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# The temporal assembly of plant-pollinator networks following restoration

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1 Abstract

The structure of networks is related to ability of communities to maintain function in the face of species extinction. Understanding network structure and how it relates to network disassembly, therefore, is a priority for system-level conservation biology. We explore the assembly of plant-pollinator communities on native plant restorations in the Central Valley of California. The assembling communities are paired with unrestored field margins (controls) and mature (non-assembling) hedgerows. We determine whether there are change points in the assembly of the communities where the network undergoes significant reorganization. We are also ask how are the individual species changing their interaction patterns? What does this mean for the topology/resilience of the network? We also attempted to adapt a financial model to mutualistic networks. Our biggest difficulty with this approach was to translate the price term to mutualistic systems. We explored a range of approaches, such as number of visits a species performs. However, it seems that financial systems cannot be easily translated to mutualistic systems. In addition, we used a Changing Point Detection Algorithm to assess weather the different communities went through a critical reorganization on their interaction patterns. We were able to identify some changing points in the communities, and also to explore some general patterns commonly used to describe ecological networks. For example, on the network level, networks become increasingly modular and less nested, whereas on the species level, species become more specialized, as resources become more reliable.

Keywords: changing points, temporal networks, hedgerows, species interactions, prefer-

₃ ential attachment

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- The structure of networks is related to ability of communities to maintain function in the face of species extinction.
- A key restoration aim is to facilitate assembly of robust networks; thus it is critical to study how restoration influences the assembly of plant-pollinator interactions.
- In general, however, few mechanisms of network assembly have been developed and analyzed.
- The mostly widely explored mechanism of network assembly, preferential attachment (Barabási & Albert, 1999), predicts that a new species (or node more generally) is more likely to interact with species that are already well-connected ("the rich-get-richer" principle, Barabási & Albert, 1999).
- To date, only a handufullof field studies have examined how networks assemble over time, often using space for time gradients.
- Olesen *et al.* (2008) was investigated day-to-day, temporal assembly of a plant-pollinator network within a season, taking advantage of the extreme seasonality of pollinator tor communities in Greenland. Olesen *et al.* (2008) found that within a season, the network assembly was similar to preferential attachment. New species tended to interact with already well-connected species, likely because these species are either more abundant or more temporally persistent.
  - Studying primary succession along a glacier foreland, Albrecht et al. (2010) found a similar pattern where nestedness, a pattern of interactions where a generalist core interacts with both specialist and generalist species, increased as the community aged.

- Even non-successional temporal dynamics suggest a stable core of generalists persist despite high turnover of peripheral species (Fang & Huang, 2012; Díaz-Castelazo et al., 2010; Alarcón et al., 2008).
- In contrast to the ordered network build-up described by preferential attachment, assembly may be punctuated by significant reorganizations of interactions. For example, as new species are added, resident species change their interaction partners to minimize competition, or become extinct. Such significant reorganizations of interactions, or changing points, have been observed in networks (Peel & Clauset, 2014).

#### 56 Materials & Methods

#### 57 Study sites and collection methods

We surveyed plant-pollinator interaction networks of assembling hedgerows (N=5), as well as in two types of non-assembling communities to serve as controls: unrestored, weedy field margins (N=XX) and established hedgerows (greater than 10 years since planting, N=XX). The sites were located in the Central Valley of California in Yolo, Colusa 61 and Solano Counties. The area is comprised of intensively managed agriculture – primarily monocultures of conventional row crops, vineyards and orchards. Hedgerows were 63 planted along field margins where they do not remove valuable land from production, 64 and are ca. 3-6m wide and approximately 350m long and border large (ca. 30-hectare) 65 crop fields. Hedgerows consist of native, perennial, shrub and tree plantings including 66 Rosa californica, Cercis occidentalis, Ceanothus spp., Heteromeles arbutifolia, Sambucus mexicana, Eriogonum spp., Baccharis spp., Salvia spp. and others. The mean distance between

- monitoring sites was 15 km, and the minimum distance between sites of the same type sampled in the same year was 2 km. The entire area surveyed spanned almost 300 km<sup>2</sup> and the crop fields adjacent to all sites were similarly managed as intensive, high-input monoculture.
- Monitoring began in 2006 and continued through 2015. Sampling of the assembling hedgerows began the year before the area was restored. For logistical reasons, no sampling of assembling hedgerows was conducted in 2010. Sites were sampled between two and five times per year (Table XX). In each round of sampling, the order in which sites were sampled was randomized. Surveys were conducted under sunny conditions when the temperature was above 21°C and wind speed was below 2.5 meters/second.
- Flower-visitors to plants in hedgerows and unrestored controls were netted for one hour of active search time (the timer was paused when handling specimens). Honeybees (*Apis mellifera*) were not collected because their abundance is determined largely by the place-81 ment of hives throughout the region by bee-keepers. All other insect flower visitors that 82 touched the reproductive parts of the flower were collected; however, here we focus only 83 on wild bees and syrphids (representing XX and XX percent of records, respectively), 84 the most abundant and effective pollinators in the system (C. Kremen, A. Klein and L. 85 Morandin, unpublished data). Bee and syrphid specimens were identified to species (or 86 morpho-species for some bee specimens in the genera *Nomada* and *Sphecodes*) by expert 87 taxonomists.

