# Community assembly on isolated islands: Macroecology meets evolution

A. J. Rominger<sup>1\*</sup>, K. R. Goodman<sup>1\*</sup>, J. Y. Lim<sup>2\*</sup>, F. S. Valdovinos<sup>3\*</sup>, E. Armstrong<sup>1,4</sup>, L. Becking<sup>1</sup>, G. M Bennett<sup>5</sup>, M. S. Brewer<sup>1</sup>, D. D. Cotoras<sup>2</sup>, C. P. Ewing<sup>4</sup>, J. Harte<sup>1</sup>, N. Martinez<sup>3</sup>, P. O'Grady<sup>1</sup>, D. Percy<sup>6</sup>, D. Price<sup>4</sup>, G. K. Roderick<sup>1</sup>, K. L. Shaw<sup>7</sup>, D. S. Gruner<sup>8#</sup>, R. G. Gillespie<sup>1#</sup>

- Environmental Science, Policy, and Management, University of California, Berkeley, California 94720-3114
- 2. Integrative Biology, University of California, Berkeley, California 94720-3140
- 3. Pacific Ecoinformatics and Computational Ecology Lab, Berkeley, California 94703
- 4. Biology, University of Hawaii, Hilo, Hawaii, 96720-4091
- 5. Integrative Biology, University of Texas, Austin, Texas 78712
- 6. Entomology, The Natural History Museum, London, UK SW7 5BD
- 7. Neurobiology and Behavior, Cornell, Ithaca, New York 14853-7601
- 8. Department of Entomology, University of Maryland, College Park, Maryland 20742-4454
- \* Contributed equally; # co-senior authors; corresponding author: R. G. Gillespie, gillespie@berkeley.edu.

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#### Abstract

- Aim Understanding how ecological and evolutionary processes synergistically determine biodiversity patterns remains a central goal in biology. In highly isolated archipelagoes such as the Hawaiian Islands, beyond the reach of frequent colonization, rapid in situ diversification has the potential to keep pace with ecological dynamics such as biotic filtering and demography in structuring local biodiversity. Using ecological theory as a conceptual guide and data from multiple arthropod lineages across the Hawaiian model system, we explore how complex communities emerge from the interplay of ecological (population dynamics, dispersal, trophic interactions) and evolutionary (genetic structuring, adaptation, speciation, extinction) processes.
- Location The Hawaiian Islands (19.5N, 155.5W).
- Methods To infer processes involved in early diversification we synthesize data on genetic structure of select arthropod species across Hawaiian landscapes of known age. Across the range of geological ages of the current high islands (< 1 my to 5 my) we also develop and analyze a plant-herbivore bipartite network. We compare the structure of these networks, measured by nestedness, modularity and the degree distributions, with theoretical predictions derived from the principle of maximum information entropy.
- Results Based on the time perspective provided by the island chronosequence and genetic information, we demonstrate that species in lower trophic levels develop local divergence more quickly than species of higher trophic levels. Higher trophic levels also show endemism, though it evolves more slowly and over larger areas. Moreover, in analyzing plant-herbivore networks across an increasing substrate age gradient we find trends of higher specialization and increasing deviation from the statistical steady state expected from theoretical predictions of food web structure.

Main conclusions We show how ecological theory can leverage natural experiments on oceanic islands of known chronologies to understand how the interplay between evolutionary and ecological processes has shaped present-day biodiversity. We advocate for combining perspectives gained from coupled molecular and community-level data analyzed in the context of ecological theory. Theory provides a lens through which to identify interesting biological outliers. We further show the utility of applying theory in a chronosequence context to better illuminate the interplay of ecological mechanisms, speciation, extinction and adaptation in driving contemporary biodiversity patterns.

#### 3 Introduction

Contemporary biodiversity is an unresolved product of speciation, extinction and dispersal all conditioned by ecological interactions with the biotic and abiotic environment. Because these processes occur on different temporal and spatial scales and may be interactive with nonlinear feedbacks and lags among them, disentangling the relative influence of local ecological mechanisms from evolutionary and historical processes is challenging (Ricklefs, 2004). The integration of ecological and evolutionary theory has the potential to reveal dynamics that generate biodiversity. The evolutionary processes of speciation and extinction tend to be viewed as regulating 41 regional species pools, occurring in a manner largely removed from local ecology (Hubbell, 2001; Cavender-Bares et al., 2009; Wiens, 2011). Ecological mechanisms tend to be viewed as packing standing diversity into local communities through competition, facilitation, and neutral ecological drift (Hubbell, 2001; Tilman, 2004; Bascompte & Jordano, 2007; Chase & Myers, 2011; Borer et al., 2014). Recent theoretical advances have further refined the causes and consequences of ecological drift (Hubbell, 2001; Rosindell & Phillimore, 2011; Rosindell et al., 2012), re-vitalized classical niche-based mechanisms such as niche parti-

tioning (Tilman, 2004; Chesson, 2000), competition and predation (Borer et al., 2014), and

put species interactions in a network theoretic context (Williams & Martinez, 2000; Brose

et al., 2006; Berlow et al., 2009). The combined advances of ecological theory, with its broad predictive power, and insights into evolutionary mechanisms based on inference from contemporary patterns of species, genetic, or phylogenetic diversity (e.g., Kreft & Jetz, 2007; Jetz et al., 2012; Wiens & Donoghue, 2004; Wiens, 2011) have set the stage to address long-standing questions of how evolutionary history can drive common patterns in contemporary ecology (Ricklefs, 1987). Here, we propose an integrative framework to study evolutionary community assembly. We then provide an initial test using arthropod lineages in the model archipelago of the Hawaiian Islands using mostly published data. We estimate metrics of evolutionary and ecological dynamics across communities that range in age from 500 yr to 5 myr: (1) The timeline for the development of genetic discontinuity and the extent to which taxa across communities differ in the rates that populations change from panmixia to fully differentiated species. This is contextualized with (2) macroecological metrics of community structure, using predictions from statistical steady state and ecological network theory to examine changes over the island chronosequence.

