> Aubl.
Iridaceae	<i>Iridaceae</i> sp.
Lamiaceae	<i>Aegiphila obducta</i> Vell.
Lamiaceae	<i>Aegiphila perplexa</i> Moldenke
Lamiaceae	<i>Amazonia campestris</i> (Aubl.) Moldenke
Lamiaceae	<i>Volkameria aculeata</i> L.
Lamiaceae	<i>Eriope latifolia</i> (Mart. ex Benth.) Harley
Lamiaceae	<i>Hyptidendron aspernum</i> (Spreng.) Harley
Lamiaceae	<i>Hyptis atrorubens</i> Poit.
Lamiaceae	<i>Hyptis cuniloides</i> Epling
Lamiaceae	<i>Hyptis leptostachys</i> Epling
Lamiaceae	<i>Hyptis pectinata</i> (L.) Poit.
Lamiaceae	<i>Hyptis</i> sp.
Lamiaceae	<i>Lamiaceae</i> sp.
Lamiaceae	<i>Leonotis nepetifolia</i> (L.) R.Br.
Lamiaceae	<i>Prunella vulgaris</i> L.
Lamiaceae	<i>Salvia arenaria</i> A.St.-Hil. ex Benth.
Lamiaceae	<i>Salvia articulata</i> Epling
Lamiaceae	<i>Salvia carnea</i> Kunth
Lamiaceae	<i>Salvia elegans</i> Vahl
Lamiaceae	<i>Salvia mocinoi</i> Benth.
Lamiaceae	<i>Salvia sellowiana</i> Benth.
Lamiaceae	<i>Salvia</i> sp.
Lamiaceae	<i>Vitex cymosa</i> Bertero ex Spreng.

1	2	3	Family	Plant Species
4	Lamiaceae			<i>Vitex divaricata</i> Sw.
5	Lecythidaceae			<i>Lecythidoideae</i> sp.
6	Loganiaceae			<i>Spigelia pulchella</i> Mart.
7	Loranthaceae			<i>Gaiadendron punctatum</i> (Ruiz & Pav.) G.Don
8	Loranthaceae			<i>Ligaria teretiflora</i> (Rizzini) Kuijt
9	Loranthaceae			<i>Loranthaceae</i> sp.
10	Loranthaceae			<i>Loranthus bicalyculatus</i> Mart. ex Schult. & Schult.f.
11	Loranthaceae			<i>Psittacanthus cordatus</i> (Hoffmanns. ex Schult. f.) Blume
12	Loranthaceae			<i>Psittacanthus cucullaris</i> (Lam.) G.Don
13	Loranthaceae			<i>Psittacanthus dichrous</i> Mart.
14	Loranthaceae			<i>Psittacanthus flavo-viridis</i> Eichler
15	Loranthaceae			<i>Loranthus furcatus</i> Mart. ex Schult. & Schult.f.
16	Loranthaceae			<i>Psittacanthus lasianthus</i> Sandwith
17	Loranthaceae			<i>Tripodanthus acutifolius</i> (Ruiz & Pav.) Tiegh.
18	Loranthaceae			<i>Tristerix longebracteatus</i> (Desr.) Barlow & Wiens
19	Lythraceae			<i>Cuphea dipetala</i> (L.f.) Koehne
20	Lythraceae			<i>Cuphea ericoides</i> Cham. & Schldl.
21	Lythraceae			<i>Cuphea melvilla</i> Lindl.
22	Lythraceae			<i>Lafoensia glyptocarpa</i> Koehne
23	Lythraceae			<i>Lafoensia</i> sp.
24	Malvaceae			<i>Abutilon aff. regnellii</i> Miq.
25	Malvaceae			<i>Abutilon bedfordianum</i> (Hook.) A.St.-Hil.
26	Malvaceae			<i>Abutilon darwinii</i> Hook.f.
27	Malvaceae			<i>Abutilon</i> sp.
28	Malvaceae			<i>Ceiba aesculifolia</i> (Kunth) Britten & Baker f.
29	Malvaceae			<i>Ceiba glaziovii</i> (Kuntze) K.Schum.
30	Malvaceae			<i>Ceiba pentandra</i> (L.) Gaertn.
31	Malvaceae			<i>Ceiba speciosa</i> (A.St.-Hil.) Ravenna
32	Malvaceae			<i>Dombeya wallichii</i> (Lindl.) K.Schum
33	Malvaceae			<i>Eriotheca gracilipes</i> (K.Schum.) A.Robyns
34	Malvaceae			<i>Eriotheca pentaphylla</i> (Vell.) A.Robyns
35	Malvaceae			<i>Guazuma ulmifolia</i> Lam.
36	Malvaceae			<i>Helicteres brevispira</i> A.Juss.
37	Malvaceae			<i>Helicteres eichleri</i> K. Schum.
38	Malvaceae			<i>Helicteres guazumifolia</i> Kunth
39	Malvaceae			<i>Helicteres sacarolha</i> A.Juss.
40	Malvaceae			<i>Helicteres velutina</i> K.Schum.
41	Malvaceae			<i>Hibiscus rosa-sinensis</i> L.
42	Malvaceae			<i>Luehea divaricata</i> Mart.
43	Malvaceae			<i>Malvaceae</i> sp.
44	Malvaceae			<i>Malvaviscus arboreus</i> Cav.
45	Malvaceae			<i>Malvaviscus palmanus</i> Pittier & Donn.Sm.
46	Malvaceae			<i>Melochia tomentosa</i> L.
47	Malvaceae			<i>Pavonia luetzelburgii</i> Ulbr.
48	Malvaceae			<i>Pavonia</i> sp.
49	Malvaceae			<i>Pavonia viscosa</i> A. St.-Hil.
50	Malvaceae			<i>Quararibea lasiocalyx</i> (K.Schum.) Vischer

Family	Plant Species
Malvaceae	<i>Spirotheca rivieri</i> (Decne.) Ulbr.
Malvaceae	<i>Urena lobata</i> L.
Malvaceae	<i>Waltheria cinerescens</i> A. St.-Hil.
Malvaceae	<i>Waltheria rotundifolia</i> Schrank
Marantaceae	<i>Calathea capitata</i> (Ruiz & Pav.) Lindl.
Marantaceae	<i>Calathea gymnocarpa</i> H.A.Kenn.
Marantaceae	<i>Calathea inocephala</i> (Kuntze) T.Durand & B.D.Jacks.
Marantaceae	<i>Calathea lasiostachya</i> Donn.Sm.
Marantaceae	<i>Calathea lutea</i> (Aubl.) E.Mey. ex Schult.
Marantaceae	<i>Calathea zingiberina</i> Körn.
Marantaceae	<i>Ischnosiphon aromatica</i> (Aubl.) Körn.
Marantaceae	<i>Ischnosiphon hirsutus</i> Petersen
Marantaceae	<i>Ischnosiphon lasiocoleus</i> K.Schum. ex Loes.
Marantaceae	<i>Maranta furcata</i> Nees & Mart.
Marantaceae	<i>Monotagma secundum</i> (Petersen) K.Schum.
Marantaceae	<i>Stromanthe thalia</i> (Vell.) J.M.A.Braga
Marcgraviaceae	<i>Marcgravia myriostigma</i> Triana & Planch.
Marcgraviaceae	<i>Marcgravia polyantha</i> Delpino
Marcgraviaceae	<i>Norantea guianensis</i> Aubl.
Marcgraviaceae	<i>Sarcopera</i> sp.
Marcgraviaceae	<i>Schwartzia brasiliensis</i> (Choisy) Bedell ex Gir.-Cañas
Melastomataceae	<i>Acinodendron sintenisii</i> (Cogn.) Kuntze
Melastomataceae	<i>Brachyotum ledifolium</i> (Desr.) Triana
Melastomataceae	<i>Brachyotum lindenii</i> Cogn.
Melastomataceae	<i>Brachyotum nutans</i> Gleason
Melastomataceae	<i>Brachyotum</i> sp.
Melastomataceae	<i>Charianthus grenadensis</i> Penneys & Judd
Melastomataceae	<i>Huilaea minor</i> (L. Uribe) Lozano & N. Ruiz
Melastomataceae	<i>Melastomataceae</i> sp.
Melastomataceae	<i>Miconia purpurea</i> (D. Don) Judd & Skean
Melastomataceae	<i>Miconia</i> sp. 1
Melastomataceae	<i>Miconia</i> sp. 2
Musaceae	<i>Musa × paradisiaca</i> L.
Musaceae	<i>Musa</i> sp. 1
Musaceae	<i>Musa</i> sp. 2
Myrtaceae	<i>Callistemon speciosus</i> (Sims) Sweet
Myrtaceae	<i>Eucalyptus</i> sp.
Myrtaceae	<i>Melaleuca leucadendra</i> (L.) L.
Myrtaceae	<i>Myrcia lasiantha</i> DC.
Myrtaceae	<i>Myrcianthes oreophila</i> (Diels) McVaugh
Myrtaceae	<i>Psidium</i> sp.
Myrtaceae	<i>Syzygium jambos</i> (L.) Alston
Myrtaceae	<i>Syzygium malaccense</i> (L.) Merr. & L.M.Perry
Nyctaginaceae	<i>Bougainvillea</i> sp.
Onagraceae	<i>Fuchsia corollata</i> Benth.
Onagraceae	<i>Fuchsia macrostigma</i> Benth.
Onagraceae	<i>Fuchsia microphylla</i> Kunth