## **SOLUTION** Change point analysis

#### Identifying change points

We employed a change point detection method (Peel & Clauset, 2014) to identify fundamental changes in large-scale pattern of interactions of plants and pollinators. A change point is caused by a merge, split, fragmentation or formation of communities (also called modules or compartments). Following Peel & Clauset (2014), we first defined a probability distribution over the networks using the generalized hierarchical random graph model (GHRG). The GHRG model is able to capture both assortative and disassortative community structure patterns at all scales in the network (Peel & Clauset, 2014). A network G is composed of vertices V and edges  $E \subseteq VV$ . The GHRG model decomposes the N vertices into a series of nested groups, the relationships among which are represented by the dendrogram T. The tips of T are the vertices of G, and the probability that two vertices U and U connect is given by the parameter U. The probability distribution of the network U thus defined as:

$$P(G|T, pr) = p_r^{E_r} (1 - p_r)^{N_r - E_r}$$
(1)

Where  $E_r$  is the observed number of edges between vertices with the common ancestor r, and  $N_r$  is the total possible edges.

Using Bayesian posterior inference and techniques from phylogenetic tree reconstruction, we fit the GHRG model to the networks (Peel & Clauset, 2014). This is accomplished by using a Markov Chain Monte Carlo (MCMC) procedure to first sample the posterior distribution of bipartitions, from which a consensus tree is derived (Peel & Clauset, 2014). ADD NEWMAN PAPER. We used  $\beta$  distributions with the hyperparameters  $\alpha = \beta = 1$  to define priors for  $p_r$ .

Once the GHRG model has been fit to the networks, we determine whether a change 111 point occurred between two time slices. To detect a change point, we compare the fit of 112 two models - one where a change point had occurred between two networks, and one where no change occurred – using posterior Bayes factors. We chose a sliding window of length, w, of four, within which to find change points. Larger windows allow for more 115 gradual changes, and four was the maximum possible with our maximum of nine years of 116 data. Lastly, to calculate a p-value for the Bayes factors, we use parametric bootstrapping 117 to numerically estimate the null distribution (Peel & Clauset, 2014). The change point 118 analysis was carried out using code published online by L. Peel. 119

We next test whether the change points occurring in maturing hedgerows were a component of the assembly process or a product of environmental shifts that lead to network
reorganizations in all types of communities. We model the number of change points as
successes and the total number of years each site was sampled as trails, and use a generalized linear model with Binomial error to test whether the probability of a change point
occurring varied by site type.

#### Characteristics of "core" and "peripheral" communities

To further elucidate the nature of the change points, we examine the characteristics of the species that contributed the reorganization of interactions. To do so, we first generate dendrograms using the GHRG model before and after each change point. We then determined which species belonged to the "core" and "peripheral" network communities at each site. The "core" network communities contain the majority of species and are more basal than the more derived, less specious "peripheral" network communities. We next use a Permutational Multivariate Analysis of Variance (PERMANOVA) (Anderson &

Walsh, 2013) to determine whether the species compositions of the species that belonged to the core and peripheral communities differed. We also test whether core or peripheral species had more variability in their species compositions (i.e., multivariate dispersion, Anderson *et al.*, 2011, 2006).

#### 38 Network structure

The changing points in network structure may contribute to the reorganization of the 139 assembling networks into predictable interaction patterns. Pollination networks exhibit 140 two main structural patterns — modularity (e.g., Olesen et al., 2007) and nestedness (e.g., 141 Bascompte et al., 2006, 2003). In modular community interactions are insular, occurring 142 within separate groups or "modules" more often than between modules. Communities in 143 the network may fragment as the network assembles, enhancing modularity. Conversely, 144 nested networks are like pyramid of interactions, where there are some species that in-145 teract with many species, other species that interact with a subset of those species, and 146 so on. If species entering the network tend to interact with the generalist base of the net-147 work pyramid, nestedness would increase through time. Alternatively, if the network is 148 accumulating specialist species or if species are beginning to limit their interaction niche 149 breath as the network assembles, this would lead to an increase in the network-level spe-150 cialization (Blüthgen et al., 2006) and nestedness would decrease through time. To test 151 whether network modularity, nestedness or specialization changed linearly with assem-152 bly, we used linear mixed models with the descriptive network metrics as the response 153 variable, year of assembly as the explanatory variable, and random effects of site and year.