## Hotspot oceanic archipelagos as model systems

Hotspot oceanic island systems are opportune model systems for studying the interplay of local ecological mechanisms and large—scale, historical, and evolutionary drivers of biodiversity patterns. Such island systems are discrete in space and in time due to their sequential formation as the tectonic plate moves over a volcanic hotspot. We hypothesize that the contributions of evolutionary and ecological assembly will vary according to geological age of the environment, taken as an indicator of the total time communities have had to assemble and over which in situ diversification could occur. Age—structured, hot—spot island archipelagoes thus have the potential to stratify the eco—evolutionary process of community assembly.

For example, younger communities by necessity originate mostly from initial immigration
(from neighboring older volcanoes and islands or from the mainland, depending on the level
of isolation), and thus should be dominated by ecological mechanisms operating on a source

pool whose evolution is removed from the local setting. Conversely, older islands could allow observation of the combined interaction and feedback of the diversification of the source pool and local ecological dynamics. The temporal stratification within such archipelagoes hence provides an opportunity to disentangle these interacting processes. Moreover, because dispersal, and hence connectivity between sites, differs between taxa, the relative role of evolutionary and ecological assembly will differ between taxa. While many archipelagos around the world share these biotic and geologic properties, the Hawaiian archipelago provides a particularly useful system for study because its geological chronology (Price & Clague, 2002) and patterns of biodiversity are well characterized (Wagner & Funk, 1995).

#### Be Development of genetic discontinuity

Movement of individuals among localities connects the population dynamics of those localities. Even moderate levels of genetic connectivity among geographically separated populations limits the potential for local divergence (Wright, 1978; Slatkin, 1987). Thus, in the face of connectivity among populations, one predicts that the structure of ecological communities will remain similar across space. By contrast, when connectivity is low, not only are the ecologies of populations in different localities free to vary, but genetic divergence is also more likely. For these reasons, the magnitude of connectivity among population provides a measure of the relative importance of ecological processes and evolutionary processes in determining differences among ecological communities. Here, by using the chronosequence, we can apply this approach to sets of communities from young to old and to taxa representing different trophic levels.

## Macroecological metrics

While we expect the mechanisms underlying the generation and maintenance of biodiversity to change across chronological sequences, studies to date have rarely moved beyond reporting basic patterns (Gillespie & Baldwin, 2010). Theory provides a necessarily simplified view of biodiversity and deviations from theory can reveal which more biologically realistic

mechanisms likely underlie observed patterns. The Maximum Entropy Theory of Ecology (METE; Harte, 2011) provides a prediction for idealized ecological communities in statistical 104 steady state. Statistical steady state describes the situation in which a system's behavior is 105 governed by a simple set of state variables and no further system-specific mechanisms are re-106 quired (Harte, 2011; Harte & Newman, 2014). Thus METE, while in a sense neutral, makes 107 fewer assumptions than neutral theory (Hubbell, 2001), allowing for the possibility that 108 myriad ecological mechanisms influence communities. However, METE assumes that these 109 mechanisms have no statistical effect on the macroscopic biodiversity patterns of the system. 110 Real world deviations from METE can provide insights into the processes driving ecology 111 away from this statistical steady state and toward alternate system states (Harte, 2011). 112 We expect that different aged communities along the Hawaiian chronosequence will devi-113 ate differently from METE, because we hypothesize the processes of speciation, extinction, 114 adaptation and colonization may themselves drive Hawaiian communities out of statistical 115 steady state. 116

METE can successfully predict various metrics of an ecological community (Harte, 2011), 117 including network metrics that describe trophic interactions between species (Williams, 2010; 118 Harte, 2011). Ecological network theory incorporates evolutionary concepts such as coevolu-119 tion (Bascompte & Jordano, 2007; Donatti et al., 2011; Nuismer et al., 2013) and has clear ties with macroecological questions of the distribution of abundance and body size across species 121 (Berlow et al., 2009; Williams, 2010; Harte, 2011). The distribution of linkages in ecological 122 networks has been used to determine whether plant-animal interaction networks assemble 123 neutrally or through deterministic processes (Vázquez et al., 2005). Analysis of other net-124 work metrics such as modularity (the degree to which species interact in semi-autonomous 125 modules) and nestedness (the degree of asymmetry in interaction between specialists and 126 generalists) can further illuminate underlying eco-evolutionary processes driving patterns of 127 species interactions (Bascompte & Jordano, 2007; Donatti et al., 2011; Nuismer et al., 2013). 128

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In this paper we integrate methods from population genetics to theoretical ecology using

the chronosequence of the Hawaiian Archipelago to understand the nexus between ecological and evolutionary forces community assembly. Moving rom young to old across the chronose-131 quence we evaluate (1) the rate and pattern of genetic connectivity among populations of 132 taxa from different trophic levels as they diversify from populations to form new species; (2) 133 the processes underlying the structure of species interaction networks given the backdrop 134 of population divergence; and (3) the processes involved in diversification as species form 135 and accumulate and how this dynamic drives deviations from statistical steady state. We 136 use data (mostly published) on population genetic structure and species interactions as a 137 proof of concept. With this framework, our goal is to show how communities develop over 138 ecological-evolutionary time, and the dynamic feedbacks involved assembly. 139