1	2	3	Family	Plant Species
4	Onagraceae			<i>Fuchsia regia</i> (Vand. ex Vell.) Munz
5	Onagraceae			<i>Fuchsia splendens</i> Zucc.
6	Orchidaceae			<i>Cattleya elongata</i> Barb.Rodr.
7	Orchidaceae			<i>Corymborkis flava</i> (Sw.) Kuntze
8	Orchidaceae			<i>Elleanthus aurantiacus</i> (Lindl.) Rchb.f.
9	Orchidaceae			<i>Elleanthus smithii</i> Schltr.
10	Orchidaceae			<i>Elleanthus sp.</i>
11	Orchidaceae			<i>Orchidaceae sp.</i>
12	Orobanchaceae			<i>Agalinis angustifolia</i> (Mart.) D'Arcy
13	Orobanchaceae			<i>Castilleja angustifolia</i> (Nutt.) G.Don
14	Orobanchaceae			<i>Castilleja fissifolia</i> L.f.
15	Orobanchaceae			<i>Castilleja irasuensis</i> Oerst.
16	Orobanchaceae			<i>Castilleja miniata</i> Douglas ex Hook.
17	Orobanchaceae			<i>Castilleja scorzonerifolia</i> Kunth
18	Orobanchaceae			<i>Castilleja tenuiflora</i> Benth.
19	Orobanchaceae			<i>Esterhazya splendida</i> J.C.Mikan
20	Orobanchaceae			<i>Vellosiella dracocephaloidea</i> (Vell.) Baill.
21	Oxalidaceae			<i>Oxalis psoraleoides</i> Kunth
22	Passifloraceae			<i>Passiflora cf. skiantha</i> Huber
23	Passifloraceae			<i>Passiflora coccinea</i> Aubl.
24	Passifloraceae			<i>Passiflora edmundoi</i> Sacco
25	Passifloraceae			<i>Passiflora kermesina</i> Link & Otto
26	Passifloraceae			<i>Passiflora luetzelburgii</i> Harms
27	Passifloraceae			<i>Passiflora mixta</i> L.f.
28	Passifloraceae			<i>Passiflora quadrangularis</i> Rodschied
29	Passifloraceae			<i>Passiflora sp.</i>
30	Passifloraceae			<i>Passiflora speciosa</i> Gardner
31	Passifloraceae			<i>Passiflora spinosa</i> (Poep. & Endl.) Mast.
32	Passifloraceae			<i>Passiflora vitifolia</i> Kunth
33	Passifloraceae			<i>Passifloraceae sp.</i>
34	Passifloraceae			<i>Turnera ulmifolia</i> L.
35	Phytolaccaceae			<i>Phytolacca rivinoides</i> Kunth & C.D.Bouché
36	Plantaginaceae			<i>Penstemon gentianoides</i> (Kunth) Poir.
37	Plantaginaceae			<i>Penstemon labrosus</i> (A. Gray) Mast. ex Hook. f.
38	Plantaginaceae			<i>Penstemon roseus</i> (Cerv. ex Sweet) G. Don
39	Plumbaginaceae			<i>Plumbago zeylanica</i> L.
40	Polygalaceae			<i>Monnieria</i> sp.
41	Polygonaceae			<i>Antigonon leptopus</i> Hook. & Arn.
42	Proteaceae			<i>Grevillea banksii</i> R.Br.
43	Ranunculaceae			<i>Aquilegia formosa</i> Fisch. ex DC.
44	Ranunculaceae			<i>Delphinium barbeyi</i> (Huth) Huth
45	Rosaceae			<i>Rubus rosifolius</i> Sm.
46	Rosaceae			<i>Rubus</i> sp. 1
47	Rosaceae			<i>Rubus</i> sp. 2
48	Rubiaceae			<i>Augusta longifolia</i> (Spreng.) Rehder
49	Rubiaceae			<i>Bouvardia ternifolia</i> (Cav.) Schldl.
50	Rubiaceae			<i>Cinchona cf. pitayensis</i> (Wedd.) Wedd.
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Family	Plant Species
Rubiaceae	<i>Coussarea hydrangeifolia</i> (Benth.) Benth. & Hook.f. ex Müll.Arg.
Rubiaceae	<i>Erythalis fruticosa</i> L.
Rubiaceae	<i>Faramea eurycarpa</i> Donn.Sm.
Rubiaceae	<i>Ferdinandusa sprucei</i> K.Schum.
Rubiaceae	<i>Genipa americana</i> L.
Rubiaceae	<i>Gonzalagunia hirsuta</i> (Jacq.) K.Schum.
Rubiaceae	<i>Hamelia patens</i> Jacq.
Rubiaceae	<i>Hamelia rostrata</i> Bartl. ex DC.
Rubiaceae	<i>Hillia triflora</i> (Oerst.) C.M.Taylor
Rubiaceae	<i>Hoffmannia arborescens</i> Donn.Sm.
Rubiaceae	<i>Isertia hypoleuca</i> Benth.
Rubiaceae	<i>Isertia parviflora</i> Vahl
Rubiaceae	<i>Isertia rosea</i> Spruce ex K.Schum.
Rubiaceae	<i>Manettia aff. sabiceoides</i> Wernham
Rubiaceae	<i>Manettia coccocypseloides</i> Wernham
Rubiaceae	<i>Manettia cordifolia</i> Mart.
Rubiaceae	<i>Manettia dominicensis</i> Wernham
Rubiaceae	<i>Manettia mitis</i> (Vell.) K.Schum.
Rubiaceae	<i>Manettia pubescens</i> Cham. & Schltdl.
Rubiaceae	<i>Morinda citrifolia</i> L.
Rubiaceae	<i>Palicourea anderssoniana</i> C.M.Taylor
Rubiaceae	<i>Palicourea angustifolia</i> Kunth
Rubiaceae	<i>Palicourea aschersonianoides</i> (Wernham) Steyerm.
Rubiaceae	<i>Palicourea cf. anacardiifolia</i> (Humb. & Bonpl. ex Schult.) Standl.
Rubiaceae	<i>Palicourea cf. nigricans</i> K.Krause
Rubiaceae	<i>Palicourea cf. vagans</i> Wernham
Rubiaceae	<i>Palicourea coriacea</i> (Cham.) K.Schum.
Rubiaceae	<i>Palicourea crocea</i> (Sw.) Schult.
Rubiaceae	<i>Palicourea demissa</i> Standl.
Rubiaceae	<i>Palicourea fastigiata</i> Kunth
Rubiaceae	<i>Palicourea gomezii</i> C.M.Taylor
Rubiaceae	<i>Palicourea guianensis</i> Aubl.
Rubiaceae	<i>Palicourea lassorrhachis</i> Oerst.
Rubiaceae	<i>Palicourea marcgravii</i> A.St.-Hil.
Rubiaceae	<i>Palicourea quadrifolia</i> (Rudge) DC.
Rubiaceae	<i>Palicourea rigida</i> Kunth
Rubiaceae	<i>Palicourea rudgeoides</i> (Müll.Arg.) Standl.
Rubiaceae	<i>Palicourea sodiroi</i> Standl.
Rubiaceae	<i>Palicourea</i> sp. 1
Rubiaceae	<i>Palicourea</i> sp. 2
Rubiaceae	<i>Palicourea subspicata</i> Huber
Rubiaceae	<i>Pentagonia monocaulis</i> C.M.Taylor
Rubiaceae	<i>Posoqueria</i> sp.
Rubiaceae	<i>Psychotria cf. nemorosa</i> Gardner
Rubiaceae	<i>Psychotria bahiensis</i> DC.
Rubiaceae	<i>Psychotria berteroana</i> DC.
Rubiaceae	<i>Psychotria capitata</i> Ruiz & Pav.

Family	Plant Species
Rubiaceae	<i>Psychotria elata</i> (Sw.) Hammel
Rubiaceae	<i>Psychotria mapourioides</i> DC.
Rubiaceae	<i>Psychotria muscosa</i> (Jacq.) Steyerm.
Rubiaceae	<i>Psychotria nuda</i> (Cham. & Schltdl.) Wawra
Rubiaceae	<i>Psychotria platypoda</i> DC.
Rubiaceae	<i>Psychotria poeppigiana</i> Müll.Arg.
Rubiaceae	<i>Psychotria suterella</i> Müll.Arg.
Rubiaceae	<i>Psychotria urbaniana</i> Steyerm.
Rubiaceae	<i>Psychotria vellosiana</i> Benth.
Rubiaceae	<i>Retiniphyllum speciosum</i> (Spruce ex Benth.) Müll.Arg.
Rubiaceae	<i>Rubiaceae</i> sp.
Rubiaceae	<i>Rudgea loretensis</i> Standl.
Rubiaceae	<i>Sabicea aspera</i> Aubl.
Rubiaceae	<i>Sabicea villosa</i> Willd. ex Schult.
Rubiaceae	<i>Schradera exotica</i> (J.F.Gmel.) Standl.
Rubiaceae	<i>Staelia galiooides</i> DC.
Rubiaceae	<i>Warszewiczia coccinea</i> (Vahl) Klotzsch
Rutaceae	<i>Decagonocarpus cornutus</i> R.S.Cowan
Rutaceae	<i>Erythrociton fallax</i> Kallunki
Rutaceae	<i>Rutaceae</i> sp.
Salicaceae	<i>Ryania pyrifera</i> (Rich.) Uittien & Sleumer
Sapindaceae	<i>Serjania aff. caracasana</i> (Jacq.) Willd.
Sapindaceae	<i>Serjania coradinii</i> Ferrucci & Somner
Sapindaceae	<i>Serjania glutinosa</i> Radlk.
Schlegeliaceae	<i>Schlegelia brachyantha</i> Griseb.
Schlegeliaceae	<i>Schlegelia fastigiata</i> Schery
Scrophulariaceae	<i>Buddleja cf. brasiliensis</i> J.Jacq.
Solanaceae	<i>Brugmansia arborea</i> (L.) Steud.
Solanaceae	<i>Cestrum corymbosum</i> Schltdl.
Solanaceae	<i>Cestrum macrophyllum</i> Vent.
Solanaceae	<i>Cestrum schlechtendahlii</i> G.Don
Solanaceae	<i>Cestrum</i> sp. 1
Solanaceae	<i>Cestrum</i> sp. 2
Solanaceae	<i>Markea coccinea</i> Rich.
Styracaceae	<i>Styrax ferrugineus</i> Nees & Mart.
Theaceae	<i>Gordonia fruticosa</i> (Schrad.) H.Keng
Tropaeolaceae	<i>Tropaeolum moritzianum</i> Klotzsch
Tropaeolaceae	<i>Tropaeolum pentaphyllum</i> Lam.
Velloziaceae	<i>Barbacenia blanchetii</i> Goethart & Henrard
Velloziaceae	<i>Barbacenia flava</i> Mart. ex Schult. & Schult.f.
Velloziaceae	<i>Barbacenia gentianoides</i> Goethart & Henrard
Velloziaceae	<i>Barbacenia luzulifolia</i> Mart. ex Schult. & Schult.f.
Velloziaceae	<i>Barbacenia williamsii</i> L.B.Sm.
Velloziaceae	<i>Vellozia epidendroides</i> Mart.
Verbenaceae	<i>Citharexylum spinosum</i> L.
Verbenaceae	<i>Lantana camara</i> L.
Verbenaceae	<i>Lantana nivea</i> Vent.

Family	Plant Species
Verbenaceae	<i>Stachytarpheta crassifolia</i> Schrad.
Verbenaceae	<i>Stachytarpheta gesnerioides</i> Cham.
Verbenaceae	<i>Stachytarpheta glabra</i> Cham.
Verbenaceae	<i>Stachytarpheta jamaicensis</i> (L.) Vahl
Verbenaceae	<i>Stachytarpheta mexiae</i> Moldenke
Verbenaceae	<i>Stachytarpheta quadrangula</i> Nees & Mart.
Verbenaceae	<i>Stachytarpheta urticifolia</i> (Salisb.) Sims
Violaceae	<i>Papaver hulkiana</i> Pulle
Vitaceae	<i>Cissus simsiana</i> Roem. & Schult.
Vitaceae	<i>Cissus verticillata</i> (L.) Nicolson & C.E.Jarvis
Vochysiaceae	<i>Qualea grandiflora</i> Mart.
Vochysiaceae	<i>Qualea multiflora</i> Mart.
Vochysiaceae	<i>Qualea parviflora</i> Mart.
Vochysiaceae	<i>Salvertia convallariaeodora</i> A.St.-Hil.
Vochysiaceae	<i>Vochysia cinnamomea</i> Pohl
Vochysiaceae	<i>Vochysia emarginata</i> Vahl
Vochysiaceae	<i>Vochysia pyramidalis</i> Mart.
Vochysiaceae	<i>Vochysia rufa</i> Mart.
Vochysiaceae	<i>Vochysia</i> sp.
Vochysiaceae	<i>Vochysia tucanorum</i> Mart.
Xanthorrhoeaceae	<i>Phormium tenax</i> J.R.Forst. & G.Forst.
Zingiberaceae	<i>Hedychium coronarium</i> J.Koenig
Zingiberaceae	<i>Renealmia alpinia</i> (Rottb.) Maas
Zingiberaceae	<i>Renealmia breviscapa</i> Poepp. & Endl.
Zingiberaceae	<i>Renealmia cernua</i> (Sw. ex Roem. & Schult.) J.F.Macbr.
Zingiberaceae	<i>Renealmia krukovi</i> Maas
Zingiberaceae	<i>Renealmia pyramidalis</i> (Lam.) Maas
Zingiberaceae	<i>Renealmia sessilifolia</i> Gagnep.
Zingiberaceae	<i>Renealmia</i> sp.
Zingiberaceae	<i>Zingiberaceae</i> sp.

