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A Network Perspective for Community Assembly

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Species interactions are responsible for many key mechanisms that govern the dynamics of ecological communities. Variation in the way interactions are organized among species results in different network structures, which translates into a community’s ability to resist collapse and change. To better understand the factors involved in dictating ongoing dynamics in a community at a given time, we must unravel how interactions affect the assembly process. Here, we build a novel, integrative conceptual model for understanding how ecological communities assemble that combines ecological networks and island biogeography theory, as well as the principles of niche theory. Through our conceptual model, we show how the rate of species turnover and gene flow within communities will influence the structure of ecological networks. We conduct a preliminary test of our predictions using plant-herbivore networks from differently-aged sites in the Hawaiian archipelago. Our approach will allow future modeling and empirical studies to develop a better understanding of the role of the assembly process in shaping patterns of biodiversity.

Keywords: evolution, coevolution, succession, turnover, island biogeography, network

1. INTRODUCTION

Species interactions shape the dynamics of ecological communities. As communities assemble, species accumulate—despite prevailing biotic and abiotic conditions—either through speciation or immigration (Hutchinson, 1957; Fukami, 2015; Mittelbach and Schemske, 2015). Antagonistic, mutualistic and facilitatory interactions among species can act as gate keepers to new species, and as a consequence, eco-evolutionary feedbacks between species can determine the trajectory of community assembly (Post and Palkovacs, 2009; Gravel et al., 2011; Pillai et al., 2011; De Meester and Pantel, 2014; Fahimipour and Hein, 2014; Fukami, 2015; Mittelbach and Schemske, 2015; Cadotte and Tucker, 2017). But how do ecological networks assemble?

Since the 1970s, researchers have searched for patterns in community assembly (e.g., Cody and Diamond, 1975; Drake, 1990; Morton et al., 1996; Chase, 2003; Tilman, 2004; Fukami, 2015), though species interaction networks were not explicitly considered until a decade later (Pimm, 1979; Post and Pimm, 1983). Subsequently, how interaction-based, eco-evolutionary feedbacks
shape the community emergence has become the subject of numerous modeling approaches (Louelle and Loreau, 2005; Fussmann et al., 2007; Brännström et al., 2011; Allhoff et al., 2015; de Andreazzi et al., 2018, see Brännström et al., 2012 for a review). To test whether models reflect likely mechanisms of community assembly, authors compare the networks generated by their models to the interaction patterns observed in real-world communities. However, a non-standard suite of summary statistics of the final simulated and empirical networks are compared—often overlooking the community’s trajectory to that endpoint (see, Yeakel et al., 2014). By ignoring the intrinsically dynamic nature of assembly, our understanding of the process is undermined. The likely reason investigators disregard the assembly process is 2-fold: (1) despite the plethora of network summary metrics that can be used to capture different aspects of network architecture (Newman, 2010), there are no clear guidelines on what aspects of network structure will change with assembly; and (2) few empirical studies exist for comparison because it is time-intensive and expensive to capture interaction assembly dynamics (see, Olesen et al., 2008; Albrecht et al., 2010; Ponisio et al., 2017; Tylankasis et al., 2018, reviewed in Bascompte and Stouffer, 2009), particularly over deep evolutionary timescales (Olesen et al., 2010).

New innovations in machine learning and metabarcoding—to allow identification of all members that make up a community and their interactions—will likely improve inference of long-term, higher resolution community network surveys in the near future (Kartzenel et al., 2015; Pornon et al., 2016; Freilich et al., 2018). To address the first impediment, here we propose a conceptual model to study and predict how networks change as communities assemble. Our conceptual model is anchored in niche theory including niche dynamics (niche breadth and overlap), dispersal, and eco-evolutionary feedbacks (taxon cycles, ecological release, and character displacement), from which we derive predictions about interaction network assembly. Our approach will help to guide future analytical modeling and empirical studies to develop a better understanding of the role ecological and evolutionary processes play in shaping community-level patterns.

We then focus on well-studied island systems to review how interactions change as a community assembles. Islands provide discrete, often differently-aged communities, allowing us to study how interactions change over an extended period of time (Brodie, 2017). By comparing the trajectory of community assembly on isolated islands with little species turnover or gene flow to near-shore islands with comparatively high turnover and gene flow, we are able to elucidate the importance of current and historic community characteristics in shaping interaction patterns. Habitat patches embedded within a less hospitable matrix may also function as biological islands (Gillespie and Roderick, 2002). Thus, understanding the mechanisms underlying the assembly of true island systems is applicable to a broad set of ecological questions. Lastly, we provide an approach for comparing network structure as a community assembles and apply those approaches within an existing dataset of Hawaiian plant-herbivore interactions.