#### 140 Methods

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#### Hawaii as an eco-evolutionary study system

The geological landscape of the Hawaiian Islands offers a matrix of volcanic substrates mapped in fine detail by chronological age and geochemical composition (Sherrod et al., 2007). On Hawaii Island in particular, independent gradients in elevation and precipitation interact within a matrix of substrate ages ranging from contemporary (active) to 500,000 years. Ongoing volcanic activity has created a dynamic mosaic of habitats with transi-146 tory to long-lasting fragmentation within landscapes. Isolation on scales of hundreds to 147 thousands of meters and hundreds to thousands of years can be sufficient for genetic differ-148 entiation among some arthropod populations among habitats (Goodman et al., 2012; Eldon 149 et al., 2013; Bennett & O'Grady, 2013), while insufficient to isolate others (Vandergast et al., 150 2004). On larger spatial and temporal scales, distinct volcanoes and islands, with their semi-151 independent histories of evolutionary community assembly from recent to ancient, comprise 152 a space-for-time geological chronosequence spanning from present day up to 5 million years 153 across Hawaii Island to Kauai. 154

To investigate how ecological patterns change in response to varied evolutionary contexts

we selected four focal sites across the chronosequence of substrate and island ages (two on Hawaii Island, one on Maui and one on Kauai). Focal sites were selected to have similar forest composition (dominated by *Metrosideros polymorpha*: Myrtaceae), elevation (1100-1400m), and climate (mean annual precipitation 2000-3000 mm), while deliberately varying substrate age. These forested montane sites are well-studied and primarily composed of native plant and arthropod species. The four sites span the chronosequence from 0.00025 million years (Kilauea and Kohala (Hawaii Island); Waikamoi (Maui), Kokee (Kauai); see Fig. 1).

#### 163 Compilation and analysis of genetic data

To evaluate the balance between region immigration and potential for local ecological differentiation, we measured how molecular variation is partitioned within species within locations of known substrate age on Hawaii Island and Maui. We compiled published and new data sets for a diversity of native Hawaiian arthropod groups that represent a spectrum of trophic levels:

- 1. Herbivorous Nesosydne planthoppers (COI and microsatellites; Goodman et al., 2012,
  GenBank accession numbers XXX-XXX); Trioza psyllids (COI, cytB; GenBank accession numbers XXX-XXX); and fungivorous Drosophila sproati: (COII; Eldon et al.,
  2013) that maintain tight host plant associations.
- 2. Detritivorous *Laupala* crickets (AFLPs; Mendelson *et al.*, 2004); and
- 3. several predatory spiders species (COI and allozymes; Roderick *et al.*, 2012; Croucher *et al.*, 2012)

In the case of previously unpublished molecular data, sequences for *Nesosydne* planthoppers were generated following protocols described in (Goodman *et al.*, 2012) and sequences
for *Trioza* psyllids were generated following protocols described in Percy (2003) with primers
given in Simon *et al.* (1994) and Timmermans *et al.* (2010). Existing genetic data from across
Hawaii Island and Maui (including, but not limited to the focal sites), provide an estimate
of how arthropod populations have accumulated genetic diversity and divergence within the

dynamic landscape of the focal sites.

We used analysis of molecular variance (AMOVA) to examine how genetic variation is 183 partitioned at two scales of population structure: among sites within volcanoes and among 184 volcanoes on both Hawaii Island and Maui Nui. All analyses of allozyme and DNA sequence 185 data were performed in Arlequin v. 3.5 (Excoffier & Lischer, 2010) using the AMOVA pro-186 cedure to compute  $F_{ST}$ , a measure of genetic variance, or, where possible  $Phi_{ST}$ , an  $F_{ST}$ 187 analog that incorporates genetic sequence information. The Laupala AFLP data were ana-188 lyzed using TFPGA v. 1.3 (Miller, 1997), using the same hierarchical approach as described 189 above. To provide a temporal framework for the population differentiation analysis we as-190 sembled divergence dating information from the literature for as many of the taxa as possible 191 and additionally implemented a new divergence dating analysis for Tetragnatha spiders (see 192 supplementary information). 193

To explicitly evaluate the role of landscape age in allowing in situ genetic diversity and potential for divergence we analyzed how within site  $F_{ST}$  varies with the geologic age of volcanoes on Hawaii Island and Maui Nui. For each volcano we calculated  $F_{ST}$  or  $\Phi_{ST}$  (Excoffier & Lischer, 2010) for each taxon between sites within volcanoes.

#### 198 Construction of plant-herbivore networks

Bipartite networks describe the topology of ecological interactions between two guilds of organisms (e.g. herbivores and their plant hosts). Quantitative information on the relative importance of interaction links can be incorporated into network analyses (Vázquez et al., 2009); however, currently available data are restricted to binary networks, those that describe only the potential for interaction between any two species, but not the relative frequency of that interaction to each species.

We compiled species lists of all endemic hemipteran herbivores for each focal site from published species accounts (see supplement for full list). Species accounts and other published sources were used to determine the presence, probable presence, or probable absence

of each Hemiptera species at each of our four focal sites. A documented presence was defined
as a known specimen collected at the focal site; a probable presence was defined as a species
whose abiotic tolerances and known geographic range (see supplement) overlap with a focal
site but no known specimen exists confirming its presence. Probable absence was assumed
when neither criteria for presence or probable presence are met. Two sets of species lists for
each focal site were compiled: a conservative data set composed of only documented presence
occurrences and a less conservative data set that also included probable presences.