### Hummingbird species

#### Phylogenetic lineage Hummingbird Species

Bee	<i>Archilochus colubris</i>
Bee	<i>Atthis heloisa</i>
Bee	<i>Calothorax lucifer</i>
Bee	<i>Chaetocercus mulsant</i>
Bee	<i>Mellisuga helenae</i>
Bee	<i>Mellisuga minima</i>
Bee	<i>Selasphorus calliope</i>
Bee	<i>Selasphorus flammula</i>
Bee	<i>Selasphorus platycercus</i>
Bee	<i>Selasphorus rufus</i>

	Phylogenetic lineage	Hummingbird Species
1	Bee	<i>Selasphorus sasin</i>
2	Bee	<i>Selasphorus scintilla</i>
3	Brilliant	<i>Aglaeactis castelnauii</i>
4	Brilliant	<i>Aglaeactis cupripennis</i>
5	Brilliant	<i>Boissonneaua flavescens</i>
6	Brilliant	<i>Boissonneaua jardini</i>
7	Brilliant	<i>Clytolaema rubricauda</i>
8	Brilliant	<i>Coeligena helianthea</i>
9	Brilliant	<i>Coeligena lutetiae</i>
10	Brilliant	<i>Coeligena prunellei</i>
11	Brilliant	<i>Coeligena torquata</i>
12	Brilliant	<i>Coeligena wilsoni</i>
13	Brilliant	<i>Ensifera ensifera</i>
14	Brilliant	<i>Eriocnemis cupreoventris</i>
15	Brilliant	<i>Eriocnemis derbyi</i>
16	Brilliant	<i>Eriocnemis mosquera</i>
17	Brilliant	<i>Eriocnemis vestita</i>
18	Brilliant	<i>Heliodoxa aurescens</i>
19	Brilliant	<i>Heliodoxa imperatrix</i>
20	Brilliant	<i>Heliodoxa jacula</i>
21	Brilliant	<i>Heliodoxa rubinoides</i>
22	Brilliant	<i>Lafresnaya lafresnayi</i>
23	Brilliant	<i>Ocreatus underwoodii</i>
24	Brilliant	<i>Pterophanes cyanopterus</i>
25	Brilliant	<i>Urosticte benjamini</i>
26	Coquette	<i>Adelomyia melanogenys</i>
27	Coquette	<i>Aglaiaocercus coelestis</i>
28	Coquette	<i>Aglaiaocercus kingii</i>
29	Coquette	<i>Chalcostigma herrani</i>
30	Coquette	<i>Heliangelus amethysticollis</i>
31	Coquette	<i>Lesbia nuna</i>
32	Coquette	<i>Lophornis chalybeus</i>
33	Coquette	<i>Lophornis magnificus</i>
34	Coquette	<i>Metallura tyrianthina</i>
35	Coquette	<i>Oreonympha nobilis</i>
36	Coquette	<i>Oreotrochilus estella</i>
37	Emerald	<i>Amazilia beryllina</i>
38	Emerald	<i>Amazilia brevirostris</i>
39	Emerald	<i>Amazilia candida</i>
40	Emerald	<i>Amazilia cyanifrons</i>
41	Emerald	<i>Amazilia cyanocephala</i>
42	Emerald	<i>Amazilia fimbriata</i>
43	Emerald	<i>Amazilia franciae</i>
44	Emerald	<i>Amazilia lactea</i>
45	Emerald	<i>Amazilia rutila</i>
46	Emerald	<i>Amazilia tobaci</i>
47	Emerald	<i>Amazilia tzacatl</i>

	Phylogenetic lineage	Hummingbird Species
1		
2	Emerald	<i>Amazilia versicolor</i>
3	Emerald	<i>Amazilia yucatanensis</i>
4	Emerald	<i>Calliphlox amethystina</i>
5	Emerald	<i>Calliphlox mitchellii</i>
6	Emerald	<i>Campylopterus curvipennis</i>
7	Emerald	<i>Campylopterus hemileucurus</i>
8	Emerald	<i>Campylopterus largipennis</i>
9	Emerald	<i>Chlorestes notata</i>
10	Emerald	<i>Chlorostilbon canivetii</i>
11	Emerald	<i>Chlorostilbon gibsoni</i>
12	Emerald	<i>Chlorostilbon lucidus</i>
13	Emerald	<i>Chlorostilbon maugea</i> ns
14	Emerald	<i>Chlorostilbon mellisugus</i>
15	Emerald	<i>Chlorostilbon olivaresi</i>
16	Emerald	<i>Chlorostilbon poortmani</i>
17	Emerald	<i>Chlorostilbon ricordii</i>
18	Emerald	<i>Chrysuronia oenone</i>
19	Emerald	<i>Cyanophrys bicolor</i>
20	Emerald	<i>Cynanthus latirostris</i>
21	Emerald	<i>Eupetomena macroura</i>
22	Emerald	<i>Eupherusa nigriventris</i>
23	Emerald	<i>Hylocharis chrysura</i>
24	Emerald	<i>Hylocharis cyanus</i>
25	Emerald	<i>Hylocharis leucotis</i>
26	Emerald	<i>Klais guimeti</i>
27	Emerald	<i>Leucochloris albicollis</i>
28	Emerald	<i>Orthorhyncus cristatus</i>
29	Emerald	<i>Phaeochroa cuvieri</i>
30	Emerald	<i>Stephanoxis lalandi</i>
31	Emerald	<i>Thalurania colombica</i>
32	Emerald	<i>Thalurania fannyi</i>
33	Emerald	<i>Thalurania furcata</i>
34	Emerald	<i>Thalurania glaukopis</i>
35	Emerald	<i>Trochilus polytmus</i>
36	Gem	<i>Eugenes fulgens</i>
37	Gem	<i>Heliodoxa constantii</i>
38	Gem	<i>Heliodoxa squamosus</i>
39	Gem	<i>Lampornis amethystinus</i>
40	Gem	<i>Lampornis calolaema</i>
41	Gem	<i>Lampornis castaneoventris</i>
42	Gem	<i>Lampornis clemenciae</i>
43	Gem	<i>Lampornis hemileucus</i>
44	Gem	<i>Lamprolaima rhami</i>
45	Gem	<i>Panterpe insignis</i>
46	Hermit	<i>Anopetia gounellei</i>
47	Hermit	<i>Eutoxeres aquila</i>
48	Hermit	<i>Glaucis hirsutus</i>

	Phylogenetic lineage	Hummingbird Species
1	Hermit	<i>Phaethornis bourcieri</i>
2	Hermit	<i>Phaethornis eurynome</i>
3	Hermit	<i>Phaethornis guy</i>
4	Hermit	<i>Phaethornis hispidus</i>
5	Hermit	<i>Phaethornis longirostris</i>
6	Hermit	<i>Phaethornis longuemareus</i>
7	Hermit	<i>Phaethornis malaris</i>
8	Hermit	<i>Phaethornis pretrei</i>
9	Hermit	<i>Phaethornis ruber</i>
10	Hermit	<i>Phaethornis squalidus</i>
11	Hermit	<i>Phaethornis striigularis</i>
12	Hermit	<i>Phaethornis stuarti</i>
13	Hermit	<i>Phaethornis subochraceus</i>
14	Hermit	<i>Phaethornis superciliosus</i>
15	Hermit	<i>Phaethornis syrmatophorus</i>
16	Hermit	<i>Ramphodon naevius</i>
17	Hermit	<i>Threnetes niger leucurus</i>
18	Hermit	<i>Threnetes ruckeri</i>
19	Mango	<i>Anthracothorax dominicus</i>
20	Mango	<i>Anthracothorax nigricollis</i>
21	Mango	<i>Anthracothorax prevostii</i>
22	Mango	<i>Anthracothorax viridis</i>
23	Mango	<i>Augastes lumachella</i>
24	Mango	<i>Augastes scutatus</i>
25	Mango	<i>Chrysolampis mosquitus</i>
26	Mango	<i>Colibri coruscans</i>
27	Mango	<i>Colibri delphinae</i>
28	Mango	<i>Colibri serrirostris</i>
29	Mango	<i>Colibri thalassinus</i>
30	Mango	<i>Doryfera ludoviciae</i>
31	Mango	<i>Eulampis holosericeus</i>
32	Mango	<i>Eulampis jugularis</i>
33	Mango	<i>Heliactin bilophus</i>
34	Mango	<i>Polytmus guainumbi</i>
35	Topaze	<i>Florisuga fusca</i>
36	Topaze	<i>Florisuga mellivora</i>
37	Topaze	<i>Topaza pyra</i>
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## Appendix S2 Phylogenetic methods

### Plant phylogenies

For each community we extracted the phylogenetic relatedness of plants from Phylomatic's megatree ([phylodiversity.net\phylomatic](http://phylodiversity.net/phylomatic)). Branch lengths were thereafter estimated with the *bladj* function from Phylocom (Webb *et al.*, 2008), using calibration dates from Wikström *et al.* (2001).

### Hummingbird phylogenies

For each community we extracted the phylogenetic relatedness of hummingbirds from the most complete to date hummingbird phylogeny, published in McGuire *et al.* (2014), which uses a time-calibrated, multi-locus Bayesian phylogenetic estimate constructed from DNA samples representing 284 hummingbird species.

The phylogenetic tree of McGuire *et al.* (2014) did not include eight hummingbird species represented in our networks, for which we had to approximate phylogenetic distances. Approximations were evaluated individually for each species, aiming to find the most accurate and conservative distance value in each case:

- Values for *Schistes geoffroy* were substituted for those of their most closely related species present in the phylogenetic tree: *Augastes lumachella* and *A. scutatus*.
- Values for *Amazilia saucerrottei* were substituted for those of their most closely related species present in the phylogenetic tree: *A. cyanifrons*.
- Values for *Campylopterus excellens* were substituted for those of their most closely related species present in the phylogenetic tree: *C. curvipennis*.
- Values for *Chlorostilbon mellisugus* were substituted for those of their most closely related species present in the phylogenetic tree: *C. gibsoni*.
- We loosely placed *Phaethornis gounellei* in the base of its clade using the minimum

1 value among *Phaethornis* species.  
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- 4 – We placed *P. stuarti* similarly using the minimum value of the “Pygmornis” clade within  
5 *Phaethornis*.  
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7 – Values for *Chlorostilbon olivaresi* and *Mellisuga helena*e were assigned as the average  
8 distance from the con-genera species present in the phylogenetic tree.  
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**16 Analysis of phylogenetic signal on bipartite interaction networks**  
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18 Following the methodology described in Ives & Godfray (2006), we examined whether closely  
19 related species interact with a similar array of partners. In other words, we quantified the  
20 effect of phylogeny on species interaction pattern. To do so, first, for each community and  
21 both for plant and hummingbirds, we used the above described phylogenies to calculate  
22 variance-covariance matrices (vcv), which represent the phylogenetic distance among  
23 species. Variance-covariance matrices were constructed using the “vcv” function from the R  
24 package *ape* (Paradis *et al.*, 2004). This approach assumes that the interaction frequency  
25 between plants and hummingbirds is given by the product of the vcv matrices. Therefore, for a  
26 given plant species *i*, the phylogenetic correlation between visits from hummingbirds *a* and *b*  
27 is given by the hummingbird phylogeny, and vice versa for hummingbirds. If both plants and  
28 hummingbirds are closely-related, visits between plant *i* and hummingbird *a* and plant *j* and  
29 hummingbird *b* are correlated. The strength of the phylogenetic association between plants  
30 and hummingbirds was then modeled statistically by fitting vcv matrices to each observed  
31 bipartite interaction matrix through a linear model using the “pbLM” function from the R  
32 package *picante* (Kembel *et al.*, 2010). This analysis results in a parameter *d* which  
33 determines the strength of the phylogenetic signal on species' interaction pattern. Specifically,  
34 this analysis provides two independent measures of phylogenetic signal (*d*), one for each  
35 trophic group ( $d_{plants}$  and  $d_{hummingbirds}$  in our case), calculated statistically through an estimated  
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2 generalized least squares model (EGLS), which does not assume that phylogenetic distance  
3 nor the interaction frequencies are normally distributed. Values of  $d=0$  represent a lack of  
4 phylogenetic correlation, whereas  $d=1$  represents a maximum correlation of phylogenetic  
5 signals. Disruptive selection processes show values above 1 (Ives & Godfray, 2006). We  
6 truncated the  $d$ -value to a maximum of  $d=2$ , as  $d_{\text{plants}}$  values for network 11 (Cuba) had an  
7 initial value of 6.43, being clearly an outlier (Table Appendix S2). Confidence intervals for  $d$   
8 were calculated by bootstrap replication data from standardized residuals through the same  
9 EGLS used to fit the original data over 2000 replication sets. Thereafter, we calculated the  
10 overall strength of the phylogenetic signal on the interaction matrix by evaluating three  
11 different models, one assuming no phylogenetic signal ( $d_{\text{plants}}=d_{\text{hummingbirds}}=0$ ; Star model), one  
12 assuming a maximum phylogenetic signal ( $d_{\text{plants}}=d_{\text{hummingbirds}}=1$ ; Brownian model), and a final  
13 one which incorporated the observed phylogenetic signals combined (estimated  $d_{\text{plants}}$  and  
14  $d_{\text{hummingbirds}}$ ; Data model). Again, these models were fitted through an EGLS approximation. To  
15 choose the model which best describes the data we quantified the goodness of fit of each  
16 model as mean squared error (MSE), so that the lowest value of MSE indicate the best fit.  
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19 Notice however that this analysis should not be acknowledged as a test for particular  
20 phylogenetic hypothesis, but rather a measure of how much does the phylogenetic  
21 relatedness among species affect their interaction pattern.

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2 **Appendix S2 Table A.** Results of the phylogenetic analysis for each community  
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Dataset ID number	MSE <sub>Dat</sub> a	MSE <sub>Star</sub>	MSE <sub>Brownia</sub> n	CI MIN		CI MAX			
				d <sub>plants</sub>	s	d <sub>plants</sub>	d <sub>hummingbir</sub> ds	d <sub>hummingbir</sub> ds	d <sub>hummingbir</sub> ds
1	15.468	19.632	36.420	0.000	0.094	0.452	0.002	0.243	0.467
2	9.887	13.973	16.665	0.000	0.293	0.639	0.000	0.310	2.395
3	25.700	25.277	183.797	0.955	0.883	1.056	0.000	0.000	0.114
4	14.194	15.611	34.123	0.000	0.161	0.403	0.000	0.043	0.257
5	54.965	59.220	101.878	0.000	0.263	0.642	0.000	0.092	0.330
6	2.270	4.220	5.354	0.354	0.700	1.018	0.000	0.000	0.156
7	109.205	116.143	207.956	0.000	0.252	0.521	0.000	0.053	0.284
8	69.545	78.534	122.920	0.000	0.093	0.364	0.000	0.295	0.574
9	16.398	21.877	40.001	0.181	0.587	1.009	0.000	0.000	0.119
10	139.116	139.116	336.877	0.000	0.000	0.000	0.000	0.000	0.000
11	137.292	247.429	242.988	0.931	2.000	17.038	0.000	0.098	0.765
12	6.258	7.458	13.106	0.034	0.654	1.280	0.000	0.000	0.293
13	36.780	37.552	55.636	0.000	0.041	0.637	0.000	0.119	0.787
14	37.913	38.519	60.741	0.000	0.041	1.873	0.000	0.120	1.400
15	16.017	16.433	25.504	0.000	0.151	2.000	0.000	0.000	0.838
16	1.454	1.642	2.379	0.068	0.290	0.499	0.000	0.155	0.319
17	1.017	1.577	1.124	0.089	1.056	2.452	0.000	0.261	1.888
18	4.161	4.471	5.857	0.118	0.706	1.387	0.000	0.065	0.377
19	12.602	21.129	18.718	0.042	1.119	3.076	0.000	0.060	0.514
20	28.627	28.718	43.951	0.000	0.000	0.734	0.000	0.000	0.458
21	1.552	2.167	2.627	0.032	0.383	0.804	0.057	0.310	0.520
22	1.761	2.415	3.948	0.268	1.703	5.247	0.000	0.000	0.291
23	1.490	1.785	1.933	0.000	0.000	0.877	0.000	0.396	0.706
24	63.498	67.344	119.495	0.071	0.420	0.841	0.000	0.006	0.268
25	9.717	9.889	21.005	0.000	0.046	0.232	0.000	0.052	0.213
26	7.456	7.581	15.548	0.000	0.045	0.396	0.000	0.000	0.237
27	2.885	3.426	3.618	0.168	0.397	0.599	0.000	0.292	0.779
28	41.405	42.071	63.699	0.000	0.112	0.423	0.000	0.053	0.306
29	15.717	18.415	29.404	0.025	0.611	1.223	0.000	0.000	0.246
30	52.055	56.779	132.613	0.000	0.009	0.331	0.000	0.084	0.413
31	3.738	4.052	9.265	0.057	0.306	0.546	0.000	0.000	0.095
32	2.873	3.296	4.361	0.056	0.271	0.455	0.056	0.267	0.436
33	0.603	0.723	0.865	0.000	0.066	0.379	0.026	0.405	0.674
34	2.385	2.397	5.426	0.000	0.000	0.186	0.000	0.000	0.130
35	12.454	19.826	23.805	0.257	1.150	2.362	0.000	0.013	0.330
36	6.404	7.762	7.988	0.000	0.240	0.827	0.000	0.499	1.383
37	10.555	10.837	13.897	0.000	0.284	2.633	0.000	0.000	1.158
38	6.022	6.340	11.381	0.007	0.214	0.430	0.000	0.028	0.197
39	23.433	23.874	44.968	0.000	0.000	0.272	0.000	0.136	0.390
40	1.956	2.205	3.601	0.132	0.350	0.551	0.000	0.000	0.142
41	1.616	2.121	2.510	0.108	0.235	0.359	0.266	0.409	0.514
42	1.716	2.047	2.875	0.000	0.009	0.296	0.047	0.388	0.607
43	1.075	1.083	2.293	0.000	0.045	0.178	0.000	0.000	0.085
44	1.783	2.562	2.289	0.157	0.668	1.275	0.001	0.340	0.609
45	1.165	1.168	1.778	0.000	0.000	0.150	0.001	0.143	0.316

Dataset ID number	$MSE_{Dat}$ a	$MSE_{Star}$	n	CI MIN		CI MAX	
				$d_{plants}$	s	$d_{plants}$	$d_{hummingbir}$
46	0.931	0.935	2.388	0.000	0.000	2.800	0.000
47	0.472	0.493	1.565	0.000	0.426	1.899	0.000
48	10.432	11.690	16.730	0.000	0.033	0.961	0.000
49	134.072	149.850	165.457	0.174	0.553	0.932	0.000
50	0.649	0.701	3.976	0.000	0.310	0.871	0.000
51	4.974	6.059	6.535	0.000	0.107	0.476	0.102
52	1.289	1.600	1.505	0.182	0.427	0.765	0.204
53	11.288	11.288	35.158	0.000	0.000	0.228	0.000
54	0.797	0.821	1.811	0.000	0.144	0.340	0.000

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2 **Appendix S3 Effect of the differences in sampling effort among data sets on network**  
3  
4 **structure**

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6 Network data were collected from various studies employing different sampling efforts, which  
7 could potentially bias resulting differences in network structure among communities. We  
8 measured sampling effort as sampling intensity, i.e. the ratio between number of interactions  
9 and the geometric mean of the number of plants and hummingbirds in the community as in  
10 Schleuning et al. (2014). Simple correlations presented in Table S2 show a relationship  
11 between sampling intensity and corrected complementary specialization and with observed  
12 modularity, both in the global and mainland datasets. Therefore, in order to reveal whether  
13 these correlations had any repercussions in the conclusions obtained from the macroecological  
14 models, we repeated the macroecological models presented in the main manuscript including  
15 sampling intensity as a predictor variable (Table Appendix S3).

16 Overall trends remained constant. For corrected complementary specialization ( $H_2'$ ), the  
17 inclusion of sampling intensity had no overall effects on the other predictor variables in the  
18 global model, while in the mainland analysis network size lost some of its importance, but the  
19 rank of importance remained the same. For observed modularity, the only difference was that  
20  $d_{\text{hummingbirds}}$  lost some of its importance in the global dataset, while insularity became an  
21 important predictor, with islands showing a lower modularity than mainland communities  
22 (Table Appendix S3), confirming findings from other performed complementary  
23 macroecological analysis (Appendix S4) which suggest differences between mainland and  
24 island communities.

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30 **REFERENCES**

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1  
2 Appendix S3 Table A. Relationship between species richness, phylogenetic signal and environment, and corrected complementary specialization ( $H_2'$ ) and observed modularity (QuanBiMo) across the Americas (Global  
3 dataset) and the mainland, i.e. when excluding insular communities (Only Mainland), when including Sampling effort as a predictor variable, as these network metrics appeared to be correlated with sampling intensity  
4 (Table S2). For each network metric, the standardized regression coefficients are reported for ordinary least squares regression for a model including all variables ("Full") and for a model averaged across all possible  
5 models using Akaike Weights (AICc  $w_i$  - "Averaged"). For the averaged models, the relative importance of each predictor variable ("Σ  $w_i$ ") is given by the Akaike Weights (AICc  $w_i$ ). Predictors with high importance are  
6 marked in bold. Phylogenetic signal among plants and hummingbirds,  $d_p$  and  $d_h$ , respectively; Mean annual temperature, MAT; Temperature seasonality, TS; Mean annual precipitation, MAP; Precipitation seasonality, PS;  
7 Velocity of temperature change since the last Glacial Maximum, MAT velocity; Velocity of precipitation change since the last Glacial Maximum, MAP velocity. The significance of Moran's  $I$  was tested with 10 distance  
8 classes and a permutation test with 10,000 iterations. In all models multicollinearity was not an issue, i.e. VIF ≤ 3.3, CN ≤ 4.2, and there was no positive spatial autocorrelation, i.e. Moran's  $I$  was non-significant.  
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			Model	AICc	R <sup>2</sup>	Network size	Species richness	Phylogenetic signal				Environmental factors								Sampling effort
								Phylogenetic signal				Environmental factors								
								d <sub>p</sub>	d <sub>h</sub>	Insularity	MAT	TS	MAP	PS	MAT velocity	MAP velocity	Sampling intensity			
Global dataset (54 networks)	<i>H<sub>2</sub>'</i> corrected	Full	-60.862	0.620	0.346	-0.027	<b>-0.228</b>	-0.311	0.285	0.138	0.237	-0.165	-0.485	0.140	0.341					
		Averaged			0.415	-0.089	<b>-0.297</b>	-0.245	0.200	0.103	0.305	-0.206	-0.289	0.069	0.342					
		Σ $w_i$			0.956	0.264	<b>0.868</b>	0.521	0.385	0.272	0.646	0.516	0.734	0.254	0.944					
	QuanBiMo observed	Full	-69.765	0.658	0.371	-0.170	-0.181	<b>-0.373</b>	0.202	-0.044	0.138	-0.087	-0.052	-0.085	-0.289					
		Averaged			0.380	-0.202	-0.225	<b>-0.345</b>	0.192	-0.071	0.305	-0.098	-0.063	-0.091	-0.304					
		Σ $w_i$			0.953	0.653	0.659	<b>0.886</b>	0.447	0.235	0.710	0.279	0.252	0.266	0.872					
	Only Mainland (45 networks)	<i>H<sub>2</sub>'</i> corrected	Full	-59.753	0.652	0.212	-0.045	<b>-0.281</b>	na	<b>0.478</b>	0.091	0.114	-0.223	<b>-0.708</b>	0.240	0.344				
		Averaged			0.249	-0.057	<b>-0.305</b>	na	<b>0.512</b>	0.044	0.208	-0.266	<b>-0.680</b>	0.277	0.370					
		Σ $w_i$			0.641	0.195	<b>0.871</b>	na	<b>0.961</b>	0.191	0.340	0.736	<b>0.985</b>	0.614	0.956					
		Full	-54.411	0.619	0.262	-0.217	<b>-0.280</b>	na	0.313	-0.118	0.052	-0.103	-0.184	-0.075	-0.311					
		Averaged			0.309	-0.239	<b>-0.292</b>	na	0.287	-0.168	0.327	-0.105	-0.221	-0.086	-0.347					
		Σ $w_i$			0.756	0.694	<b>0.754</b>	na	0.583	0.356	0.549	0.262	0.373	0.237	0.809					

**Appendix S4 Results of macroecological models on observed network metrics**

As an additional analysis, we performed an equivalent macroecological analysis on observed network metrics instead of on null model corrected metrics as presented in the main manuscript. By analyzing observed network metrics, we are investigating the "raw" level of specialization, regardless of the community context (species and interaction richness, interaction frequency distribution). This complementary analysis allow us to understand more fully the effect of correcting for network metrics and the relationships between the studied variables (Tables S4A and S4B).

The amount of variability explained by the models is similar to that of corrected metrics, but important differences in the role of the different predictor variables arose. First of all, species richness lost its high importance except for modularity in the global dataset. In other words, in general, larger communities did not show higher levels of observed specialization (complementary specialization and modularity). Together with the results from the corrected-metrics analysis, where species richness exhibited a strong and widespread association with higher levels of specialization, we can conclude that *only when compared to communities with the same attributes as empirical matrices* do species richness impact network structure.

Similarly, hummingbird phylogenetic signal, the other variable exhibiting a constant association with corrected network structure, is only related with complementary specialization, and not modularity, when examining observed metrics. As our null model correction does not account for species' phylogenetic relatedness, we may conclude that higher levels of corrected modularity are indeed achieved when closely-related hummingbirds form distinct modules.

As with corrected metrics, mean annual temperature and temperature velocity were related only to higher levels of complementary specialization on mainland communities, both

1 showing the highest regression coefficients of all predictors in the full models.  
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Additionaly, insularity and mean annual precipitation became important determinants of observed specialization, with mainland and humid communities showing higher levels of complementary specialization and modularity than insular communities or networks from drier environments. The effect of these predictor variables disappeared when modeling corrected metrics, suggesting that insularity and MAP may affect either species/interaction richness or interaction distribution, leading to higher observed levels of specialization (Dalsgaard *et al.*, 2011).

## REFERENCES

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For Review

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 2 Appendix S4 Table A. Relationship between species richness, phylogenetic signal and environment, and complementary specialization ( $H_2'$ ) and modularity (QuanBiMo) across the Americas (Global dataset) and the  
 3 mainland, i.e. when excluding insular communities (Only Mainland). Complementary specialization measures the exclusiveness of the interactions in the community, whereas modularity describes a community pattern  
 4 where species interact more frequently within subsets of species and loosely among subsets. For each network metric, the standardized regression coefficients are reported for ordinary least squares regression for a  
 5 model including all variables ("Full") and for a model averaged across all possible models using Akaike Weights (AICc  $w_i$  - "Averaged"). For the averaged models, the relative importance of each predictor variable (" $\Sigma w_i$ ")  
 6 is given by the Akaike Weights (AICc  $w_i$ ). Phylogenetic relatedness among plants and hummingbirds,  $d_p$  and  $d_h$ , respectively; Mean annual temperature, MAT; Temperature seasonality, TS; Mean annual precipitation,  
 7 MAP; Precipitation seasonality, PS; Velocity of temperature change since the last Glacial Maximum, MAT velocity; Velocity of precipitation change since the last Glacial Maximum, MAP velocity. The significance of  
 8 Moran's  $I$  was tested with 10 distance classes and a permutation test with 10,000 iterations. In all models multicollinearity was not an issue, i.e. VIF  $\leq 4.0$ , CN  $\leq 3.3$ , and there was no positive spatial autocorrelation, i.e.  
 9 Moran's  $I$  was non-significant.

			Species richness			Phylogenetic signal			Environmental factors						
			Model	AICc	$R^2$	Network size	$d_p$	$d_h$	Insularity	MAT	TS	MAP	PS	MAT velocity	
15 16 17 18 19 20 21 22 23	Global dataset (54 networks)	$H_2'$	Full	-35.112	0.578	0.210	-0.069	<b>-0.231</b>	<b>-0.453</b>	0.355	0.085	<b>0.344</b>	-0.085	-0.317	0.129
			Averaged			0.249	-0.119	<b>-0.254</b>	<b>-0.401</b>	0.290	0.040	<b>0.423</b>	-0.101	-0.225	0.115
			$\Sigma w_i$			0.659	0.323	<b>0.767</b>	<b>0.916</b>	0.573	0.218	<b>0.965</b>	0.268	0.417	0.295
	QuanBiMo	Full	-65.124	0.602	<b>0.273</b>	-0.119	-0.117	<b>-0.449</b>	0.203	-0.076	<b>0.301</b>	-0.049	-0.037	-0.128	
			Averaged			<b>0.259</b>	-0.139	-0.112	<b>-0.446</b>	0.202	-0.098	<b>0.446</b>	-0.028	-0.063	-0.095
			$\Sigma w_i$			<b>0.794</b>	0.411	0.324	<b>0.993</b>	0.483	0.276	<b>0.989</b>	0.216	0.261	0.280
	Only Mainland (45 networks)	$H_2'$	Full	-31.253	0.574	0.076	-0.067	<b>-0.300</b>	na	<b>0.518</b>	0.001	0.238	-0.105	<b>-0.491</b>	0.197
			Averaged			0.145	-0.094	<b>-0.382</b>	na	<b>0.525</b>	-0.065	0.327	-0.167	<b>-0.475</b>	0.223
			$\Sigma w_i$			0.325	0.250	<b>0.858</b>	na	<b>0.924</b>	0.223	0.731	0.351	<b>0.832</b>	0.477
	QuanBiMo	Full	-51.864	0.561	0.159	-0.221	-0.234	na	0.314	-0.149	<b>0.232</b>	-0.069	-0.179	-0.136	
			Averaged			0.181	-0.232	-0.210	na	0.298	-0.181	<b>0.455</b>	-0.047	-0.230	-0.101
			$\Sigma w_i$			0.412	0.637	0.496	na	0.585	0.383	<b>0.938</b>	0.221	0.384	0.261

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**Appendix S4 Table B.** Coefficients of determination ( $R^2$ ) from partial regression local models separating the effect of species richness, species' evolutionary history and environmental factors as predictors of corrected network structure of American communities, when including climate change velocity as an estimate of historical climate stability. We estimated  $R^2$  values in "full" models, i.e. in models including all ten predictor variables (see Appendix S4 Table A). "Total model" reflects the total amount of variation in network structure explained by the full model; "Total species richness", "Total phylogenetic signals" and "Total environmental factors" reflect the variation explained by species richness, phylogenetic signal and environmental factors, respectively; "Only species richness", "Only phylogenetic signal" and "Only environmental factors" reflect the amount of variation in network structure explained by the specified predictor-type and not by any other (i.e. the unique variation); Finally, "shared" reflects the variation attributable to the pair of specified predictor types or to all factors combined. See Table S4A for standardized coefficients of each variable and more details of model fit.

	Global dataset (54 networks)	Only Mainland (45 networks)		
	$H_2'$ observed	QuanBiMo observed	$H_2'$ observed	QuanBiMo observed
Total model	0.58	0.60	0.57	0.56
Total species richness	0.19	0.27	0.11	0.14
Total phylogenetic signal	0.23	0.18	0.25	0.23
Total environmental factors	0.52	0.54	0.51	0.48
Unique species richness	0.03	0.04	0.00	0.02
Unique phylogenetic signal	0.04	0.02	0.06	0.07
Unique environmental factors	0.19	0.21	0.23	0.23
Shared species richness – phylogeny	0.00	0.00	0.00	0.00
Shared species richness – environmental factors	0.14	0.17	0.09	0.08
Shared phylogeny – environmental factors	0.16	0.11	0.17	0.12
Shared all factors	0.04	0.06	0.02	0.05

## 1 2 Appendix S5 Complementary macroecological analyses 3

4 As different climate change metrics may capture different aspects important for biodiversity  
5 (Garcia *et al.*, 2014), to evaluate more comprehensively the impact of historical climate-  
6 change we calculated another summarizations of historical climatic stability: *the anomaly of*  
7 *mean annual temperature and mean annual precipitation* between the LGM and the present,  
8 calculated as pre-industrial minus LGM precipitation/temperature (i.e. positive values indicate  
9 that conditions were drier or colder at LGM than at present). Whereas climate change  
10 velocities reflect the speed of climate change irrespective of the direction of the change,  
11 anomalies reflect the direction of climate change. As with climate change velocities,  
12 projections of the global climate during the LGM and pre-industry were generated by Hadley  
13 Centre Coupled Model Version 3 (HadCM3) with a resolution of  $3.75 \times 2.5$  arc degrees  
14 (Singarayer & Joy 2010), and were downscaled to  $0.1 \times 0.1$  arc degrees.  
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17 We performed a set of models equivalent to those presented in the main text (Diniz-  
18 Filho *et al.* 2008) but which included climate change anomaly instead of climate change  
19 velocity. Temperature anomaly was  $\log_{10}$ -transformed. Models including climatic anomaly  
20 gave a poorer fit to our data, explaining less variability and raising multicollinearity concerns.  
21 Nevertheless, strong patterns disclosed by the macroecological models with climatic velocity  
22 persisted (Tables S4A-B).  
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25 The significance and importance of species richness and hummingbird phylogenetic  
26 signal was again widespread **and showed similar values than the models which included**  
27 **change velocity as measure of historical climate stability**. We found a stronger role, although  
28 still non-significant, of the phylogenetic signal of plants, associated with lower levels of  
29 modularity on both datasets. Finally, and most importantly, neither temperature nor  
30 precipitation anomaly had any effect on any of the studied variables.  
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**Appendix S5 Table A.** Relationship between species richness, phylogenetic signal and environment, and complementary specialization ( $H_2'$ ) and modularity (QuanBiMo) across the Americas (Global dataset) and the mainland, i.e. when excluding insular communities (Only Mainland), when including historical climate anomaly as a predictor of historical climatic stability. Complementary specialization measures the exclusiveness of the interactions in the community, whereas modularity describes a community pattern where species interact more frequently within subsets of species and loosely among subsets. For each network metric, the standardized regression coefficients are reported for ordinary least squares regression for a model including all variables ("Full") and for a model averaged across all possible models using Akaike Weights (AICc  $w_i$  - "Averaged"). For the averaged models, the relative importance of each predictor variable ( $\sum w_i$ ) is given by the Akaike Weights (AICc  $w_i$ ). Phylogenetic signal among plants and hummingbirds,  $d_p$  and  $d_h$ , respectively; Mean annual temperature, MAT; Temperature seasonality, TS; Mean annual precipitation, MAP; Precipitation seasonality, PS; Velocity of temperature change since the last Glacial Maximum, MAT velocity; Velocity of precipitation change since the last Glacial Maximum, MAP velocity. The significance of Moran's  $I$  was tested with 10 distance classes and a permutation test with 10,000 iterations. In all models [multicollinearity was not an issue](#), i.e. VIF  $\leq 5.1$ , CN  $\leq 4.9$ , and [there was no positive spatial autocorrelation](#), i.e. Moran's  $I$  was non-significant.

			Species richness		Phylogenetic signal		Environmental factors								
			Model	AICc	R <sup>2</sup> <sub>adj</sub>	Network size	d <sub>p</sub>	d <sub>h</sub>	Insularity	MAT	TSN	MAP	PSN	MAT anomaly	MAP anomaly
Global dataset (54 networks)	$H_2'$	Full	-46.828	0.474	0.512	-0.168	-0.306	-0.046	-0.004	0.111	0.026	-0.251	0.009	-0.046	
		Averaged			0.516	-0.165	-0.328	-0.067	-0.024	0.080	0.101	-0.249	0.004	-0.023	
		$\Sigma w_i$			0.999	0.458	0.936	0.240	0.218	0.261	0.276	0.738	0.216	0.224	
	QuanBiMo	Full	-99.080	0.619	0.630	-0.186	-0.237	-0.210	0.110	0.075	-0.087	-0.242	0.022	-0.217	
		Averaged			0.636	-0.194	-0.277	-0.141	0.024	0.054	0.118	-0.214	-0.053	-0.124	
		$\Sigma w_i$			1.000	0.667	0.908	0.362	0.227	0.239	0.317	0.696	0.242	0.343	
	Only Mainland (45 networks)	$H_2'$	Full	-45.189	0.477	0.422	-0.146	-0.337	na	0.135	0.019	-0.388	-0.350	0.216	-0.215
		Averaged			0.424	-0.116	-0.330	na	0.079	0.039	-0.067	-0.275	-0.218	-0.047	
		$\Sigma w_i$			0.968	0.282	0.839	na	0.237	0.217	0.246	0.695	0.479	0.229	
	QuanBiMo	Full	-76.043	0.558	0.539	-0.217	-0.301	na	0.169	0.028	-0.285	-0.291	-0.081	-0.332	
		Averaged			0.534	-0.192	-0.293	na	0.101	0.019	0.079	-0.251	-0.135	-0.183	
		$\Sigma w_i$			0.999	0.510	0.794	na	0.265	0.209	0.282	0.676	0.312	0.410	

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**Appendix S5 Table B.** Coefficients of determination ( $R^2$ ) from partial regression local models separating the effect of species richness, environmental factors and species' evolutionary history as predictors of corrected network structure of American communities when using climatic anomaly as an estimate for historical climatic stability. We estimated  $R^2$  values in "full" models, i.e. in models including all ten predictor variables (see Appendix S5 Table A for standardized coefficients of each variable and more details of model fit). "Total model" reflects the total amount of variation in network structure explained by the model; "Total species richness", "Total phylogenetic signals" and "Total environmental factors" reflect the total variation explained by species richness, phylogenetic signal and environmental factors, respectively, whereas "Only species richness", "Only phylogenetic signal" and "Only environmental factors" reflect the unique variation in network structure explained by species richness, phylogenetic signal and environmental factors, respectively. Finally, "shared" reflect the shared variation described by the specified predictor sets, and "shared all factors" reflect variation attributable to species richness, phylogenetic signal and environmental factors.

	Global dataset (54 networks)		Only Mainland (45 networks)	
	$H_2'$	QuanBiMo	$H_2'$	QuanBiMo
Total model	0.47	0.62	0.48	0.56
Total species richness	0.25	0.44	0.20	0.33
Total phylogenetic signal	0.19	0.19	0.18	0.19
Total environmental factors	0.25	0.32	0.28	0.28
Unique species richness	0.16	0.24	0.12	0.20
Unique phylogenetic signal	0.08	0.06	0.08	0.08
Unique environmental factors	0.06	0.06	0.11	0.09
Shared species richness – phylogeny	0.00	0.00	0.00	0.00
Shared species richness – environmental factors	0.07	0.14	0.06	0.09
Shared phylogeny – environmental factors	0.08	0.07	0.08	0.06
Shared all factors	0.04	0.06	0.03	0.05

Table S1. Coordinates, description, location and data references for each studied community.

ID number	Latitude	Longitude	Site description and general location	Data Source Reference
1	-20.75	-42.92	Atlantic forest, SE Brazil	Abreu, C.R.M. & Vieira, M.F. (2004) Os beija-flores e seus recursos florais em um fragmento florestal de Viçosa, sudeste brasileiro. <i>Lundiana</i> , <b>5</b> , 129–134.
2	34.22	-116.95	Montane meadow, SW USA	Alarcón, R., Waser, N.M & Ollerton, J. (2008) Year-to-year variation in the topology of a plant-pollinator interaction network. <i>Oikos</i> , <b>117</b> , 1796–1807.
3	38.98	-106.97	Sub-alpine meadow, Central USA	Ollerton, J., Alarcón, R., Waser, N.M, Price, M.V., Watts, S., Cranmer, L., Hingston, A., Peter, C. & Rotenberry, J. (2009) A global test of the pollination syndrome hypothesis. <i>Annals of Botany</i> , <b>103</b> , 1471–1480.
4	-23.35	-44.83	Atlantic forest, SE Brazil	Araujo, A.C. (1996) Beija-flores e seus recursos florais numa área de planicie costeira do litoral norte de São Paulo, sudeste do Brasil. <i>MSc Thesis</i> . Universidade Estadual de Campinas, Brazil.
5	19.50	-105.05	Tropical dry deciduous forest, W Mexico	Arizmendi, M.C. & Ornelas, J.F. (1990) Hummingbirds and their floral resources in a tropical dry forest in Mexico. <i>Biotropica</i> , <b>22</b> , 172–180.
6	-31.80	-52.42	Pampa, S Brazil	Vizentin-Bugoni, J. & Rui, A.M. <i>Unpublished data</i> .
7	-22.50	-44.83	Highland Atlantic forest, SE Brazil	Canela, M.B.F. (2006) Interações entre plantas e beija-flores numa comunidade de floresta atlântica montana em Itatiaia, RJ. <i>Ph.D thesis</i> . Universidade Estadual de Campinas, Brazil.
8	-7.87	-36.40	Caatinga, NE Brazil	Las Casas, F.M.G., Azevedo Júnior, S.M. & Dias Filho, M.M. (2012) The community of hummingbirds (Aves: Trochilidae) and the assemblage of flowers in a Caatinga vegetation. <i>Brazilian Journal of Biology</i> , <b>72</b> , 51–58.
9	-11.48	-41.32	Altitudinal caatinga, NE Brazil	Moura, A.C. & Machado, C.G. Hummingbirds and their flowers at altitudinal dryland vegetation in Chapada Diamantina, northeast Brazil. <i>In preparation</i> .
10	-3.82	-70.27	Amazonian rainforest, SE Colombia	Cotton, P.A. (1998) The hummingbird community of a lowland Amazonian rainforest. <i>Ibis</i> , <b>140</b> , 512–521.
11	22.28	-81.20	Swamp forest, Hurricane disturbed, Cuba	Baquero, A.C. (2014) Evolutionary and ecological insight into hummingbird-plant communities in the Caribbean. <i>MSc Thesis</i> . University of Copenhagen, Denmark.
12	15.35	-61.30	Elfin forest, Dominica	Dalsgaard, B., Martín González, A.M., Olesen, J.M., Ollerton, J., Timmermann, A., Andersen, L.H. & Tossas, A.G. (2009) Plant–hummingbird interactions in the West Indies: floral specialisation gradients associated with environment and hummingbird size. <i>Oecologia</i> , <b>159</b> , 757–766.
13	15.25	-61.37	Coastal dry scrubland, Dominica	Dalsgaard, B., Martín González, A.M., Olesen, J.M., Ollerton, J., Timmermann, A., Andersen, L.H. & Tossas, A.G. (2009) Plant–hummingbird interactions in the West Indies: floral specialisation gradients associated with environment and hummingbird size. <i>Oecologia</i> , <b>159</b> , 757–766.

ID number	Latitude	Longitude	Site description and general location	Data Source Reference
14	12.10	-61.70	Elfin forest, Grenada	Dalsgaard, B., Martín González, A.M., Olesen, J.M., Ollerton, J., Timmermann, A., Andersen, L.H. & Tossas, A.G. (2009) Plant–hummingbird interactions in the West Indies: floral specialisation gradients associated with environment and hummingbird size. <i>Oecologia</i> , <b>159</b> , 757–766.
15	12.10	-61.68	Rainforest, Grenada	Dalsgaard, B., Martín González, A.M., Olesen, J.M., Ollerton, J., Timmermann, A., Andersen, L.H. & Tossas, A.G. (2009) Plant–hummingbird interactions in the West Indies: floral specialisation gradients associated with environment and hummingbird size. <i>Oecologia</i> , <b>159</b> , 757–766.
16	1.25	-77.43	High andean forest, Colombia	Gutierrez Zamora, E.A. & Rojas Nossa, S.V. (2001) Dinámica anual de la interacción colibrí-flor en ecosistemas altoandinos del volcán Galeras, Sur de Colombia. <i>BSc. Thesis</i> . Universidad Nacional de Colombia, Colombia.
17	18.35	-77.65	Evergreen lowland rainforest, Jamaica	Ingversen, T. (2006) Plant-Pollinator Interactions on Jamaica and Dominica – The Centrality, Asymmetry, and Modularity of Networks. <i>MSc. Thesis</i> . University of Aarhus, Denmark.
18	-27.27	-49.01	Atlantic Forest, low elevation	Kohler, G. (2011) Redes de interação planta-beija-flor em um gradiente altitudinal de Floresta Atlântica no Sul do Brasil. <i>MSc. Thesis</i> . Universidade Federal do Paraná, Brazil.
19	-27.26	-49.02	Atlantic Forest, mid elevation	Kohler, G. (2011) Redes de interação planta-beija-flor em um gradiente altitudinal de Floresta Atlântica no Sul do Brasil. <i>MSc. Thesis</i> . Universidade Federal do Paraná, Brazil.
20	-27.26	-49.02	Atlantic Forest, high elevation	Kohler, G. (2011) Redes de interação planta-beija-flor em um gradiente altitudinal de Floresta Atlântica no Sul do Brasil. <i>MSc. Thesis</i> . Universidade Federal do Paraná, Brazil.
21	19.23	-98.97	Highland temperate mosaic forest, Central Mexico	Lara, C. (2006) Temporal dynamics of flower use by hummingbirds in a highland temperate forest in México. <i>Ecoscience</i> , <b>13</b> , 23–29.
22	19.28	-98.23	Suburban forest, Central Mexico	Lara, C. <i>Unpublished data</i> .
23	19.40	-96.80	Protected cloud forest, Central Mexico	Lara, C. <i>Unpublished data</i> .
24	0.07	-72.45	Rainforest, Colombia	Rosero, L. (2003) Interações planta/beija-flor em três comunidades vegetais da parte sul do Parque Nacional Natural Chiribiquete, Amazonas (Colombia). <i>Ph.D. Thesis</i> . Universidade Estadual de Campinas, Brazil. Rosero-Lasprilla, L. & Sazima, M. (2004) Interacciones planta-colibrí en tres comunidades vegetales de la parte suroriental del Parque Nacional Natural Chiribiquete, Colombia. <i>Ornitología Neotropical</i> , <b>15</b> (Suppl.), 183–190.
25	-13.12	-41.58	Highland caatinga, NE Brazil	Machado, C.G. (2009) Hummingbirds (Aves: Trochilidae) and their floral resources in an area of caatinga vegetation in the Chapada Diamantina, Bahia State, Northeast Brazil. <i>Zoologia</i> , <b>26</b> , 55–65.
26	-13.12	-41.57	Open cerrado, NE	Machado, C.G. (2014) The hummingbird community and the plants which they visit at a savannah in the

1	4	ID number	Latitude	Longitude	Site description and general location	Data Source Reference
2	5				Brazil	Chapada Diamantina, Bahia, Brazil. <i>Bioscience Journal</i> , <b>30</b> , 1578–1587.
3	6	27	-12.98	-41.33	Campos rupestres, NE Brazil	Machado, C.G., Coelho, A.G., Santana, C.S., Rodrigues, M. (2007) Hummingbirds and their flowers in the 'campos rupestres' of Chapada Diamantina, Bahia, northeastern Brazil. <i>Revista Brasileira de Ornitológia</i> , <b>15</b> , 215–227.
4	7	28	10.44	-84.01	Cloud forest, low elevation, Costa Rica	Maglianesi, M.A., Blüthgen, N., Böhning-Gaese, K. & Schleuning, M. (2014) Morphological traits determine specialization and resource use in plant–hummingbird networks in the Neotropics. <i>Ecology</i> , <a href="http://dx.doi.org/10.1890/13-2261.1">http://dx.doi.org/10.1890/13-2261.1</a>
5	8	29	10.27	-84.08	Cloud forest, mid elevation, Costa Rica	Maglianesi, M.A., Blüthgen, N., Böhning-Gaese, K. & Schleuning, M. (2014) Morphological traits determine specialization and resource use in plant–hummingbird networks in the Neotropics. <i>Ecology</i> , <a href="http://dx.doi.org/10.1890/13-2261.1">http://dx.doi.org/10.1890/13-2261.1</a>
6	9	30	10.18	-84.11	Cloud forest, high elevation, Costa Rica	Maglianesi, M.A., Blüthgen, N., Böhning-Gaese, K. & Schleuning, M. (2014) Morphological traits determine specialization and resource use in plant–hummingbird networks in the Neotropics. <i>Ecology</i> , <a href="http://dx.doi.org/10.1890/13-2261.1">http://dx.doi.org/10.1890/13-2261.1</a>
7	10	31	-18.99	-48.30	Cerrado, Central Brazil	Araújo, F.P., Sazima, M. & Oliveira, P.E. (2013) The assembly of plants used as nectar sources by hummingbirds in a Cerrado area of Central Brazil. <i>Plant Systematics and Evolution</i> , <b>299</b> , 1119–1133. Machado, A.O. (2012) Diversidade de recursos florais para beija-flores nos cerrados do Triângulo mineiro e região. <i>Ph.D Thesis</i> . Universidade Federal de Uberlândia, Brazil. Maruyama, P.K., Oliveira, G.M., Ferreira, C., Dalsgaard, B. & Oliveira, P.E. (2013) Pollination syndromes ignored: importance of non-ornithophilous flowers to Neotropical savanna hummingbirds. <i>Naturwissenschaften</i> , <b>100</b> , 1061–1068.
8	11	32	-19.16	-48.39	Cerrado, Central Brazil	Araújo, F.P., Sazima, M. & Oliveira, P.E. (2013) The assembly of plants used as nectar sources by hummingbirds in a Cerrado area of Central Brazil. <i>Plant Systematics and Evolution</i> , <b>299</b> , 1119–1133. Machado, A.O. (2012) Diversidade de recursos florais para beija-flores nos cerrados do Triângulo mineiro e região. <i>Ph.D Thesis</i> . Universidade Federal de Uberlândia, Brazil. Maruyama, P.K., Oliveira, G.M., Ferreira, C., Dalsgaard, B. & Oliveira, P.E. (2013) Pollination syndromes ignored: importance of non-ornithophilous flowers to Neotropical savanna hummingbirds. <i>Naturwissenschaften</i> , <b>100</b> , 1061–1068.
9	12	33	-17.78	-48.68	Cerrado, Central Brazil	Machado, A.O. (2012) Diversidade de recursos florais para beija-flores nos cerrados do Triângulo mineiro e região. <i>Ph.D Thesis</i> . Universidade Federal de Uberlândia, Brazil.
10	13	34	19.50	-96.95	Protected cloud forest, Central Mexico	Ornelas, J.F. <i>Unpublished data</i> .
11	14	35	-19.52	-56.98	Pantanal wetland, SW Brazil	Araujo, AC & Sazima, M. 2003. The assemblage of flowers visited by hummingbirds in the "capões" of Southern Pantanal, Mato Grosso do Sul, Brazil. <i>Flora</i> , <b>198</b> , 427–435

ID number	Latitude	Longitude	Site description and general location	Data Source Reference
36	18.13	-66.76	Elfin forest, Puerto Rico	Dalsgaard, B., Martín González, A.M., Olesen, J.M., Ollerton, J., Timmermann, A., Andersen, L.H. & Tossas, A.G. (2009) Plant–hummingbird interactions in the West Indies: floral specialisation gradients associated with environment and hummingbird size. <i>Oecologia</i> , <b>159</b> , 757–766.
37	17.95	-66.82	Highly preserved coastal dry scrubland, Puerto Rico	Dalsgaard, B., Martín González, A.M., Olesen, J.M., Ollerton, J., Timmermann, A., Andersen, L.H. & Tossas, A.G. (2009) Plant–hummingbird interactions in the West Indies: floral specialisation gradients associated with environment and hummingbird size. <i>Oecologia</i> , <b>159</b> , 757–766.
38	-19.25	-43.52	Campos rupestres, SE Brazil	Rodrigues, L.C. & Rodrigues, M. (2014) Flowers visited by hummingbirds in the open habitats of the southeastern Brazilian moutaintops: species composition and seasonality. <i>Brazilian Journal of Biology</i> , <b>74</b> , 659–676.
39	-20.44	-54.65	Cerrado, W Brazil	Rodrigues, L.C. & Araujo, A.C. (2011) The hummingbird community and their floral resources in an urban forest remnant in Brazil. <i>Brazilian Journal of Biology</i> , <b>71</b> , 611–622.
40	-22.73	-45.58	Montane Forest, SE Brazil	Sazima, I., Buzato, S. & Sazima, M. (1996) An assemblage of hummingbird-pollinated flowers in a montane forest in southeastern Brazil. <i>Botanica Acta</i> , <b>109</b> , 149–160.
41	10.67	-61.28	Mixed forest, Trinidad	Snow, B.K. & Snow, D.W. (1972) Feeding niches of humingbirds in a Trinidad Valley. <i>The Journal of Animal Ecology</i> , <b>41</b> , 471–485.
42	4.53	-73.85	Andean forest, Colombia	Snow, D.W. & Snow, B.K. (1980) Relationships between hummingbirds and flowers in the Andes of Colombia. <i>Bulletin of the British Museum of Natural History (Zoology)</i> , <b>38</b> , 105–139.
43	5.90	-73.42	Andean forest, Colombia	Snow, D.W. & Snow, B.K. (1980) Relationships between hummingbirds and flowers in the Andes of Colombia. <i>Bulletin of the British Museum of Natural History (Zoology)</i> , <b>38</b> , 105–139.
44	5.92	-73.53	Andean forest, Colombia	Snow, D.W. & Snow, B.K. (1980) Relationships between hummingbirds and flowers in the Andes of Colombia. <i>Bulletin of the British Museum of Natural History (Zoology)</i> , <b>38</b> , 105–139.
45	-23.63	-45.85	Coastal cloud forest, SE Brazil	Snow D.W. & Snow, B.K. (1986) Feeding ecology of hummingbirds in the Serra do Mar, southeastern Brazil. <i>Hornero</i> , <b>012</b> , 286–296.
46	-17.51	-63.64	Primary forest, Bolivia	Abrahamczyk, S. & Kessler, M. (2010) Hummingbird diversity, food niche characters, and assemblage composition along a latitudinal precipitation gradient in the Bolivian lowlands. <i>Journal of Ornithology</i> , <b>151</b> , 615–625.
47	-16.96	-65.41	Primary forest, Bolivia	Abrahamczyk, S. & Kessler, M. (2010) Hummingbird diversity, food niche characters, and assemblage composition along a latitudinal precipitation gradient in the Bolivian lowlands. <i>Journal of Ornithology</i> , <b>151</b> , 615–625.
48	-19.95	-43.90	Campo rupestre, W Brazil	Vasconcelos, M.F. & Lombardi, J.A. (1999) Padrão sazonal na ocorrência de deis espécie de beija-flores (Apodiformes: Trochilidae) em uma localidade de campo rupestre na Serra do Curral, Minas Gerais. <i>Ararajuba</i> , <b>7</b> , 71–79.

	ID number	Latitude	Longitude	Site description and general location	Data Source Reference
1	49	-23.28	-45.05	Atlantic forest, SE Brazil	Vizentin-Bugoni, J., Maruyama, P.K. & Sazima, M. (2014) Processes entangling interactions in communities: forbidden links are more important than abundance in a hummingbird–plant network. <i>Proceedings of the Royal Society of London B</i> , <b>281</b> , 1–8.
2	50	-13.22	-72.12	Subtropical humid montane forest, Perú	Watts, S. (2008) Plant-flower visitor interaction in the Sacred Valley of Perú. <i>PhD. Thesis</i> . University of Northampton, UK.
3	51	-12.85	-69.37	Lowland primary forest, Perú	Watts, S. (2008) Plant-flower visitor interaction in the Sacred Valley of Perú. <i>PhD. Thesis</i> . University of Northampton, UK.
4	52	-0.02	-78.77	Andean rainforest, mid-elevation, Ecuador.	Walther, B.A. & Brieschke, H. (2001) Hummingbird-flower relationships in a mid-elevation rainforest near Mindo, northwestern Ecuador. <i>International Journal of Ornithology</i> , <b>4</b> , 115–135.
5	53	9.57	-83.73	Elfin forest, Costa Rica	Wolf, L.L., Stiles, F.G. & Hainsworth, F.R. (1976) Ecological organization of a tropical, highland hummingbird community. <i>The Journal of Animal Ecology</i> , <b>45</b> , 349–379.
6	54	9.48	-83.48	Undisturbed highland páramo, Costa Rica	Wolf, L.L., Stiles, F.G. & Hainsworth, F.R. (1976) Ecological organization of a tropical, highland hummingbird community. <i>The Journal of Animal Ecology</i> , <b>45</b> , 349–379.

Review

**Table S2.** Correlations between corrected and observed indices of network structure ( $H_2'$ , complementary specialization; and QuanBiMo, quantitative bipartite modularity) and the predictor variables included in the OLS models: species richness/network size, phylogenetic signal in the interaction pattern of plants ( $d_p$ ) and of hummingbirds ( $d_h$ ), and all included environmental variables (except insularity as it is a dummy variable) calculated within a 10km buffer zone around each locality. MAT, Mean annual temperature; TS, Temperature seasonality; MAP, mean annual precipitation; PS, precipitation seasonality; MAT vel, velocity of temperature change; MAP vel, velocity of precipitation change; MAT anom, temperature anomaly; MAP anom, precipitation anomaly; SE, Sampling effort/intensity. Correlations for the global dataset are below the diagonal ( $N = 54$ ), and for mainland networks above the diagonal ( $N = 45$ ). \*\* $P < 0.01$ ; \* $P < 0.05$  when significance level is based on degrees of freedom corrected for spatial auto-correlation using Dutilleul's (1993) method; † significant when using traditional non-spatial statistics, but non-significant when correcting for spatial auto-correlation; ns non-significant. Significant correlations are marked in bold.

	$H_2'$ observed	$H_2'$ corrected	QuanBiMo observed	QuanBiMo corrected	Species richness	$d_p$	$d_h$	MAT	TS	MAP	PS	MAT vel	MAP vel	MAT anom	MAP anom	SE
$H_2'$ <i>observed</i>	-	+0.77**	+0.77**	+0.65**	+0.33*	-0.17 <sup>ns</sup>	-0.43**	+0.42**	-0.29 <sup>ns</sup>	+0.56**	-0.34*	-0.01 <sup>ns</sup>	+0.09 <sup>ns</sup>	-0.47**	+0.31†	-0.28 <sup>ns</sup>
$H_2'$ <i>corrected</i>	+0.82**	-	+0.51**	+0.78**	+0.45*	-0.14 <sup>ns</sup>	-0.37*	+0.23 <sup>ns</sup>	-0.17 <sup>ns</sup>	+0.34*	-0.36†	-0.14 <sup>ns</sup>	+0.19 <sup>ns</sup>	-0.40*	-0.15 <sup>ns</sup>	+0.27 <sup>N S</sup>
QuanBiMo <i>observed</i>	+0.78**	+0.55**	-	+0.76**	+0.38**	+0.32*	-0.30*	+0.34*	-0.43†	+0.61**	-0.27 <sup>ns</sup>	-0.09 <sup>ns</sup>	-0.11 <sup>ns</sup>	-0.43**	+0.40†	-0.41*
QuanBiMo <i>corrected</i>	+0.70**	+0.78**	+0.81**	-	+0.57**	-0.26 <sup>ns</sup>	-0.30*	+0.26 <sup>ns</sup>	-0.27 <sup>ns</sup>	+0.42**	-0.30†	-0.10 <sup>ns</sup>	+0.03 <sup>ns</sup>	-0.38†	+0.25 <sup>ns</sup>	+0.14 <sup>N S</sup>
Species richness	+0.44**	+0.50**	+0.52**	+0.66**	-	-0.23 <sup>ns</sup>	+0.05 <sup>ns</sup>	+0.21 <sup>ns</sup>	-0.41*	+0.31*	+0.01 <sup>ns</sup>	-0.06 <sup>ns</sup>	+0.22 <sup>ns</sup>	-0.34†	+0.13 <sup>ns</sup>	+0.14 <sup>N S</sup>
$d_p$	-0.26 <sup>ns</sup>	-0.25 <sup>ns</sup>	-0.34*	-0.33*	-0.26 <sup>ns</sup>	-	-0.19 <sup>ns</sup>	-0.11 <sup>ns</sup>	+0.21 <sup>ns</sup>	-0.14 <sup>ns</sup>	-0.07 <sup>ns</sup>	+0.11 <sup>ns</sup>	-0.05 <sup>ns</sup>	+0.05 <sup>ns</sup>	-0.07 <sup>ns</sup>	+0.07 <sup>N S</sup>
$d_h$	-0.36**	-0.32*	-0.21 <sup>ns</sup>	-0.24 <sup>ns</sup>	+0.09 <sup>ns</sup>	-0.14 <sup>ns</sup>	-0.21 <sup>ns</sup>	-0.28 <sup>ns</sup>	-0.05 <sup>ns</sup>	-0.29*	+0.36*	-0.26 <sup>ns</sup>	-0.12 <sup>ns</sup>	+0.12 <sup>ns</sup>	+0.03 <sup>ns</sup>	-0.03 <sup>ns</sup>
MAT	+0.15 <sup>ns</sup>	+0.05 <sup>ns</sup>	+0.06 <sup>ns</sup>	+0.05 <sup>ns</sup>	-0.01 <sup>ns</sup>	+0.01 <sup>ns</sup>	-	-0.21 <sup>ns</sup>	-0.09 <sup>ns</sup>	+0.25 <sup>ns</sup>	+0.02 <sup>ns</sup>	+0.66**	+0.32†	-0.10 <sup>ns</sup>	-0.16 <sup>ns</sup>	+0.00 <sup>N S</sup>
TS	-0.20 <sup>ns</sup>	-0.13 <sup>ns</sup>	-0.30†	-0.20 <sup>ns</sup>	-0.30*	+0.21 <sup>ns</sup>	-0.06 <sup>ns</sup>	-0.17 <sup>ns</sup>	-	-0.47†	+0.14 <sup>ns</sup>	+0.18 <sup>ns</sup>	+0.02 <sup>ns</sup>	+0.17 <sup>ns</sup>	-0.35†	+0.23 <sup>N S</sup>
MAP	+0.51**	+0.34**	+0.50**	+0.36**	+0.25 <sup>ns</sup>	-0.15 <sup>ns</sup>	-0.20 <sup>ns</sup>	+0.25 <sup>ns</sup>	-0.48†	-	-0.47*	-0.14 <sup>ns</sup>	-0.20 <sup>ns</sup>	-0.53**	+0.71**	-0.47*
PS	-0.25 <sup>ns</sup>	-0.28†	-0.15 <sup>ns</sup>	-0.19 <sup>ns</sup>	+0.08 <sup>ns</sup>	-0.03 <sup>ns</sup>	+0.29*	-0.08 <sup>ns</sup>	+0.20 <sup>ns</sup>	-0.50*	-	+0.06 <sup>ns</sup>	-0.03 <sup>ns</sup>	+0.21 <sup>ns</sup>	-0.20 <sup>ns</sup>	+0.12 <sup>N S</sup>
MAT vel	-0.03 <sup>ns</sup>	-0.13 <sup>ns</sup>	-0.05 <sup>ns</sup>	-0.06 <sup>ns</sup>	+0.01 <sup>ns</sup>	+0.19 <sup>ns</sup>	-0.22 <sup>ns</sup>	+0.53**	+0.22 <sup>ns</sup>	-0.20 <sup>ns</sup>	+0.13 <sup>ns</sup>	-	+0.58**	+0.42*	-0.37*	+0.21 <sup>N S</sup>
MAP vel	+0.01 <sup>ns</sup>	+0.07 <sup>ns</sup>	-0.13 <sup>ns</sup>	-0.01 <sup>ns</sup>	+0.17 <sup>ns</sup>	+0.08 <sup>ns</sup>	-0.09 <sup>ns</sup>	+0.29*	+0.05 <sup>ns</sup>	-0.24 <sup>ns</sup>	+0.03 <sup>ns</sup>	+0.61**	-	+0.40**	-0.41*	-0.39*
MAT anom	-0.23 <sup>ns</sup>	-0.13 <sup>ns</sup>	-0.24 <sup>ns</sup>	-0.17 <sup>ns</sup>	-0.16 <sup>ns</sup>	-0.16 <sup>ns</sup>	+0.07 <sup>ns</sup>	-0.14 <sup>ns</sup>	-0.46†	-0.43**	+0.18 <sup>ns</sup>	+0.33*	+0.30†	-	-0.49*	+0.32*
MAP anom	+0.14 <sup>ns</sup>	+0.04 <sup>ns</sup>	-0.15 <sup>ns</sup>	+0.08 <sup>ns</sup>	-0.04 <sup>ns</sup>	-0.03 <sup>ns</sup>	+0.05 <sup>ns</sup>	+0.04 <sup>ns</sup>	-0.37†	+0.67*	+0.27†	+0.38**	+0.38**	-0.44*	-	-0.37†
SE	-0.18 <sup>ns</sup>	+0.29*	-0.33*	+0.16 <sup>ns</sup>	+0.11 <sup>ns</sup>	-0.07 <sup>ns</sup>	-0.07 <sup>ns</sup>	+0.04 <sup>ns</sup>	-0.29†	-0.38*	+0.08 <sup>ns</sup>	+0.15 <sup>ns</sup>	+0.32*	+0.36**	-0.29†	-

**REFERENCES** Dutilleul, P. (1993) Modifying the t test for assessing the correlation between two spatial processes. *Biometrics*, **49**, 305–314.

1           **Table S3.** Coefficients of determination ( $R^2$ ) from partial regression local models separating the effect of species  
 2 richness, [phylogenetic signal](#) and environmental factors as predictors of corrected network structure when including  
 3 climate change velocity as an estimate of historical climate stability. We estimated  $R^2$  values in “full” models, i.e. in  
 4 models including all ten predictor variables (see Table 1). “Total model” reflects the total amount of variation in network  
 5 structure explained by the full model; “Total species richness”, “[Total phylogenetic signals](#)” and “Total environmental  
 6 factors” reflect the variation explained by species richness, phylogenetic signal and environmental factors, respectively;  
 7 “Only species richness”, “[Only phylogenetic signal](#)”, “Only environmental factors” reflect the amount of variation in  
 8 network structure explained by the specified predictor-type and not by any other (i.e. the unique variation); Finally,  
 9 “shared” reflects the variation attributable to the pair of specified predictor types or to all factors combined. See Table 1  
 10 for standardized coefficients of each variable and more details of model fit, [and Figure 3 for a graphical representation of](#)  
 11 [these results](#).

	Global dataset (54 networks)	Only Mainland (45 networks)	
	$H_2'$	QuanBiMo	$H_2'$
Total model	0.54	0.62	0.58
Total species richness	0.25	0.44	0.20
Total <a href="#">phylogenetic signal</a>	0.19	0.19	0.18
Total environmental factors	0.38	0.35	0.45
Unique species richness	0.13	0.24	0.07
Unique <a href="#">phylogenetic signal</a>	0.06	0.05	0.08
Unique environmental factors	0.13	0.06	0.22
Shared species richness – phylogeny	0.00	0.00	0.00
Shared species richness – environmental factors	0.10	0.14	0.11
Shared phylogeny – environmental factors	0.10	0.08	0.09
Shared all factors	0.05	0.08	0.03

**Figure S1.** Scatter plots showing the linear associations between corrected complementary specialization,  $H_2'$ , and corrected quantitative bipartite modularity, *QuanBiMo*, and species richness and hummingbird phylogenetic signal, the predictor variables exhibiting the most consistent relationships. Complementary specialization measures the exclusiveness of the interactions in the community, whereas modularity quantifies whether species interact more frequently with subsets of available species within a community. Full and open circles denote mainland and island networks, respectively. Notice the differences in the regression lines fitted to the Global and Only Mainland datasets, in thick and dashed lines, respectively.