2. CONCEPTUAL MODEL

2.1. Niche Theory: Limiting Similarity and the Benefits of Specialization

Theory concerned with predicting interaction occurrence often involves two principles thought to increase an individual’s fitness. The first is limiting similarity (Macarthur and Levins, 1967; Abrams, 1983), since increasing sharing of resources or other ecological factors—referred to as niche overlap—intensifies competition. Thus, one individual will either out-compete the other, or each will modify their niche space in order to minimize competition (Macarthur and Levins, 1967; Abrams, 1983). In the original description of this principle, limiting similarity was a property of populations (Gause, 1932; Hutchinson, 1959), but since then, this principle has been expanded to capture individual plasticity in resource use (Bolnick et al., 2002; Brosi, 2016). When applied to interaction networks, the interaction niche is the identity of the possible partners with which a species can interact, and interaction niche overlap is partner sharing between species of the same trophic level (Box 1). Here we use “interaction partner” to describe any type of interaction between two species, for example a mutualistic interaction between a bee and a flower, or a consumptive interaction between a caterpillar and plant.

The second principle is the benefits of specialization, often represented by the idiomatic phrase: “a jack of all trades is a master of none” (Gause, 1936; Holt, 1977; Parrish and Bazzaz, 1979; Waser et al., 1996; Chase and Leibold, 2003; Morlon et al., 2014). If resources are available and stable in the environment, species that use a narrower range of resources, or have narrower niche breadth (also referred to as niche width), have a fitness advantage because they are able to more efficiently use the available resources (i.e., specialists) (Leigh Jr, 1975; MacArthur, 1984). In contrast, specialization will be less likely to have a fitness advantage when resources are unstable (Herrera, 1988, 1996; Gómez and Zamora, 2006). When considering networks, the interaction niche breadth is the total of all possible partners with which a species can interact (Box 1).

2.2. Evolutionary Context: Evolutionary and Phenotypic Distinctness

Phylogenetic and phenotypic distinctness will influence the interactions in which species can participate (Vermeij, 1994; Jordano, 1995; Thompson, 2005; Ives and Godfray, 2006; Petchey and Gaston, 2006; Rezende et al., 2007; Donatti et al., 2011; Schlenuing et al., 2015; Coux et al., 2016; Pigot et al., 2016). Studies found that species that were distantly related to other community members were not only phenotypically distinct, but also more specialized (Coux et al., 2016; Pigot et al., 2016, but see Donatti et al., 2011; Fritschie et al., 2014). The distribution of phenotypic and/or phylogenetic distinctness in a community, thus, will affect both the average and variance in interaction niche overlap and breadth. In addition, distinct species with specialized interaction needs (either in consumer-resource, mutualistic, or facilitatory interactions) are often unsuccessful at entering a community (Holt et al., 1999; Piechnik et al., 2008; Gravel et al., 2011; Fukami, 2015). This limitation has
been considered especially important in island biogeography (Wallace, 1891; Carlquist, 1974; Armbruster and Baldwin, 1998; Nobre et al., 2010).

2.3. Ecological Context: Intra- and Inter-specific Competition

The distribution of interaction niche overlaps will determine where intra- or inter-specific competition is the most limiting to population growth. When inter-specific competition is strong, a species’ individuals may specialize on the resources, constricting the population-level interaction niche (Day and Young, 2004; Bolnick et al., 2010; Brosi and Briggs, 2013; Fründ et al., 2013). Similarly, release from inter-specific competition, known as competitive or ecological release, increases population niche breadth by increasing among-individual variation (Bolnick et al., 2010). On islands, the expansion of niche breadth in island taxa compared to continental relatives, driven by the absence of antagonistic interactions including competition, has long been noted as an island syndrome in evolutionary biology (Yeaton, 1974; Werner and Sherry, 1987).

Conversely, when competition within a species is strong, individuals may specialize on distinct interaction partners, broadening the population-level niche breadth (Bolnick, 2001; Bolnick et al., 2002; Svanbäck and Bolnick, 2007; Araújo et al., 2011; Brosi, 2016). Thus interaction niche breadth and overlap will change at the individual and population level based on the strength of inter- and intra-specific competition.