Host plants for each hemipteran species were determined from published species accounts.

Data on host plant use at each specific site were not available so we assumed that if a known host plant was present at a site it would eventually be used. Host plant occurrence in the focal sites was determined using distribution models for 1158 species of Hawaiian plants (Price, 2012). Each focal site was spatially joined in a geographic information system with all coincident plant distribution models that fell within its boundaries. Two sets of resulting focal site-specific networks were constructed: one using the conservative data set of hemipteran species presences and the other using the less conservative data set.

#### Analysis of plant-herbivore networks

We hypothesize that communities differentially depart from statistical steady state along the continuum form those dominated by ecological processes to those with potential complex evolutionary feedbacks We used METE (Williams, 2010; Harte, 2011) to compute the statistical steady state for the hemipteran degree distribution (distribution of the number 227 of plant hosts to each hemipteran species). To evaluate how well METE predicts the data 228 we simulated METE-conforming communities of the same number of species and links as 229 observed. We then calculated the log-likelihood of each simulated data set and compared 230 the resultant distribution of log-likelihoods under the hypothesis that METE is true, to the 231 observed log-likelihood. This comparison is identical in approach to a z-score test using a 232 Monte Carlo simulation to estimate the sampling distribution of log-likelihoods. R scripts used for METE estimation and Monte Carlo methods are available in the supplement.

To further investigate how in situ diversification leaves a potentially unique signature on 235 network structure we analyzed the number of links assigned to each hemipteran species (the 236 degree distribution) separately for island endemics (those species found on only one island 237 and thus likely derived from in situ diversification) versus island cosmopolitans (those species 238 found on multiple islands). To compare species' degree distributions between endemics and 239 cosmopolitans across sites of different ages we conducted a generalized linear model with 240 binomial error, treating site identity as a categorical predictor. Binomial errors effectively 241 account for network size due to the bounded support of the binomial distribution. 242

To understand how other network properties change with ecosystem substrate age, we calculated two widely used descriptive network metrics across sites—nestedness, which describes the degree of asymmetry of species interactions connecting specialists and generalists (Bascompte et al., 2003; Ulrich et al., 2009), and modularity which describes the degree to which interactions are concentrated within subsets of species but not between subsets (Newman & Girvan, 2004; Olesen et al., 2007).

We calculated nestedness using the NODF metric (Almeida-Neto et al., 2008) as imple-249 mented in the R package vegan (Oksanen et al., 2013) and modularity using a variety of algorithms implemented in the R package igraph (Csardi & Nepusz, 2006). These metrics 251 are not directly comparable across networks of different size and connectance (Ulrich et al., 252 2009), so for each metric in each network we calculate z-scores using a null model that ran-253 domizes network structure while maintaining certain aggregate network properties (Ulrich 254 et al., 2009). These z-scores are calculated as the difference between the observed network 255 metric minus the mean of the null model divided by the null model standard deviation, or 256  $(x_{obs} - \bar{x}_{sim})/sd_{sim}$ ). Because z-scores can be highly sensitive to the choice of null model 257 (Ulrich et al., 2009) we implemented both a probabilistic null model (Bascompte et al., 2003) 258 and a null model that strictly constrains the degree distributions of plants and herbivores 259 (Ulrich et al., 2009). The probabilistic null using the frequency of interactions as the proba-260

bility that a randomized link gets assigned to that cell in the interaction matrix (Bascompte et al., 2003); thus the probabilistic null constrains row and column sums in probability but not absolutely.

#### Results

#### Population genetic inference of discontinuity among populations

The analysis of molecular variance (AMOVA) revealed evidence of significant population 266 genetic structure from the smallest to the largest spatial scales examined, all within a very 267 recent timeframe. For mitochondrial loci, the amount of statistically significant molecular 268 variation partitioned to among sites within volcanoes ranged from 0.037-0.92 and to the 269 among volcanoes from 0-0.30. Corresponding variation at multilocus nuclear loci between-270 sites within volcanoes ranged from 0.21-0.58 and among volcanoes, 0.04-0.34. Taxa in the 271 lower trophic levels (herbivorous sap-feeding Hemiptera: planthoppers and psyllids) had as 272 much or more molecular variation partitioned at the among-site, within volcano level than 273 the among volcano level, while the predatory spiders were less structured at localities within volcanoes compared to between (Table 1). The analysis of genetic population structure across the chronosequence of localities revealed a similar pattern. The herbivores show high genetic population structure among localities on young volcanoes relative to between localities on 277 older volcanoes (Fig. 2). By contrast, predatory spiders exhibited higher genetic population 278 structure only on older volcanoes (e.g. Maui). 279 The observed levels of genetic divergence have evolved rapidly. For example, within 280

The observed levels of genetic divergence have evolved rapidly. For example, within species genetic divergence in planthoppers has evolved in as little as 2,600 years (Goodman et al., 2012). For species from Hawaii Island for which phylogenetic data provide divergence times, estimates of dates of species divergence range from 0.341.15 million years, with additional within-species genetic divergence developed subsequently (Table 1).

#### 285 Evolving network structure

The Hemiptera species degree distribution varied across the chronosequence with both the youngest and oldest sites deviating most from the statistical steady state maximum entropy predictions (Fig. 3). In the middle aged site of Kohala, minor deviations from maximum entropy are no different than expected by chance indicating the Kohala Hemiptera assemblage matches the predictions of maximum entropy.