#### 56 Network robustness

Lastly, we test the changes in interaction patterns associated with network assembly affect 157 the robustness of the network to species loss and to cascading effects. Following Mem-158 mott et al. (2004), we simulate the extinction of plant species the subsequent extinction 159 cascades of pollinator species. Because the reproduction of plant species if facilitated by 160 active restoration efforts, it is unlikely the extinction of pollinators would affect plant pop-161 ulations in the hedgerows. However, plants ceasing to bloom, for example in response to 162 drought, will likely affect the pollinators that depend on them. Plants species were elimi-163 nated based on their degree or abundance, and the number of pollinators that secondarily 164 went extinct is calculated. The area below the extinction curve is used as a measure of net-165 work robustness. We also explored how the robustness to cascading effects changed as 166 community assembled, using algebraic connectivity— the second smallest eigenvalue of 167 the Laplacian matrix (?)— as a proxy for network robustness (e.g. Gaiarsa et al., sub-168 mitted). Algebraic connectivity was first used to describe spectral properties of complex 169 graphs other than ecological networks, and it is related to how difficult it is to turn a net-170 work into completely disconnected groups of nodes, or species (??), and also to flows in . 171 The larger the algebraic connectivity, the more robust a network is to perturbations (e.g. Gaiarsa et al., submitted).

#### 174 Results

## 75 Change point analysis

#### 176 Identifying change points

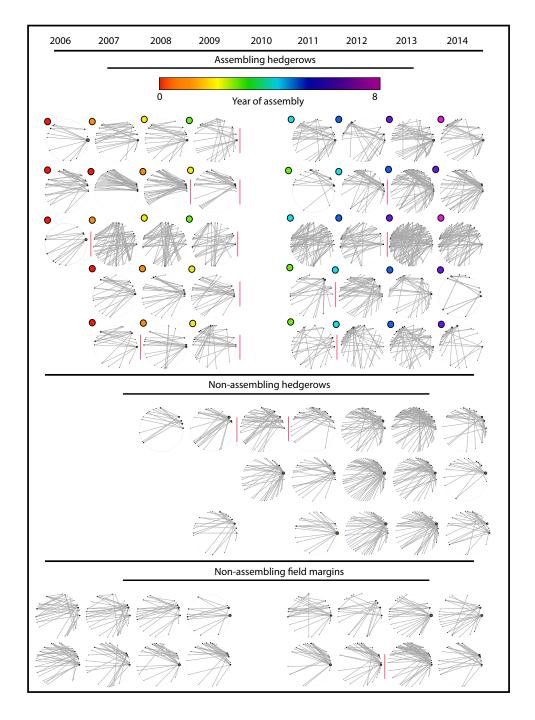


Figure 1: XXX

177 Characteristics of "core" and "peripheral" communities

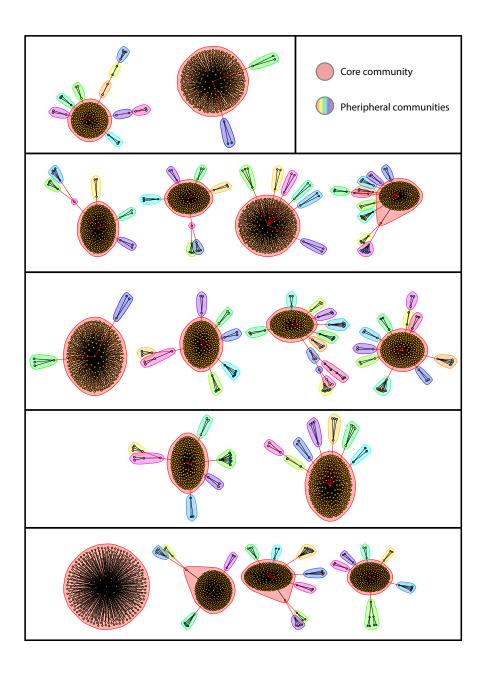


Figure 2: XXX

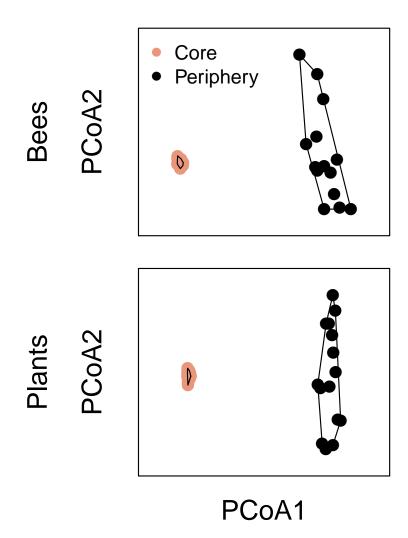


Figure 3: XXX

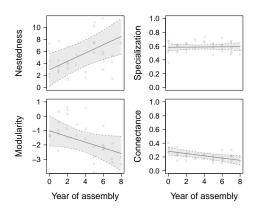


Figure 4: XXX

# Network structure

#### Network robustness

# 180 Discussion

# 181 Acknowledgments

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