2.4. Eco-Evolutionary Feedbacks

The ecological context in which species are embedded will affect the evolutionary/phenotypic distinctness of species in a community. Traits that enable species interactions are shaped by reciprocal selection and, over evolutionary time scales, by patterns of speciation. Faced with intra-specific competition, individual specialization can lead to frequency-dependent disruptive selection (Bolnick, 2004; Bolnick and Lau, 2008). Similarly, with high inter-specific competition, character displacement can lead to interaction partner displacement (Grant and Grant, 2006), resulting in a species’ realized niche breadth eventually becoming its fundamental niche (Rosenzweig, 1978), and ultimately greater species-level specialization. Given the hypothesized benefit of specialization and limiting similarity, consistent selection between interaction partners may favor specialization and niche partitioning among species (e.g., the taxon cycle, Ricklefs and Bermingham, 2002). Such an interaction pattern would also only result if there were no other limits to evolutionary and coevolutionary adaptions and the network is a true representation of the interacting community (Box 2).
Conversely, continual turnover in either the environment and/or interaction partners will favor generalization (Waser et al., 1996; Hawkins, 2005; Thompson, 2005; Gómez and Zamora, 2006; Montoya et al., 2006; Loeuille and Leibold, 2008; Urban et al., 2008) because a species with a narrow interaction niche breadth will be restricted by its biotic requirements (Wilson, 1961; Ricklefs and Bermingham, 2002). In addition, gene flow between populations will dilute the selective pressure on specific, pairwise species interactions (Herrera, 1988, 1996; Waser et al., 1996; Thompson, 2005; Gómez and Zamora, 2006). Thus over evolutionary time scales, communities with high gene flow between populations and partner turnover will not favor interaction patterns in which species or groups of species are specialized on distinct sets of resources (Waser et al., 1996; Gómez and Zamora, 2006).

2.5. Networks and Interaction Niches

Niche patterns may be translated into different network architectures depending on the degree of, and variance in, interaction niche breadth and overlap (Cohen and Stephens, 1978; Araújo et al., 2011; Valdivinos et al., 2016). Two of the most used patterns to describe network architecture are modularity and nestedness. Modularity measures the extent to which species in a community form tightly interacting groups of species, or modules (Box 1). On the other hand, in nested networks, specialist species interact with a subset of the species that generalists interact with (Box 1). Modular networks can also have compartments that are nested (Figure 1C). Modularity emerges from increased niche partitioning and decreased niche breadth (Figures 1A, C, E), whereas nestedness emerges from increased niche overlap and a high variance in niche breadth (Figure 1D). Both measures of network architecture capture essential structural features affected by niche breadth and overlap, and are expected to be tightly linked with community stability (Thébault and Fontaine, 2010; Valdivinos et al., 2016). Networks dominated by generalist or specialist species with substantial interaction overlap will neither be nested nor modular (Figures 1B, F).

Depending on current ecological context and evolutionary history, interaction patterns will differ in their degree of niche overlap and average niche breadth, as well as the variance of both between species. As discussed above, the turnover of interaction partners and gene flow will influence an individual’s ability to limit similarity and specialize. The evolutionary and phenotypic distinctness of species entering a community will also influence the distribution of niche overlap and breadth. Lastly, the strength of inter- and intra-specific competition will influence a species’s realized niche overlap and breadth.

The community characteristics (species turnover, gene flow, intra- and inter-specific competition, evolutionary/phenotypic distinctness) that shape niche breadth and overlap, in turn, are a function of a community’s stage of assembly and isolation (MacArthur and Wilson, 1967; Levin, 1981). Just as species richness is determined by the balance between species extinction rate and the rate at which new species enter a community (either due to colonization or in situ speciation), our conceptual model predicts that: (1) the rates of species and gene flow within communities will affect the patterns observed in interaction networks, and (2) as assembly progresses, isolated systems will support greater interaction specialization and niche partitioning (Figures 2, 3). In contrast, we predict the opposite trend in proximate environments, where high species and genetic turnover will result in communities having high levels of niche overlap and limited specialization (Figures 2, 3).