The generalized linear model revealed that there are also significant differences between 291 the degree distributions of island endemics (those species found on only one island) ver-292 sus island cosmopolitans (those species found on multiple islands; Fig. 3). Endemics show 293 significantly lower degree distributions overall (i.e., more specialization) compared to more 294 generalist cosmopolitan species. Endemics become significantly more generalist on the mid-295 dle aged Maui site; however this pattern disappears when analyzing links to plant genera 296 instead of species. The slightly younger Kohala shows increased generalization overall. When 297 considering the degree distribution defined by trophic links to plant genera instead of plant 298 species, the pattern of increased generalization holds for the Kohala but endemics on Maui 299 no long show a difference in their degree distributions from other island endemics. This 300 change in pattern suggests that increased generality of Maui endemics may be driven by 301 increased intra-genus plant diversity on that island. 302

Network nestedness decreased with age while modularity increased (Fig. 4). This trend is found in networks constructed from both more and less stringent geographic criteria (supplemental Fig. 3). Choice of null model changed the magnitude of modularity and the sign of nestedness z-scores; however, the relative pattern of decreasing nestedness and increasing modularity remained across the different null models used to standardize network metrics (supplemental Fig. 2). The patterns are also robust to sampling intensity, as demonstrated by a rarefaction analysis (supplemental Fig. 4).

#### Discussion

By combining disparate data with a novel combination of analytical approaches that in-311 corporate population genetics, bipartite networks and maximum entropy theory, our results 312 present evidence for the timeline over which evolution begins to keep pace with ecology in 313 determining the local diversity of communities. Taxa in the lower trophic levels, as compared 314 to higher trophic guilds, developed genetic discontinuities more quickly along the chronose-315 quence and at much smaller spatial scales (Table 1, Fig. 2), allowing them the opportunity 316 to diverge ecologically. Network nestedness decreased while modularity increased with age 317 (Fig. 4), indicating a possible shift from assembly driven by ex situ immigration early on, to one based on in situ co-diversification with host plants (Bascompte & Jordano, 2007; Donatti et al., 2011). This possibility is further strengthened by the observation that single island 320 endemics show more specialization compared to more broadly distributed species (Fig. 3). 321 At intermediate modularity and nestedness, the distributions of the number of links assigned 322 to each hemipteran species showed the least deviation from the METE prediction (Fig. 3), 323 suggesting that at the transition from primary succession to evolutionary assembly, these 324 plant-herbivore communities reach statistical steady state. 325

#### Development of genetic discontinuity at different trophic levels

The analysis of available genetic data presented here indicates that divergence is occurring 327 within the islands at small spatial scales and over short time periods (Table 1, Fig. 2). 328 Furthermore, the scale of population structure varies with trophic position, with the sap-320 feeding herbivores in this study showing structure at smaller scales compared to detritivorous 330 crickets and predatory spiders (Table 1, Fig. 2). Population structure within species allows 331 for populations to take independent evolutionary trajectories, especially when aided by other 332 evolutionary processes that may be acting differentially across each species' range. A variety 333 of factors have been implicated in the genetic divergence of populations and species in lineages 334 described here, including combinations of genetic drift associated with geographic isolation (Percy, 2003; Gillespie, 2005; Mendelson & Shaw, 2005; O'Grady et al., 2011; Goodman et al., 2012), adaptation associated with ecological interactions of competition, predation, and mutualism (Gillespie, 2004; Blackledge & Gillespie, 2004; Roderick & Percy, 2008), and sexual signaling (Mendelson & Shaw, 2005; Percy & Kennedy, 2006; Magnacca et al., 2008; Goodman et al., in review).

The sap-feeding Hemiptera group Nesosydne (Goodman et al., 2012) provide evidence 341 that some period of geographic isolation preceded divergence of sexual signals (Goodman 342 et al., in review). Shifts in plant host use are also involved in the process of diversification 343 in this group (Roderick & Percy, 2008). In a similar radiation of leafhoppers, Nesophrosyne 344 (Bennett & O'Grady, 2013), host plant specialization was implicated in driving species radi-345 ations up until approximately 1 million years ago, when plant niches were mostly exhausted 346 on Maui; following this period, speciation, largely on the Hawaii Island, shifted to geographic 347 mechanisms of diversification. Our network analysis indicates that specialization and modu-348 larity begin to show pronounced signals in network data on Maui (Figs. 4, 3), in agreement 349 with the Nesophrosyne results and indicating that an approximate age of 1 million years 350 may be necessary for host plant specialization to become the dominant process in the se-351 quence of diversification. Other taxa at lower trophic levels, such as the herbivorous Trioza psyllids, detritivorous Laupala crickets and fungivorous Drosophila, show similar signals of geographic isolation combined with ecological and sexual processes driving genetic diver-354 gence and diversification across sites as young as those on the Hawaii Island (Percy, 2003; 355 Percy & Kennedy, 2006; Mendelson & Shaw, 2005; Magnacca et al., 2008; O'Grady et al., 356 2011). As a contrast, spiders, which are predatory, only develop genetic discontinuities at 357 larger spatial and temporal scales. Most important in the context of community assembly is 358 that endemic sap-feeding herbivores developed structure quickly (on the order of less than 359 0.1 million years; Table 1), with predatory spiders showing local endemicity more slowly 360 (Table 1). 361

