Studies

Evaluating the structure of commensalistic epiphyte–phorophyte networks: a comparative perspective of biotic interactions

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Abstract. Epiphytic vascular plants comprise an essential part of the tropical flora and are a key component for ecosystem functioning. Some recent studies have used a network approach to investigate the interaction of epiphytes with host phorophytes at the community level. However, knowledge on commensalistic epiphyte–phorophyte network structure still lags behind with regard to other biotic interaction networks. Our goal was to provide a more complete overall perspective on commensalistic epiphyte–phorophyte interaction and its placement with respect to other better studied mutualistic interactions. We hypothesized that the intensity of the fitness effect of the different types of biotic interactions would determine the degree of specialization of the interacting organisms. Thus, commensalistic epiphyte–phorophyte interactions would have lower specialization than mutualistic interactions. We compiled and analysed the structural properties (nestedness, network specialization and modularity) of 12 commensalistic epiphyte–phorophyte networks and compared them with the same metrics to 11 ant–myrmecophyte, 86 pollination and 13 seed dispersal mutualistic networks. Epiphyte–phorophyte networks were nested and modular with regard to the corresponding null models and had greater nestedness than mutualistic networks, whereas specialization and modularity were significantly lower. Commensalistic epiphyte–phorophyte networks of interactions are both nested and modular, and hence, are structured in a similar way to most other types of networks that involve co-evolutionary interactions. Nevertheless, the nature and intensity of the ecological processes involved in the generation of these patterns is likely to differ. The lower values of modularity in commensalistic epiphyte–phorophyte networks are probably due to the low levels of specialization and the lack of co-evolutionary processes between the interacting partners.

Keywords: Ecological interactions; modularity; nestedness; orchids; specialization; tropics.

Introduction

The establishment of biotic interactions is a topic of great interest in ecology. For a long time, ecologists have searched for patterns and processes related to species distribution and interspecific interactions in communities (Janzen 1974; Thompson 1999; Wisz et al.)

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These interactions can be described as networks in which each species is connected to one or more different species (Bascompte et al. 2003; Bascompte and Jordano 2014). The network structure has relevant implications for the coexistence and stability of species and has been described as the architecture of biodiversity (Bascompte and Jordano 2007). The topological properties of these networks provide information on community organization and offer a general framework to evaluate the different types of interactions between species (Proulx et al. 2005; Blick and Burns 2009; Bascompte and Jordano 2014). Furthermore, network analysis has been widely used to measure ecological specialization (Blüthgen et al. 2007; Schleuning et al. 2012; Lara-Romero et al. 2016; Carstensen et al. 2018; Dugger et al. 2019; but see Mello et al. 2015). Several types of mutualistic (Bascompte et al. 2003; Bascompte and Jordano 2007; Bascompte 2009; Vázquez et al. 2009; Fortuna et al. 2010; Pastor and Garca 2015) and antagonistic networks (Cagnolo et al. 2011; Hagen et al. 2012; Elias et al. 2013; Morris et al. 2014) have been extensively studied contributing to an understanding of the factors influencing the structure of these networks.

Two widely accepted topological properties of these networks, with theoretical and practical implications, are nestedness (Bascompte et al. 2003; Thébault and Fontaine 2010) and modularity (Olesen et al. 2007; Thébault and Fontaine 2010). The first implies that specialist species interact with a proper subset of the partners that interact with generalist species (Bascompte et al. 2003), whereas the second denotes a pattern in which some subsets (modules) of species are more linked to each other than to species in other modules (Fortuna et al. 2010). Nestedness and modularity are linked to community stability and robustness against perturbations and environmental changes (Thébault and Fontaine 2010; Dalsgaard et al. 2013; Robinson et al. 2015; Nuwagaba et al. 2017). However, how these properties emerge and what is their relevance to network stability, is still under debate, especially for nestedness (Burgos et al. 2007; Minoarivelo and Hui 2016; Landi et al. 2018). Although nestedness and modularity have traditionally been considered antagonistic properties of networks (Fortuna et al. 2010), some studies emerged in the past years have shown that they are not mutually exclusive, and that they can interact with one another (Flores et al. 2013; Pinheiro et al. 2016).