2.6. A Note About Different Types of Interactions

Most of the theory on which we base our conceptual model is focused on competitive interactions. We do this because competition is widespread, has a long history of study especially as a part of adaptive radiation (Schultz, 2000), and because other types of interactions are likely to contain competitive processes within guilds or trophic levels (Ehrlich and Raven, 1964; Armbruster and Muchhala, 2009). We note, however, that other types of interactions can have different effects on selection and diversification (Yoder and Nuiser, 2010; Hembry et al., 2014), and we thus do not necessarily expect all interaction types.
FIGURE 1 | Conceptual figure illustrating the interplay of interaction niche overlap and interaction niche breadth in determining the arrangement of interactions in a community, and specifically nestedness and modularity (A–F). In the accompanying network figures, white circles represent one trophic level (e.g., pollinators) and dark squares represent the other trophic level (e.g., plants). Each color represents a species. Within our review, we could not find evidence of communities exhibiting a structure as depicted in (F).

3. CONCEPTUAL MODEL PREDICTIONS

3.1. Partner Predictability and Low Gene Flow in Isolated Systems

Island biogeography theory predicts that dispersal to isolated communities will be rare in all stages of assembly. Sporadically colonizing species, initially freed from predators, parasites, and inter-specific competition, will be confronted with an array of open niche space (Figure 3A). Species that depend on specific interaction partners are rarely early colonizers due to the low likelihood of both species simultaneously reaching a new community. Consequently, the filter is high for specialized consumer-resource interactions, mutualisms, and facilitative interactions (Figure 3A, Holt et al., 1999; Piechnik et al., 2008; Gravel et al., 2011; Fukami, 2015). Having colonized, antagonistic interactions that limit population growth and establishment in mainland systems, such as herbivory and predation, will be less important because of the low likelihood the antagonists colonize with the host/prey species (Figure 3A).

Early arrival may allow some species to adapt and monopolize resources (i.e., priority and monopolization effects, Gillespie, 2004; Fukami, 2015; Shaw and Gillespie, 2016; Vanoverbeke et al., 2016), precluding other species from establishing or persisting, a phenomenon known as a priority effect. Because priority effects are more likely in systems with resource and composition stability (Fukami, 2015), their effects can be particularly strong on isolated islands, depending on the disturbance regime. We thus expect that networks of early stage isolated communities will be composed of predominantly generalist species due to either (1) ecological release of interaction specialization subsequent to colonization; or (2) being generalist prior to colonization.

to follow these same mechanisms and assembly rules under all circumstances. Throughout, where it is useful and illustrative, we invoke examples from non-competitive antagonistic and mutualistic interactions alongside those from competition.
The degree of niche overlap will depend on the evolutionary history and traits of the species that happen to colonize. Thus the community at very early assembly stages will be composed of generalists with either little (Figure 1A) or some niche overlap (Figure 1, between A and B), resulting in a non-nested networks that are either also non-modular (Figure 1B) or somewhat modular (Figure 1, between A and B).

As colonists’ abundance increases, intra-specific competition also increases (Figure 3A, negative frequency-dependence, Chesson, 2000; HilleRisLambers et al., 2012), potentially favoring species that are able to partition resources amongst individuals by increasing individual specialization on specific interaction partners. This individual-level specialization (Roughgarden (1972)’s “within niche component”) results in a broader niche at the species level, potentially increasing overlap between species (Roughgarden (1972)’s “between niche component”) (Wilson, 1961; Roughgarden, 1972). The species interaction networks at this stage will continue to be generalized but also shift toward more niche overlap, resulting in low nestedness and modularity (Figure 1B).

As assembly continues, subsets of the population that specialize on different resources and experience distinct selection pressures may become reproductively isolated, ultimately leading to adaptive radiations via ecological speciation for some lineages (Figure 3A, Roughgarden, 1972; Rainey and Travisano, 1998; Grant, 1999; Travisano and Rainey, 2000; Ackermann and Doebeli, 2004; Day and Young, 2004). In situ specialization becomes an alternative source of new species (weakening the dispersal filter, Figure 3A, Mittelbach and Schemske, 2015). The increase in species richness will also increase and intensify inter-specific competition (Figure 3A), favoring resource partitioning and partner specialization between species (Wilson, 1961; Schluter, 2000; Ricklefs and Bermingham, 2002; Maherali and Klironomos, 2007; Knelman and Nemergut, 2014). For example, when presented with potential interactions with the Galápagos flora, Darwin’s finches evolved specialized beak morphologies to consume the seeds of different plants under variable abiotic conditions (Grant, 1999; Guerrero and Tye, 2009). Similar radiations to exploit novel interaction partners have been documented in lizards on oceanic islands and Hawaiian honeycreepers (Olesen and Valido, 2003; Lerner et al., 2011). Assuming that evolutionarily similar species interact with similar partners (Coux et al., 2016; Pigot et al., 2016), we expect species-level networks to be modular, with niche overlap of closely related species within a module and limited overlap between modules (Figure 1C). Given enough time, partner and resource stability leading to specialization via niche differentiation, networks in isolated islands would approach specialized, partitioned networks (Figure 1E).

Interestingly, some of the empirical literature on oceanic island interaction networks suggests very different predictions than those derived from our conceptual model, and provides some intriguing insights into the processes that might dictate deviations from our predictions. In particular, mutualistic species interaction networks on oceanic islands often appear to be nested and sometimes dominated by super-generalist taxa which interact with nearly all potential partners in their community (Olesen et al., 2002; Traveset et al., 2015; Nogales et al., 2016).
but see (Trøjelsgaard et al., 2013); this can even occur in cases where other factors like interaction intimacy might otherwise be expected to drive specialization (Hembry et al., 2018). It may be that smaller islands, like the Galápagos, Canary, and Society islands where these studies took place, even if they are isolated, do not provide sufficient area for as much diversification as we describe in our model (Losos and Schluter, 2000; Gillespie et al., 2017). Because relatively less diversification is able to take place, these communities are paused in an intermediate stage in this process of assembly where ecological release is the dominant determinant of interactions. Thus, we tend to observe many generalized and even super-generalized interactions in food webs (Piechnik et al., 2008) and mutualistic networks on oceanic islands (reviewed in Kaiser-Bunbury et al., 2010).

3.2. Metacommunity Dynamics and Gene Flow in Proximate Systems

The examples described so far take place in isolated locations with low rates of colonization and gene flow. At sites with many colonists, however, assembly will have a different trajectory. Metacommunity theory is commonly used to study the assembly of less isolated environments (Leibold et al., 2004), as illustrated through the well-known study of Florida’s defaunated mangrove islands (Simberloff, 1974; Piechnik et al., 2008).

Following defaunation, the mangrove islands quickly accumulated species through dispersal. This led to high interspecific competition, with species that were pre-adapted to the existing conditions prevailing (species sorting and “slotting in”, Simberloff and Wilson, 1969; Simberloff, 1974). Because the available abiotic and biotic resources on these islands were limited, the most successful species were generalized (Piechnik et al., 2008). In these proximate islands, dispersal and intra-specific competition are weak filters, while environmental and inter-specific competition are strong ones (Figure 3B). Priority and monopolization effects are less important during assembly in these communities than in more isolated islands, because poorly-adapted early colonizers have insufficient time to adapt or reach sufficient abundances before being out-competed by pre-adapted later colonists (Fukami, 2015; Vanoverbeke et al., 2016). The networks of these early-stage proximate communities will be highly generalized with substantial niche overlap, and thus neither nested nor modular (Figure 1B).

Given the principles of limiting similarity, inter-specific competition should favor niche partitioning among colonizing species, but the lack of isolation of these communities and the presence of constant gene flow makes it less likely that existing species will adapt to local conditions (Vanoverbeke et al., 2016). In addition, the high rate of species turnover—due to high

**FIGURE 4 |** Network nestedness and modularity ($z$-scores, Top panels), niche overlap and average niche breadth of herbivores (Middle panels) and plants (Bottom panels) along the Hawaiian chronosequence (from youngest to oldest: Kilauea, Hawaii; Kohala, Hawaii; Haleakala, Maui; Kokee, Kauai). Error bars for nestedness and modularity represent the 95% confidence intervals generated by the null community simulation.
colonization rates by new species and the continual extirpation of residents—leads to unstable resource availability and dilutes the selective pressure of any one species on its partners, ultimately inhibiting the evolution of specialized interactions (Thompson, 2005; Gómez and Zamora, 2006).

However, eventually it is likely that specialist species that can exploit the available resources more efficiently than generalists will colonize from the mainland (i.e., species sorting, Leibold et al., 2004; Pienknot et al., 2008). If specialist species rely on other specialist interaction partners where there is high species turnover, they will likely go extinct. Therefore, specialists that interact with generalist species would be more likely to persist. Combined high niche overlap and asymmetric specialization of intermediate to late stage proximate island communities will generate nested interaction networks (Figure 1D). Eventually, if the environment remains stable, species particularly adapted to the habitat will predominate, increasing intra-specific competition (Figure 3B). Modules, each with a nested structure, will emerge (Figure 1C).

4. METHODS FOR TESTING OUR CONCEPTUAL MODEL

Using a published dataset of plant-herbivore interaction networks along the chronosequence of the Hawaiian islands (Rominger et al., 2016), we demonstrate a network approach to study community assembly by examining changes in interaction patterns through extended time (Pellissier et al., 2017). We extracted qualitative data on interactions between native herbivorous Hemiptera species and native plants from published species accounts and host-plant association at four focal sites on three islands: Kauai (1 site; 5.1 Ma), Maui (1 site; 0.75 Ma) and Hawaii (2 sites; 0.15 and 0.003 myo), and used these data to assemble species interaction networks. Focal sites had a similar forest composition (dominated by Metrosideros polymorpha; Myrtaceae), elevation (1,100–1,400 m) and rainfall (mean annual precipitation 2,000–3,000 mm). This published dataset represents one of very few empirical studies of interaction networks across an evolutionary timescale (but see, Trojselgaard et al., 2013; Hembry et al., 2018), however, these data are very sparse (presence/absence data for interactions at only four sites on three islands). More extensive sampling to capture other stages of assembly are necessary to explore interaction assembly on these islands; however, our intention is to illustrate the possible methods for testing our framework’s predictions.

We calculated the niche overlap of plants and herbivores using Jaccard’s similarity index (Box 1) and the interaction niche breadth (average degree, Box 1) within each trophic level. Because the networks have different numbers of species and thus different maximum diet breadths, we divided by the total number of possible species with which to interact. We calculated network nestedness using the estimator NODF (Box 1, Almeidano et al., 2008). We estimate modularity using a random walk algorithm (Pons and Latapy, 2005; Gabor and Tamas, 2006). We calculated standardized z-scores so that nestedness, modularity, and niche overlap could be compared across communities (Ulrich et al., 2009; Pellissier et al., 2017). To generate z-score values, we first assembled 999 null networks by shuffling the interactions between species while keeping each species’ total number of interactions constant (the quasiswap null model in the vegan package, Ulrich et al., 2009; Oksanen et al., 2017).

Overall, the majority of our predictions aligned with network analysis results from these Hawaiian plant-herbivore communities. Plant-herbivore networks tended to be significantly modular and unnested (Fontaine et al., 2011), and modularity increased through time while nestedness decreased (Figure 4). These results corroborate our prediction that over time, evolution/coevolution and in situ speciation will result in increasing modularity and decreased nestedness. Except for the oldest site (Kokee, Kauai), niche overlap of plants and herbivores decreased with substrate age (Figure 4). As expected, niche breadth of herbivores decreased with island age (Figure 4). However, niche breadth peaked at intermediate island ages in plants (Figure 4). Given that our aim is to illustrate how one might test our conceptual model’s predictions, we will refrain from speculating further about the reasons certain islands depart from some of our predictions in this example.

5. CONCLUSION

The role of species interactions in the assembly trajectory of isolated islands, such as Hawaii, and proximate islands, such as mangrove islets, implies that communities will diverge over time, even if the starting conditions are the same, resulting in different network structures. Because patterns of selection imposed by interacting species varies across space, we expect that evolution and coevolution will lead to distinct outcomes in different communities, as predicted by the geographic mosaic theory of coevolution (Thompson, 2005). Both isolated and proximate systems favor interactions that minimize inter- and intra-specific competition and maximize resource use efficiency. However, interactions will differ depending on the rate of species turnover and gene flow, both of which are higher on proximate islands. Our approach will help guide future modeling and empirical studies to improve the understanding of how communities assemble through the lens of species interaction networks.

AUTHOR CONTRIBUTIONS

LCP, FSV, and RG planned a workshop which the other authors attended to outline this perspective. FSV and KTA designed simulations to help formulate the framework. LCP, PRG, and MPG designed the conceptual figures. LCP wrote the first version of the manuscript and all other authors contributed to revisions.

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