#### Macroecological metrics: Network structure and steady state

On the geologically youngest volcano, Kilauea, ecological assembly should be the dominant 363 process there. The results of network analysis support this hypothesis with Kilauea showing 364 substantial nestedness and limited modularity (Fig. 4). Nestedness is likely to result if new 365 species arriving by immigration have a high probability to eat or be eaten by the generalist 366 species already present at the site (Bascompte et al., 2003). In this way we might expect 367 Kilauea to also conform to the statistical steady state predication of maximum entropy. 368 However, the observed deviations from maximum entropy at Kilauea are largely driven by 369 a surplus of singleton links (Fig. 3). These in turn likely result from incomplete assembly, 370 and thus lower species richness, of the plant and herbivore biotas. Future research should 371 focus on the observation from genetic analysis that indicates discontinuities can arise within 372 species on short timescales that, in some taxa, include the greater landscape of Kilauea (Table 373 1). Conversely, Kohala shows a statistically significant agreement with maximum entropy 374 perhaps because the Kohala site, at intermediate age (150 ky), has experienced complete 375 ecological succession but is still too young to be driven away from statistical steady state by 376 specialization and rapid in situ diversification driven by host plant preference 377

The older Maui and Kauai sites show strong deviations from expectations of maximum 378 entropy theory (Fig. 4), which is consistent with our hypothesis that the influence of evo-379 lutionary assembly on these biotas drives them away from statistical steady state. The 380 application of maximum entropy to ecology does not currently take into account evolution 381 (Harte, 2011). Indeed the use of maximum entropy in ecology is inspired by its application to 382 physical systems whose change through time is simple and lacks the evolutionary memory of 383 biological systems, potentially a far cry from the complex change through time produced by 384 speciation, extinction and adaption to novel ecosystems (Eldredge, 1989). Maui and Kauai 385 show strong evidence of evolutionary assembly driven by specialization and diversification 386 on host plants, particularly demonstrated by decreased nestedness and increased modularity 387 (Fig. 4). Modularity is known to result from coevolution selectively driving the traits of 388

interacting species towards convergence (Donatti et al., 2011; Nuismer et al., 2013).

The analysis of island endemic and cosmopolitan (archipelago-wide) Hemiptera species 390 sheds further light on the evolution of the networks they form. Endemics are always more 391 specialized than cosmopolitans, further supporting the hypothesis that in situ diversifica-392 tion and evolutionary assembly favor coevolution. At the Kohala site, which showed the 393 best fit to maximum entropy theory, endemic and cosmopolitan species alike show increased 394 generalization (i.e. higher degree; Fig 3), while at the youngest site Kilauea, specialist en-395 demics are limited by low plant diversity and thus show more apparent specialization (Fig. 396 3). Conversely at the oldest site on Kauai, where plant diversity is not limiting (Kitayama 397 & Mueller-Dombois, 1995), endemics again show decreased degree and thus genuine spe-398 cialization (Fig. 3). On Maui, endemics show statistically significant increases in apparent 399 generalization but this pattern disappears when analyzing the data at the resolution of plant 400 genera, thus suggesting that Hemiptera species endemic to Maui are no more generalized on 401 plant genera but instead may benefit from the diversification of plant species within genera 402 on Maui. 403

#### 404 Future Research

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The analyses presented here indicate strong patterns of a dynamic assembly process leading to contrasting hypotheses concerning the relative importance of ecological and evolutionary depending on the evolutionary age of the community under observation. In future work we will tackle these hypotheses using detailed quantitative ecological and genomic data collected from across the Hawaiian archipelago.

- 1. In younger communities we hypothesize that
- (a) during periods of ecological assembly, communities strongly influenced by immigration will resemble random samples from regional source pools and thus metrics describing these communities will largely match expectations of statistical steady state after primary succession has completed;

- (b) the exception will be communities still undergoing primary succession, which will change rapidly through time and represent non–random samples of source pools;
- (c) we also predict that these communities will exhibit a nested network structure, assuming new species will eat or be eaten by the generalist species (Bascompte et al., 2003) already present in the community.
- 2. Following the same logic, in older communities we hypothesize that

- (a) during periods of evolutionary assembly, if processes such as niche exploration, adaptation and speciation happen fast enough to keep pace with immigration, the resultant communities could be driven into alternate evolutionary states that fail to meet the predictions of purely statistical theories that do not account for evolutionary dynamics (Harte, 2011);
- (b) networks in such communities should exhibit higher levels of specialization and modularity (Bascompte & Jordano, 2007; Donatti et al., 2011; Nuismer et al., 2013).
- 3. Systems undergoing rapid ecological and evolutionary change are generally expected to deviate most from statistical steady state; thus we expected populations in such communities to show genetic signatures of rapid change, from bottlenecks or population expansion to selection.

#### Evolutionary data: Diversification within species

The current study demonstrates that taxa of different trophic guilds differ in the scale at which differentiation occurs, and highlights the importance of fragmentation of the landscape in facilitating differentiation. For some taxa fragmentation clearly allows genetic separation. For others, in particular those that are more connected, the fragmentation can provide a way of enhancing adaptive differentiation (Gillespie & Roderick, 2014). Future work is aimed at gathering genomic SNP data for focal taxa within this system that represent different trophic levels. We will use it to understand taxonomic differences in the rate of differentiation, to assess the roles of genetic fusion and fission, and to detail the relative rates of speciation and

extinction across the island chronosequence.