Interactions between vascular epiphytes (henceforth referred to as epiphytes) and host trees (phorophytes) are considered to be commensalistic because epiphytes establish on the host tree for support without causing harm or benefit (Zotz 2016). Epiphytic vascular plants comprise an essential part of the tropical and subtropical flora (Kreft et al. 2004; Krömer et al. 2005) and are a key component for ecosystem functioning and diversity (Kreft et al. 2004). Despite their importance, our understanding of the mechanisms structuring epiphyte communities is still rather poor (Burns and Zotz 2010; Wagner et al. 2015). This gap is in part due to the lack of a rigorous theoretical and empirical framework that could guide investigations of epiphyte assemblages (Burns and Zotz 2010; Wagner et al. 2015). Yet in recent years, novel studies have used a network approach to investigate these interactions at the community level. They have found that the composition of epiphyte–phorophyte interactions appears to be structured deterministically, as happens with mutualistic and antagonistic interactions. Specifically, they have found that these networks are highly nested (Blick and Burns 2009; Silva et al. 2010; Sáyago et al. 2013; Ceballos et al. 2016; but see Burns and Zotz 2010), even more than in mutualistic and antagonistic networks (Piazzon et al. 2011), and that there is no phylogenetic signal in species interaction patterns (Silva et al. 2010). These studies also support that the host specificity of epiphytes is small and that most interactions occur among generalist epiphytes and generalist phorophytes (Laube and Zotz 2006; Silva et al. 2010; Sáyago et al. 2013). However, varying traits in host phorophytes, such as bark texture and size and temporal and spatial distribution of species, can have significant effects on the performance of epiphyte species (Hietz 1999; Zotz et al. 1999; Callaway et al. 2002; Laube and Zotz 2006), and therefore, they have been invoked to determine the degree of epiphyte host specificity (Wagner et al. 2015). This is particularly important because there is a long-standing and ongoing discussion on whether vascular epiphytes show host specificity (Cornelissen and Steege 1989; Benzing 1990; Laube and Zotz 2006; Vergaro-Torres et al. 2010, reviewed in Wagner et al. 2015).

Despite progress made in recent years, the knowledge on the structure of commensalistic epiphyte–phorophyte networks lags behind those of other biotic interaction networks (Sáyago et al. 2013) and thorough comparisons between the former and latter are missing. This is in part because network studies in these commensalistic interactions are still scarce and scattered and both data source and the methodology differ across studies, making it difficult to detect general patterns (but see Piazzon et al. 2011). Furthermore, modularity has not been yet characterized for this type of interaction, and therefore, we do not know the specific importance of this basic network build-up mechanism. To provide a more complete overall perspective on epiphyte–phorophyte networks and their placement with respect to the networks of other more studied mutualistic interactions,
we analysed 12 epiphyte–phorophyte networks, and compared them with the same metrics to ant–myrmecophyte (plant–ant), pollination and seed dispersal mutualistic networks. Our general goal is to understand how epiphyte–phorophyte networks are structured. Mutualistic networks were selected as a baseline for comparison because their overall degree of specialization covers a broad range (Blüthgen et al. 2007). Pollination mutualisms tend to be more specialized than seed dispersal mutualisms, probably due to evolutionary considerations (Wheelwright and Orians 1982; Blüthgen et al. 2007), whereas obligate-myrmecophytic symbioses represent one of the most specialized mutualistic interactions (Blüthgen et al. 2007). Characterizing ecological specialization is challenging, because it entails different definitions and can be assessed at different scales (Blüthgen 2010; Devictor et al. 2010; Poisot et al. 2011). Here we considered specialization as a niche-breath correlate, which can be estimated for species interacting at the community level by well-established network metrics (Blüthgen et al. 2006; Dormann 2011; Poisot et al. 2012). Considering that the different types of biotic interactions have distinctive effects on the fitness of the interacting organisms, we hypothesized that the magnitude of fitness change generated by the different types of biotic interactions would determine the degree of specialization of the interacting organisms. Thus, organisms involved in biotic interactions which greatly affect their fitness would have greater specialization. Since epiphytes are assumed to establish on their hosts without impairing their fitness (Zotz 2016), we predicted that the epiphyte–phorophyte interaction would entail low degree of specialization. Moreover, considering that specialization and host specificity is negatively associated with nestedness (Bascompte et al. 2003) and positively associated with modularity (Albrecht et al. 2014; Dormann and Strauss 2014), we also predicted that epiphyte–phorophyte networks would be more generalist and nested, but less modular than the mutualistic ones. To the best of our knowledge, this is the first study to compare the modularity of epiphyte–host tree networks with that of mutualistic networks. Our study greatly expands the scope of Piazzon et al. (2011) who carried out a comparative study about nestedness that only included epiphyte–phorophyte networks of the same study region.

Materials and Methods

Data set
To compare epiphyte–phorophyte networks with mutualistic networks, we compiled a data set of 122 interaction networks [see Supporting Information—Table S1]. We restricted data to tropical and subtropical areas to minimize potential biases in topological metrics due to heterogeneous environmental conditions. Twelve published data sets that measured quantitatively interactions between vascular epiphyte and host species in tropical and subtropical areas were included in the study (Table 1). Epiphyte species vary across studies, but they are distributed mainly among the families Orchidaceae and Bromeliaceae [see Supporting Information—Table S1]. The data set also included 86 pollination, 13 seed dispersal and 11 ant–myrmecophyte (plant–ant) networks from published studies considering only studies carried out in tropical and subtropical areas, available in Web of life (http://www.web-of-life.es) and additional works [see Supporting Information—Table S1]. The main characteristics of the analysed networks are described in Supporting Information—Table S1; detailed description of data sets and field sampling procedures can be found in the attached references.

Data analysis
For each of the 122 networks, data were completely reanalysed to compute the following network metrics: (i) nestedness, (ii) complementary specialization and (iii) modularity. Nestedness quantifies the degree to which species with few interactions are connected to highly connected species (Bascompte et al. 2003). Nestedness was evaluated with the nestedness metric based on overlap and decreasing fill (NODF) index (Almeida-Neto et al. 2008). As the NODF metric is dependent on network size and sampling intensity (Ulrich et al. 2009), the significance of NODF was evaluated against a fixed-fixed null distribution derived from 1000 random networks with the same number of nodes and interactions as the observed networks. To characterize, network specialization at the community level, we used the index of complementary specialization $H^2$, which quantifies the degree of niche divergence of elements within an entire bipartite network, that is, whether species in a network tend to partition or share their interaction partners (Blüthgen et al. 2006). It ranges from 0 (low specialization, high niche overlap) to 1 (high specialization, low niche overlap). Modularity detects the degree to which the network is structured as cohesive subgroups of species (modules) in which the density of interactions is higher within subgroups than among subgroups (Dormann and Strauss 2014). Modularity was estimated using the QuaBiMo algorithm (Q), which is based on a hierarchical random graph approach, adapted for quantitative bipartite networks (Dormann and Strauss 2014). As the algorithm is a stochastic process, results may vary among computations. For each network, we therefore
ran the QuaBiMo algorithm 10 times and retained the optimal modular configuration, i.e. the iteration with highest Q value. Because Q also can be affected by network size (Dormann and Strauss 2014), the significance of Q was assessed against a null distribution derived from 100 random networks generated as for NODF. The value of modularity ranges between 0 (random network with no modules) and 1 (maximum modularity). Modularity increases with increasing link density within modules and decreasing connectedness between different modules (i.e. as specialization increases) (Albrecht et al. 2014; Dormann and Strauss 2014). Thus, we expected considerable lower Q values in epiphyte–phorophyte networks compared to mutualistic networks due to higher generalization. NODF calculations were based on the binary data matrix and $H^2$ and Q were all computed based on interaction frequencies (quantitative matrix). NODF calculations were based on the binary data matrix because weighted metrics explicitly assume that nested patterns are determined by the abundance of interacting species (Almeida-Neto and Ulrich 2011), and thereby it is not in accordance with the classical concept of nestedness, more focused on assessing the role of species as specialist and generalist. In any case, we also estimated weighted NODF, finding that NODF and wNODF were highly correlated (Pearson’s correlation $r = 0.86$, $P < 0.0001$, $n = 118$) and yielded similar results in comparisons across types of biotic interactions [see Supporting Information—Fig. S1].

To characterize specialization at the species level, we quantified the roles of species within networks with two species-level metrics: complementary specialization ($d'$) and between-module connector values ($c$-values). The two metrics define a specialist as an organism having stronger link strengths in association with a limited subset of its possible resources (Poisot et al. 2012). The metric $d'$ measures the level of specialization of each species based on its discrimination from a random selection of partners (Blüthgen et al. 2006). It is analogous to the calculation of $H^2$ at the community level and ranges from 0 (no specialization, species that interact with their partners proportionally to their availability) to 1 (perfect specialists, species that disproportionately interact with rare partners) (Blüthgen et al. 2006). $C$-value determines the importance of a species in connecting different modules by interactions with species from other modules, thereby reducing modularity (e.g. Schleuning et al. 2014; Dugger et al. 2019). It has a 0 value if all interactions of a species are within its own module and it is close to 1 if the interactions of a species are evenly distributed among modules. Network analyses were conducted in R v. 3.1.2 (R Development Core Team 2011) with the add-on libraries bipartite v. 2.04 (Dormann et al. 2008) and vegan v. 2.4-5 package (Oksanen et al. 2007). The network structure was explored and visualized with the R package igraph (Csardi and Nepusz 2006) using the Fruchterman–Reingold force-directed layout algorithm (Fruchterman and Reingold 1991), which is an

| Reference          | Locality           | Habitat | Field sampling | Lat     | Long     | Network size | NODF  | $H^2$ | Q     |
|--------------------|--------------------|---------|----------------|---------|----------|--------------|-------|-------|-------|
| Ceballos et al. (2016) | Tucumán, Argentina | TMF     | GR             | −26.76  | −65.33   | 44           | 69.53 | 0.11  | 0.18  |
| Dejean et al. (1995)  | Quintana Roo, Mexico | LIF     | na             | 19.38   | −87.79   | 20           | 67.77 | 0.24  | 0.22  |
| Laube and Zotz (2006) | San Lorenzo, Panamá | TDF     | CA, GR         | 8.3     | −82.1    | 107          | 20.81 | 0.28  | 0.28  |
| Martínez-Meléndez et al. (2008) | Cerro Quetzal, Mexico | ECF     | CA             | 15.72   | −92.92   | 25           | 59.89 | 0.10  | 0.13  |
| C. Naranjo, unpubl. data | Zamora, Ecuador     | TMF     | CA             | −3.99   | −76.1    | 146          | 56.98 | 0.21  | 0.17  |
| Piazzon et al. (2011)  | Caúlín Forest, Chile | TRF     | GR             | −41.83  | −73.6    | 17           | 64.29 | 0.09  | 0.11  |
| Piazzon et al. (2011)  | Senda Darwin, Chile | TRF     | GR             | −41.88  | −73.67   | 16           | 74.65 | 0.09  | 0.09  |
| Piazzon et al. (2011)  | Llanquihue, Chile   | TRF     | GR             | −41.85  | −73.57   | 22           | 74.54 | 0.06  | 0.08  |
| Piazzon et al. (2011)  | Quilicr, Chile      | TRF     | GR             | −41.92  | −73.60   | 19           | 68.95 | 0.12  | 0.12  |
| Sáyago et al. (2013)   | Jalisco, Mexico     | TDF     | GR             | 20.66   | −103.5   | 62           | 62.89 | 0.23^  | na    |
| Vergara-Torres et al. (2010) | San Andrés Cal, Mexico | TDF | CA, GR | 18.95 | −99.08 | 16 | 76.67 | 0.1 | 0.09 |
| Zhao et al. (2015)     | Xishuangbanna, China | TMF     | GR             | 22.01   | 100.8    | 180          | 16.4a | 0.5^  | na    |

Table 1. Network properties of 12 epiphyte–phorophyte networks analysed in this study. TMF, tropical montane forest; TDF, tropical dry forest; LIF, low inundated forest; TRF, temperate rainforest; ECF, evergreen cloud forest; NODF, nestedness index; $H^2$, specialization index; Q, modularity; GR, ground-based survey; CA, canopy-based survey. Network size is the sum of epiphytes and phorophytes. *Denotes that data were obtained from the original source. na denotes that data were not available from the original source and bipartite matrix was not published and therefore we could not estimate the metric. Values that are statistically significant from random expectations ($Z$-test: $P < 0.05$) are indicated in bold.
energy-minimization plotting algorithm that optimizes
the placement of the species in the graph.

To contrast network- and species-level metrics among epiphyte–phorophyte, pollination, seed dispersal and ant–myrmecophyte networks, we applied linear mixed models (LMMs). All models included metrics as the response variables, type of interaction as a fixed factor and locality as random factor to control for potential spatial non-independence (Zuur et al. 2009). Network size was included as a continuous predictor variable in models fitted for network-level metrics, given that network metrics can depend on network size or sampling intensity (Blüthgen et al. 2008; Ulrich et al. 2009; Dormann and Strauss 2014). We assumed Gaussian error for all LMMs, and NODF was In-transformed to reach normality and homoscedasticity. Model residuals were checked graphically for normality and homogeneity of variances using diagnostic plots (Zuur et al. 2009). Using the Anova function in the R package ‘car’, we tested the null hypothesis that the response means are identical across type of interaction. When the overall analyses of variance indicated a significant difference (P < 0.05), a Tukey’s test for post hoc multiple contrasts was conducted using glht in the package multcomp v. 1.4-6 (Hothorn et al. 2008). Models were assessed for goodness-of-fit to the data using the marginal (Rm²) and conditional (Rc²) R² described by Nakagawa and Schielzeth (2013). Rm² represents the variance explained by fixed factors while Rc² is interpreted as variance explained by both fixed and random factors (i.e. the entire model). Linear mixed models were fitted with the R package ‘lme4’ (Bates et al. 2015).

Results

All three network metrics varied significantly among the different types of interactions (all LMM tests; P < 0.0001; Figs 1 and 2; see Supporting Information—Table S2). Specialization and modularity were significantly lower in the epiphyte–phorophyte networks than in the rest of the networks (Figs 1 and 2, all tests: P < 0.016), whereas nestedness was significantly higher (Fig. 2, P < 0.02). Network size influenced NODF values (LMM tests: P < 0.01; see Supporting Information—Table S2), but did not significantly affect H 2 and Q (LMM tests: all P > 0.13; see Supporting Information—Table S2). Rc², which represents variance explained by the entire model, ranged between 0.59 and 0.62 [see Supporting Information—Table S2], with the variance explained by fixed factors (Rm²) representing a large fraction in all cases (41–67 %). Regarding species-level metrics, d’ and c-values varied significantly among types of interactions (Fig. 3, d’: χ² = 74.93, df = 3, P < 0.0001, Rm² = 0.19, Rm² = 0.1; c-value: χ² = 100.5, df = 3, P < 0.0001, Rc² = 0.26, Rm² = 0.13). C-value was significantly higher in the epiphyte–phorophyte networks than in the rest of the networks (Fig. 3, all tests: P < 0.0001), whereas d’ was significantly lower (Fig. 3, all tests P < 0.0001).

Focusing on epiphyte–phorophyte networks, the three network metrics varied considerably (NODF = 44.44 ± 18.05, H/2 = 0.18 ± 0.13, Q = 0.15 ± 0.06; Table 1). NODF values were significantly higher than those from the null models in 9 of 12 analysed networks, while Q was significantly higher than random expectations in 7 of the 10 analysed networks (Table 1).

Discussion

In this paper, we evaluated the network structure of 12 epiphyte–phorophyte networks and compared it to those found in ant–myrmecophyte (plant–ant), pollination and seed dispersal mutualistic networks. Results confirmed our hypothesis that epiphyte–phorophyte networks are more nested and connected and have lower modularity and specialization than mutualistic networks.

Most studies agree that nestedness observed in epiphyte–phorophyte networks is largely explained by differences in species abundance, i.e. species interact randomly resulting in interaction frequencies that are proportional to relative species abundances (Burns 2007; Silva et al. 2010; Piazzon et al. 2011; Sáyago et al. 2013; Ceballos et al. 2016). This is further supported by the low values of the index of network specialization (H/2) found in our study, which suggests that colonization of phorophytes by epiphytes is proportional to the abundance of phorophytes (Blüthgen et al. 2007). However, 9 out of 12 networks analysed were more nested than null model expectations, which implies that nestedness is not entirely explained by abundance. This is in line with previous case studies where reported nested patterns were attributed to ecological processes (Burns 2007; Silva et al. 2010; Piazzon et al. 2011; Sáyago et al. 2013; Ceballos et al. 2016). In any case, current observational data are rather controversial, and there is still no general agreement about the key processes shaping nestedness in epiphyte–phorophyte interactions, extending the debate observed in other types of interaction networks (Burgos et al. 2007; Minoarïvelo and Hui 2016; Pinheiro et al. 2016).

Modularity (Q) differed from null expectations, which indicated a certain level of ecological compartmentalization. Epiphyte networks therefore appear to be structured similarly to most other types of species interactions as they showed both significant nestedness and
Modularity (Bascompte et al. 2003; Olesen et al. 2007; Thébault and Fontaine 2010). Modularity implies that some subsets (modules) of species are more linked to each other than to species in other modules. This is the first time that modularity is reported for epiphyte-phorophyte networks, and so far, we lack information about the mechanisms that underpin modularity. In mutualistic interactions, modules have been usually viewed as potential co-evolutionary units of biological significance that arise due to trophic specialization, divergent selection regimes and phylogenetic clustering of closely related species (Olesen et al. 2007; Martín-González et al. 2012; Watts et al. 2016; Morente-López et al. 2018). These mechanisms are not plausible in
epiphyte–phorophyte networks because interactions among species are not mutually advantageous and therefore, natural selection does not favour the convergence and complementarity of traits in interacting species. This is further supported by the lack of evidence of phylogeny as an explanatory factor structuring epiphyte–phorophyte networks (Silva et al. 2010; Sáyago et al. 2013). However, although there is not strong specialization, varying traits in host phorophytes do provide contrasting microhabitat conditions that influence the interactions with epiphytes (Hietz 1999; Zotz et al. 1999; Callaway et al. 2002; Laube and Zotz 2006). Differential use of these resources can give rise to phorophyte-specific epiphyte spectra (Zotz et al. 1999). We propose that modularity may arise as a result of ‘structural host specificity’ (sensu Wagner et al. 2015), i.e. differences in the performance of the focal epiphyte species on a given host phorophyte relative to another host phorophyte. Hence, modules may contain a set of species with convergent traits, which influence epiphyte performance. These traits may be related to physical bark characteristics, leaf and bark chemistry or branch architecture of the phorophytes and microclimatic specificity, duration of life cycle, size at maturity or diaspore characteristics of the epiphytes (Wagner et al. 2015). Identification of modules may bring us closer to an understanding of the structure of complex networks of interaction (Olesen et al. 2007). Thus, a more explicit focus on modules as study objects may open pathways towards a better mechanistic understanding of the epiphyte–phorophyte interaction.

Our results provide insight on the current discussion of whether vascular epiphytes show low levels of specialization (Wagner et al. 2015). Species-level metrics indicated lower degree of specialization in epiphyte–phorophyte than in mutualistic networks, consistent with our results at the network level. Nested structure indicates that host specificity of vascular epiphytes is small and instead, specialists engaged in few interactions are connected predominantly with generalists (Silva et al. 2010; Piazzon et al. 2011; Taylor et al. 2016). This implies that strict specialization is rare. Furthermore, low values found for $H^2$ and $Q$ in epiphyte–phorophyte networks compared to mutualistic networks indicated high niche overlap among species or modules (Blüthgen et al. 2006; Schleuning et al. 2014; Dugger et al. 2019). This is not surprising because driving forces for co-evolution are expected to be weaker in epiphyte–phorophyte interaction as epiphytes do not have fitness effects on their host. Furthermore, epiphytes are structurally dependent plants that cannot actively search for appropriate

Figure 2. Variation in network metrics across network interaction types: commensalistic epiphyte–phorophyte (EP), seed dispersal (SD), pollination (PO) and ant–myrmecophyte (AM) networks. (A) Nestedness; (B) specialization; (C) modularity. Different letters denote significant differences among network types ($P$-value < 0.05) after Tukey’s correction for multiple comparisons.
phorophytes and thus only have the possibility of establishing at the location where diaspores are deposited by chance (Wagner et al. 2015). This leads to concluding that a general pattern of host generality is the rule (Wagner et al. 2015).

There are some possible sources of variation that cannot be tested in our study due to the limited number of networks available. For example, we were not able to assess differences in network structure across types of habitat within tropical regions. Studies examining vascular epiphyte communities are particularly scarce outside neotropic and most of the studies were performed in tropical dry forest (Burns and Zotz 2010; Silva et al. 2010; Sáyago et al. 2013). This is important because, for example, specialization of vascular epiphytes may differ between dry forests and rainforests (Vergara-Torres et al. 2010) altering the interaction structure. However, the robustness of our results, along with the similar outcomes found in other regions (Burns 2008), reinforces the idea that commensalistic epiphyte–phorophyte networks are deterministically structured. Another factor to consider is the effect of sampling strategies, i.e. ground-vs. canopy-based inventories, on network structure. Both sampling methods provide different effectiveness in recording species richness and frequency (Flores-Palacios and García-Franco 2001). Again, the limited number of studies available prevent us from making any formal analysis. Previous comparisons between ground-based and canopy-based inventories did not identify substantial effects on network structure (Burns 2008). Yet, sampling bias cannot be ruled-out entirely and comparisons between ground-based and canopy-based inventories should be carried out to identify potential biases that might result from the use of different methods. Finally, several of the networks analysed were comprised of only one type of epiphyte (i.e. orchids, bromeliads or angiosperm epiphytes). This may lead to a phylogenetic bias in the network metric values because different types of epiphytes can use different strategies to establish, grow and reproduce (Zotz and Schmidt 2006), which can affect the interaction structure. This may be particularly important when only certain phylogenetic groups of a richer epiphyte–phorophyte community are considered, which may entail important consequences for measurement of network structure (Lewinsohn et al. 2006; Flores et al. 2013).

Conclusions

This study showed that commensalistic epiphyte–phorophyte networks are more nested, but less modular, than the mutualistic ones. Overall, these findings confirmed that the interaction between vascular epiphytes and host phorophytes is predominantly generalist. The study also identified for first time that epiphyte–phorophyte networks of interactions are both nested and modular, and hence, are structured in a similar way to most other types of networks that involve co-evolutionary interactions. A better understanding of underpinning mechanisms that drive nestedness and modularity is required.
to gain insight on the epiphyte–phorophyte interaction. Further research is also needed to evaluate the generalization of our results across regions and systems, and to identify the factors responsible for the convergence between epiphyte–phorophyte and mutualistic networks.

**Data**

The data sets generated during and/or analysed during the current study are available in the Figshare Digital Repository: [https://doi.org/10.6084/m9.figshare.7751189.v1](https://doi.org/10.6084/m9.figshare.7751189.v1). The R-code can be downloaded from GitHub at [https://github.com/CarlosLara/R-ecology](https://github.com/CarlosLara/R-ecology).

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**Contributions by the Authors**

The work was designed by C.N., M.L.R. and C.L.-R.; C.L.-R. analysed the results with the contribution of C.N. and J.M.I.; C.L.-R., J.M.I. and C.N. wrote the manuscript with the contribution of M.L.R.

**Conflict of Interest**

None declared.

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**Supporting Information**

The following additional information is available in the online version of this article—

**Table S1.** Bibliographic reference, taxonomic focus, location and topological properties of the epiphyte–phorophyte and mutualistic networks analysed in this study.

**Table S2.** Linear mixed models performed to contrast metrics among type of interactions.

**Figure S1.** Distribution of weighted NODF values across network interaction types.

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