For lineages characterized by extensive ecological diversification, recent work has high-443 lighted the potential role of multiple colonizations and admixture in enhancing variability: while a break in gene flow is necessary for adaptive differentiation, hybridization and genetic 445 admixture are key in the generation of adaptive variation and functional novelty (Seehausen 446 & Schluter, 2004; Rius & Darling, 2014). Numerous studies demonstrate that the negative 447 effects of genetic founder effects may be offset if different colonization events result in mul-448 tiple genotypes within the introduced population (see Rius & Darling, 2014, and citations 449 therein). This highlights the potential role of admixture among successively introduced pop-450 ulations in providing the genetic variation to allow adaptive evolution over surprisingly short 451 time scales. 452

#### Ecological data: Assembly of species into communities

Our results show that the island chronosequence can reveal important insights into the process of community assembly, namely that ecological processes dominate in younger environments, with evolutionary processes becoming more important later. However, in order to understand the nature of the assembly process and the dynamic nature of the feedbacks involved, future work is focusing on conducting broad sampling of all macroscopic arthropod taxa at a number of sites across the age gradient, thus allowing assessment of changes in overall species composition and diversity across all players in the time-calibrated landscape (sensu Gruner, 2007).

Such data will also allow us to test how entire arthropod communities of different aged substrates deviate from statistical steady state as predicted by METE (Harte, 2011). For example, predators, whose assemblages are likely more dominated by immigration and ecological assembly (Fig. 2) are hypothesized to never show strong deviations from METE predictions whereas herbivores are hypothesized to show increasing deviation with age and potentially at the youngest sites as well in agreement with the network results of this paper (Fig. 3).

The current study provides a framework for the use of chronologically arranged oceanic island systems to examine the interplay between evolutionary and ecological processes in shaping biodiversity. We analyze preliminary molecular and community—level data together in the context of ecological theory to demonstrate how this approach can provide insights into how communities develop over ecological—evolutionary time, and the dynamic feedbacks involved in assembly.

## 475 Acknowledgement

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#### Biosketch

The authors are part of a large team involved in understanding how biodiversity and com-643 plex ecosystems emerge from ecological and evolutionary processes. The project, funded by 644 the National Science Foundation's "Dimensions in Biodiversity", focuses on the geological chronosequence provided by the Hawaiian Islands. Each of the co-authors are involved in 646 the effort, although Rominger, working closely with Goodman, Lim, and Valdavinos, played the key role in tying the elements together for the current manuscript. Macroecological 648 tools that have been employed for the study include those developed by Harte and Martinez. From the more empirical side, Gruner has long standing projects in the Hawaiian Islands on the ecological underpinnings of community diversity. Rominger, Lim and Fernandez bring 651 a macroecological perspective to the project. From the evolutionary perspective, Shaw's 652 research focuses on Laupala crickets; Ewing on nitidulid beetles, Goodman Bennet and Rod-653 erick on Hawaiian planthoppers; Percy on endemic psyllids; O'Grady and Price on Hawaiian 654 *Drosophila*; and Cotoras with Gillespie on spiders. 655

#### 656 Tables

Group	Species	Island(s)	Data (# individuals: length/ # markers)	Between Volcanoes % variation (p-value)	Within volcanoes % variation (p-value)	Data source	Timeframe citation
Planthoppers	Nesosydne chambersi	Hawaii	mtDNA COI (187:653) msat (292:13)	0.05***  0.04***	0.77***  0.21***	Goodman et al. 2012	WSD <sup>1</sup> : Saddle Road pops=2.6 (95% HPD: 1.2- 35.1) x 10 <sup>3</sup> yrs WSD <sup>1</sup> : Mauna Kea/Mauna Loa pops: 20.1 (95%HPD: 7.4-135.0) x 10 <sup>3</sup> yrs
							Goodman et al. 2012
	Nesosydne raillardiae	Hawaii	mtDNA COI (33:581)	0.26***	0.49***	this paper	na
	Nesosydne bridwelli	Maui	mtDNA COI (34:677)	na	0.18**	this paper	na
Psyllids	Trioza HB <sup>4</sup>	Hawaii	mtDNA COI & cytB (29:857)	-0.14***	0.92***	this paper	na
	Trioza HC <sup>4</sup>	Hawaii	mtDNA COI & cytB (17:857)	0.17**	0.53**	this paper	na
Fly	Drosophila sproati	Hawaii	mtDNA COII (232:570)	0.11***	0.81***	Eldon et al. 2013	Max age <sup>2</sup> =1.15 (95%HPD: 0.75-1.5) my. Magnacca and Price, in review
Cricket	Laupala cerasina	Hawaii	AFLP (631)	0.30***	0.58***	Mendelson and Shaw 2005	na
Spiders	Tetragnatha anuenue	Hawaii	mtDNA COI (162: 607) allozymes (12:9)	0.23*** na	0.041*** na	Roderick et al. 2012	na
	Tetragnatha brevignatha	Hawaii	mtDNA COI (54:605)	0.16*	0.00	Roderick et al. 2012	T. macracantha Max age2=0.34 (95%HPD: 0.14-0.58) my. Supplementary info, this paper
	Tetragnatha quasimodo	Hawaii	mtDNA COI (149:439) allozymes (46:9)	0.09*** 0.34***	0.037*** na	Roderick et al. 2012	Node age <sup>2</sup> =0.80 (95%HPD: 0.50-1.16) my. Supplementary info, this paper
	Theridion grallator	Hawaii	mtDNA COI (209:1270) allozymes (224:8)	0.30*** 0.19***	0.05*** na	Roderick et al. 2012	Node age <sup>3</sup> =0.56 (95%HPD: 0.37-0.75) my. Croucher et al. 2012
	Ariamnes spp.	Hawaii	mtDNA COI (8:420)	0.05	na	Roderick et al. 2012	na

<sup>\*\*\* &</sup>lt; 0.001, \*\* < 0.05, \*<0.10; na = no information available

Table 1: Proportion of genetic variation distributed at between volcanoes and among sites within volcanoes.

<sup>&</sup>lt;sup>1</sup>WSD = Within-species divergence, estimated using IM.

<sup>&</sup>lt;sup>2</sup>Max age = the node age of the phylogenetic split between this species and its sister species, calculated using divergence dating analyses performed in BEAST. In most cases, this will be an overestimate of the node age of the species itself, but is the best

<sup>&</sup>lt;sup>3</sup>Node age = Age of the most recent common ancestor of the monophyletic group on Hawaii Island, estimated using BEAST.

<sup>4</sup> These *Trioza* species are in the process of being described; HB and HC are their provisional identifiers (Percy in prep).

# Figure captions

Figure 1. Substrate age map of the islands of Kauai, Maui and Hawaii Island. Colors correspond to substrate age from young (yellow) to old (blue). Focal sites are shown as black circles while sampling sites for genetic data are represented by gray circles.

Figure 2. Genetic population structure  $(F_{ST})$  among sites within volcanoes with volcanoes age for insects and spiders. The plant-feeding groups, specifically the sap-feeding Hemiptera, show high genetic structure among sites on young volcanoes relative to older volcanoes whereas detritivores (crickets), fungivores (Drosophila), and in particular predators (spiders), show little structure on young volcanoes. For spiders, substantial structure develops only later in the chronosequence, e.g., on Maui at approximately 1 million years. Number refer to different species: 1.  $Nesosydne\ chambersi$ , 2.  $Nesosydne\ raillardiae$ , 3.  $Nesosydne\ bridwelli$ , 4.  $Trioza\ HB$ , 5.  $Trioza\ HC$ , 6.  $Drosophila\ sproati$ , 7.  $Laupala\ cerasina$ , 8.  $Tetragnatha\ anuenue$ , 9.  $Tetragnatha\ brevignatha$ , 10.  $Tetragnatha\ quasimodo$ , 11.  $Theridion\ grallator$ .

**Figure 3.** Patterns in degree distributions across sites and different biogeographic classifications of taxa. Top panels show that networks deviate most from MaxEnt on youngest and 671 oldest sites. Inset figures show the distribution of simulated mean squared errors; if the red 672 line falls within the gray region (95% confidence interval) the data conform to maximum en-673 tropy; thus the observed minor deviation on Kohala is not different than expected by chance. 674 Kohala shows minimal modularity, and maximal connectance. The bottom panel shows the 675 number of links for island endemics versus island cosmopolitans. Endemics show lower link-676 age overall, but significantly increase on the middle aged site Maui (highlighted with dotted 677 box). Kohala shows increased linkage overall (highlighted with solid box). When looking 678 at links to plant genera this pattern holds except that endemics on Maui no long show a 679 difference in generality, indicating that the pattern is driven in part by plant diversity.

Figure 4. Trends in network metric nestedness and modularity through time. Nestedness decreases while modularity increases. Error bars come from a null model simulation.
While the sign of the z-score depends on null model and method of calculating modules
(see supplemental figure) the overall trend is robust. Some level of nestedness is likely a
statistical property of these networks, however it could also be driven by stabilizing ecological mechanisms. Modularity is thought to be a sign of coevolution driving convergence in
traits of plants and herbivores. Note the very interesting peaks on Maui where adaptive
diversification may be at its maximum.

# Figures

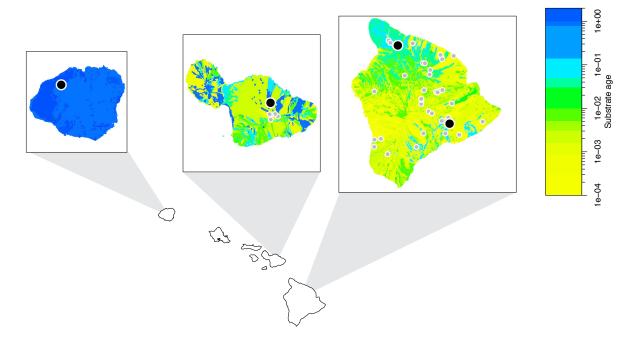


Figure 1:

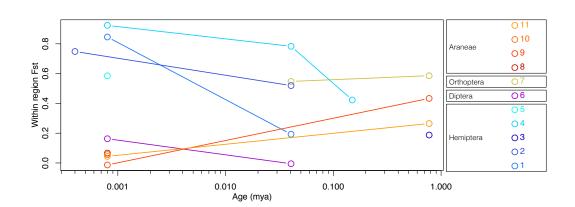


Figure 2:

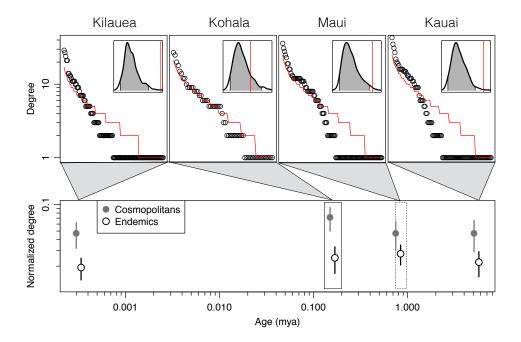


Figure 3:

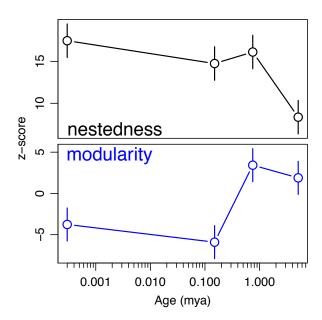


Figure 4: