



RESEARCH
PAPER

Diversity regulation at macro-scales: species richness on oceanic archipelagos

Kostas A. Triantis^{1,2,3*†}, Evan P. Economo^{4,5*†}, François Guilhaumon^{3,6} and Robert E. Ricklefs⁷

¹Department of Ecology and Taxonomy, Faculty of Biology, National and Kapodistrian University, Athens, Greece, ²Biodiversity Research Group, Oxford University Centre for the Environment, Oxford, UK, ³CE3C – Centre for Ecology, Evolution and Environmental Changes/Azorean Biodiversity Group and Universidade dos Açores – Angra do Heroísmo, Açores, Portugal, ⁴Department of Ecology and Evolutionary Biology, University of Michigan, Ann Arbor, MI, USA, ⁵Okinawa Institute of Science and Technology, Onna, Okinawa, Japan, ⁶IRD, MARBEC, University of Montpellier, Montpellier, France, ⁷Department of Biology, University of Missouri, St Louis, MO, USA

*Correspondence: Kostas A. Triantis, Department of Ecology and Taxonomy, Faculty of Biology, National and Kapodistrian University, Athens GR-15784, Greece.
E-mail: ktriantis@biol.uoa.gr; island.biogeography@gmail.com
Evan P. Economo, Okinawa Institute of Science and Technology, 1919-1 Tancha, Onna, Okinawa 904-0495, Japan.
E-mail: evaneconomo@gmail.com
†These authors contributed equally to this work.

ABSTRACT

Aim Understanding the mechanisms that generate diversity patterns requires analyses at spatial and temporal scales that are appropriate to the dispersal capacities and ecological requirements of organisms. Oceanic archipelagos provide a range of island sizes and configurations which should predictably influence colonization, diversification and extinction. To explore the influence of these factors on archipelagic diversity, we relate the numbers of native and endemic species of vascular plants, birds, land snails and spiders – taxa having different dispersal capabilities and population densities – to the number and sizes of islands in the major oceanic archipelagos of the globe.

Location Fourteen major oceanic archipelagos of the globe.

Methods Species richness was collated for native and endemic species in each archipelago. We used linear mixed effect models to quantify the influence on diversity of total area, number of islands, isolation and latitude. We then applied process-based modelling in a Bayesian framework to evaluate how speciation, colonization and extinction are influenced by characteristics of archipelagos associated with species richness, i.e. area, isolation and number of islands.

Results We found parallel scaling of species richness among taxa with respect to total area and number of islands across groups. The process-based model supported effects of isolation on colonization and of area and number of islands on extinction rates, with the scaling exponents mostly similar across taxa. Data are consistent with a range of scaling exponents for speciation rate, implying that those relationships are difficult to infer from the data used.

Conclusions We demonstrate an unexpected parallel scaling of species richness of four taxa with area and number of islands for the major oceanic archipelagos of the globe. We infer that this scaling arises through similar effects of the physical characteristics of archipelagos on extinction, colonization and speciation rates across these disparate taxa, indicating that similar mechanisms have created variation in diversity.

Keywords

Anagenesis, Bayesian inference, colonization, cladogenesis, extinction, linear mixed effect models, macroecology, scale, speciation, species–area relationship.

INTRODUCTION

Although theories in biogeography and macroecology have often emerged from observations of repeated patterns in nature (MacArthur, 1972), patterns themselves rarely provide substantial insight into underlying processes (Rahbek, 2005; Ricklefs,

2007). Indeed, a general consensus about the causes of global patterns of geographic variation in the number of species has yet to develop (Ricklefs, 2004).

The interpretation of large-scale species diversity patterns is dominated by a fundamental conceptual dichotomy, which contrasts (1) environmental limits to regional diversification and

the number of locally coexisting species with (2) the influence of regional geography and history independent of local environmental conditions (Hutchinson, 1959; MacArthur, 1972; Ricklefs, 1987, 2007; Rosenzweig, 1995; Rabosky, 2009; Rabosky & Glor, 2010; Cornell, 2013). As multiscale palaeontological, biogeographic, genetic and ecological data have become more readily available, it is increasingly evident that these alternatives are not mutually exclusive. Large-scale patterns of diversity represent a balance between ecological and historical processes, and, although influenced by ecological conditions, the regional relationship between diversity and ecological space is not constrained to a particular environmentally determined relationship (see Brown *et al.*, 2001; Ricklefs, 2004; Rahbek, 2005; Phillimore & Price, 2008; Rabosky & Glor, 2010; Kisel *et al.*, 2011; Cornell, 2013; Rabosky, 2013; Economo *et al.*, 2015). A key challenge to understanding the mechanisms that create diversity patterns is to frame analyses at the appropriate spatial and temporal scales.

Biotic provinces are self-contained areas within which diversity primarily reflects a balance between speciation and extinction, with colonization from the outside having a minor role. Accordingly, biotic provinces are appropriate spatial and evolutionary units for interpreting the causes of large-scale diversity patterns (Rosenzweig, 1995). Oceanic archipelagos are particularly useful biotic provinces for the analysis of diversity (Triantis *et al.*, 2008; Cabral *et al.*, 2014; Patiño *et al.*, 2014; Valente *et al.*, 2014). Being of volcanic origin and formed over oceanic plates, they have never been connected to continental landmasses and are among the most isolated biological systems of the globe. Moreover, most oceanic islands present a typical developmental sequence from youth, to maturity, old age and eventual disappearance under the ocean surface, giving oceanic archipelagos a similar temporal dynamic (Whittaker & Fernández-Palacios, 2007; Whittaker *et al.*, 2008; Cameron *et al.*, 2013) (Fig. S1 in Appendix S1 in Supporting Information). Finally, these archipelagos exhibit high endemism for most taxa, even the most mobile organisms, arising from a typically small number of colonizing species. For example, of the 29 species of land birds of the Galápagos, 83% are endemic to the islands; all seven bird species of Tristan da Cunha are archipelagic endemics. For less mobile taxa, such as land snails, endemism is high across all the major oceanic archipelagos, reaching up to 99% in Hawaii (e.g. Gillespie & Clague, 2009).

Here we used a two-stage approach to investigate the diversity of indigenous, native (indigenous but not endemic) and endemic species of vascular plants, birds, land snails and spiders of the 14 major volcanic archipelagos of the globe (Table 1 & Table S1 in Appendix S2). These four taxa differ greatly in their ecological requirements, dispersal abilities and typical population sizes, yet they occur on all oceanic archipelagos and are therefore good candidates for identifying commonalities and differences across spatial scales of biological organization (Whittaker & Fernández-Palacios, 2007; Gillespie & Clague, 2009).

First, we used linear mixed effect models in a regression framework to quantify the influence on diversity of potentially

important macroecological drivers: total area, number of islands, isolation, geological age and latitude. At the archipelagic scale, how total area is partitioned among individual islands and how the islands are arranged spatially is likely to affect speciation rates, extinction rates and, ultimately, species richness (e.g. Kisel *et al.*, 2011; Cabral *et al.*, 2014). Second, to disentangle ecological and evolutionary mechanisms underlying patterns identified during the first descriptive step, we developed a process-based modelling framework in which species richness is a steady-state outcome of underlying processes: anagenetic and cladogenetic speciation, extinction and colonization. Our understanding of biodiversity has been advanced recently by the description, parameterization and empirical evaluation of biodiversity dynamics models that are built from simple assumptions (e.g. Gravel *et al.*, 2011; Rosindell & Phillimore, 2011; Valente *et al.*, 2014). In our framework, features of archipelagos, such as area and number of islands, affect the steady-state endemic and native richness through their influence on underlying processes. We took a Bayesian approach to inferring whether characteristics of archipelagos, as well as those of different taxa, influence richness patterns through their effects on underlying rates of colonization, extinction and speciation. We demonstrate an unexpected parallel scaling of species richness with area for birds, spiders, land snails and plants, regardless of variation in their ecology and propensity for colonization and dispersal within archipelagos. Our results suggest that this parallel scaling arises as a result of the similar effects of physical archipelagic characteristics on biogeographic rates across disparate taxa.

METHODS

Archipelagos and physical attributes

We examined 14 oceanic archipelagos distributed across the globe (Table 1 & Table S1 in Appendix S2). The four taxa considered – land birds, vascular plants, spiders and land snails – are present on all the main oceanic archipelagos but have distinct ecological requirements and features. For each archipelago we recorded (1) latitude, (2) isolation (distance to the closest source area for colonists; see Rosindell & Phillimore, 2011), (3) maximum elevation, (4) maximum geological age for each of the archipelagos, defined as the age of the oldest island among the currently existing islands, (5) land area of the archipelago, and (6) number of islands constituting the archipelago. The configuration of islands varies greatly through time; sea level minima during the Pleistocene produced connections between some adjacent islands, turning them to single islands, while volcanism can both join and sometimes subdivide island areas (Whittaker & Fernández-Palacios, 2007; Carracedo & Tilling, 2003; Grant & Grant, 2008; Figs. S1 & S2 in Appendix S1 and Appendix S2; for further details see Cameron *et al.*, 2013). As a simple description of archipelago dynamics, we considered previously connected islands in each archipelago as single islands (see below and Table S1 in Appendix S2).

The main sources of data were the UNEP Islands Directory (<http://islands.unep.ch/>) and several data compilations (Steadman, 2006; Whittaker & Fernández-Palacios, 2007; Whittaker *et al.*, 2008; Gillespie & Clague, 2009; Rosindell & Phillimore, 2011; Cameron *et al.*, 2013; Weigelt *et al.*, 2013; Cabral *et al.*, 2014; see also Appendix S4). For most of the archipelagos, if not all, the maximum geological age exceeds the age of the current islands. For example, the maximum geological age for the Canary Islands is estimated as 68 Ma, whereas the oldest current island in the archipelago, Fuerteventura, is only 20 Ma (Fig. S1A in Appendix S1). Although some lineages present on the archipelagos have histories that extend beyond the maximum age of the current islands (e.g. Givnish *et al.*, 2009), the maximum age of the existing islands of an archipelago can be considered as a conservative reference point for the time over which biological processes have established modern diversity patterns (Price & Clague, 2002; Whittaker *et al.*, 2008). All the archipelagos have maximum geological ages exceeding *c.* 3 Ma.

Species richness

For land birds, we included only species described as breeding, or probably breeding, on a given archipelago; for plants and land snails, we included all species considered indigenous or probably indigenous. Because species assemblages of these taxa on most oceanic islands have been affected by historic and pre-historic human activities (Olson & James, 1984; Steadman, 2006; Whittaker & Fernández-Palacios, 2007; Gillespie & Clague, 2009), we included both extant species and historically or pre-historically extinct species (Steadman, 2006) whenever such information was available. Olson & James (1984) noted that the pre-human bird faunas of the three main Atlantic archipelagos, i.e. the Azores, Madeira and Canary, are generally unknown (but

see Sánchez-Marco, 2010), and so we also ran the analyses without them. However, excluding these faunas provided similar results (see Results and Appendix S3). Our sources of data on species richness were several published compilations (Steadman, 2006; Whittaker & Fernández-Palacios, 2007; Whittaker *et al.*, 2008; Gillespie & Clague, 2009; Givnish *et al.*, 2009; Sánchez-Marco, 2010; Rosindell & Phillimore, 2011; Cameron *et al.*, 2013; Cabral *et al.*, 2014; see also Appendix S4) and personal contacts with local experts (see Acknowledgements). For spiders, data were available for 12 archipelagos out of the 14 considered for the other three taxa, and only for the archipelagic endemics. Although different sampling and taxonomic efforts were most probably invested for different taxa and archipelagos, the species richness data used here are the most recent available.

Multiple regressions

We used linear mixed effect models (LMMs) in a model selection framework (Burnham & Anderson, 2002) to determine the best models for describing \log_{10} -transformed numbers of indigenous, native (indigenous without endemics) and endemic species (12 archipelagos for spiders and 14 for the rest). We selected an initial set of five explanatory variables after investigating multicollinearity using Pearson correlations: latitude, number of islands, isolation, geological age and archipelagic land area (elevation was excluded from these analyses because it is strongly collinear with area, $r = 0.743$, $n = 14$). The continuous descriptor variables, apart from latitude, were \log_{10} -transformed. We evaluated taxonomic differences in regression slopes for the relationship between species richness and the above-mentioned variables by investigating the selection of taxon \times variable interactions. By considering all the taxa in a single analysis, we created replicated data for each archipelago. This pseudoreplication was accounted

Archipelago	Indigenous			Archipelagic endemics			
	Plants	Birds	Snails	Plants	Birds	Snails	Spiders*
Azores	230	23	104	86	1	74	24
Canary	1254	78	260	585	10	232	300
Comoros	937	47	146	500	16	100	45
Galápagos	549	29	100	236	24	94	88
Gulf of Guinea	903	49	110	107	34	83	92
Hawaii	1166	102	757	1024	98	752	162
Juan Fernández	207	12	40	121	6	20	28
Madeira	754	30	188	150	4	168	58
Marquesas	331	22	99	161	11	75	27
Mascarene	1224	60	200	749	51	180	–
Northern Marianas	180	5	30	21	0	12	3
Samoa	765	32	94	252	9	60	47
Society	575	26	161	250	16	136	–
Tristan da Cunha	91	7	15	45	7	15	4

Table 1 The number of indigenous and archipelagic species for plants, land birds, land snails and spiders for the archipelagos considered.

*For the archipelagic endemic spider species, data were available for 12 out of the 14 archipelagos considered.

for by including a random intercept for archipelagos in the LMMs (see Bunnefeld & Phillimore, 2012). The best combinations of fixed effects were selected using maximum likelihood (ML) methods and model selection based on the corrected Akaike information criterion (AICc). We used the dredge function in the MuMIn package in R (version 0.13.17) to run a complete set of models with all possible combinations of the fixed effects and to identify the set of 'best models' according to the accepted criterion for different AICc values: $\Delta\text{AICc} < 2$ (Burnham & Anderson, 2002). Comparable R^2 values (with the same meaning as in simple or multiple linear regression) are not easy to obtain for LMMs (Zuur *et al.*, 2009). We thus used an R^2 measure that compares the deviance of the LMM with the deviance of a linear intercept-only model (Kvålseth, 1985):

$$R^2 = 1 - \frac{\sum (y - \hat{y})^2}{\sum (y - \bar{y})^2}.$$

We used these R^2 values to quantify the proportion of the total variation among archipelagos in species (indigenous, natives and endemics) explained (accounted for) by selected LMMs. If more than one model was selected as 'best' then we estimated for each variable the weight of evidence, i.e. the sum of AICc weights ($w\text{AICc}$) derived from the AICc for the models in which the variable is included, which measures the relative importance of each variable (Burnham & Anderson, 2002; see also Cameron *et al.*, 2013). The variables with a $w\text{AICc} = 1$ are included in all best models. For the number of islands in each archipelago, we ran analyses for current island number and after considering palaeo-connections separately.

Endemism

Similar proportions of endemics among taxa could lead to converging results between indigenous and endemic species. We tested whether the proportion of endemics differs among snails, birds and plants (for spiders, data are available for endemics only). The proportions differ statistically [Kruskal–Wallis test: H (d.f. = 2, $n = 42$) = 12.3, $P = 0.002$]. Removing birds from the three Atlantic archipelagos did not alter this result [Kruskal–Wallis test: H (d.f. = 2, $n = 39$) = 12.8, $P = 0.002$].

Model-based inference of biogeographic rate scaling

After determining which predictor variables correlate with species richness, we performed additional analyses to investigate which biogeographic rates, i.e. cladogenesis, anagenesis, extinction and colonization, are likely to be affected by those variables. Due to the complexities related to island age, and to simplify the analysis, we retained only the three biogeographic variables with the maximum importance values in the multiple regressions: area, isolation and number of islands (see Results, Table 2). We sought to determine whether each variable influenced species numbers through its effects on the scaling of colonization, extinction or speciation rates. Furthermore, we asked whether similarities in the scaling of species richness with predictor

variables across taxonomic groups were reflected by similarities in the scaling of biogeographic rates.

We focused on the numbers of archipelagic endemic species and the number of native species (i.e. indigenous but not archipelagic endemic) as our response variables. The numbers of archipelagic endemics and natives can change through the following events (Fig. 1): (1) colonization from the outside (adding one non-endemic species); (2) anagenetic speciation (subtracting one non-endemic and adding one endemic species); (3) cladogenetic speciation (one endemic becomes two endemics); and (4) extinction (subtracting a species from either category). For our main analysis, we assumed that non-endemic species cannot undergo cladogenetic speciation. However, we also considered the alternative assumption, that non-endemics could undergo cladogenetic speciation, and compared the results (see Fig. S5 in Appendix S1). Because we focus at the archipelago level, dispersal from one island to the other does not affect archipelagic species richness directly, and so we did not consider intra-archipelago dispersal in our model.

We assumed that colonization, speciation and extinction events are Poissonian in nature, an assumption that is commonly used to model biogeographic (MacArthur & Wilson, 1967; Chen & He, 2009) and metapopulation dynamics (e.g. Hanski, 1999), and their respective rates are modelled as functions of predictor variables. The expected rate of change of native species (x) and endemic species (y) can be written as:

$$\begin{aligned} \frac{dx}{dt} &= r_c - r_g x - r_e x \\ \frac{dy}{dt} &= r_g x + r_s y - r_e y \end{aligned}$$

where r_c is the rate of colonization, r_g is the per-species rate of anagenetic speciation, r_s is the per-species rate of cladogenetic speciation and r_e is the per-species rate of extinction. These rates may vary with island characteristics, such as area and isolation. They may also change dynamically as richness changes; for example, extinction rates may increase as species richness increases. In general, the form of the equations relating the Poisson rates to the predictor variables is similar to a Poisson regression in a generalized linear model framework, where the Poisson rate parameter is a power function of the predictor variables. To avoid overfitting, we sought to minimize the parameter number by allowing a predictor variable to affect a given rate only when supported by theory or empirical observation; we restricted the maximum number of predictor variables for each process to two. We assumed that colonization rate is affected by area and isolation, that cladogenetic speciation and extinction rates are affected by area, number of islands and species richness (e.g. MacArthur & Wilson, 1967; Phillimore & Price, 2008; Whittaker *et al.*, 2008; Chen & He, 2009; Kisel *et al.*, 2011; Rosindell & Phillimore, 2011; Valente *et al.*, 2014), but that anagenetic speciation rate is not affected by the predictor variables (e.g. Stuessy *et al.*, 2006), although it can vary across taxa. Our rates were given by:

Table 2 Most parsimonious linear mixed effect models for indigenous, native, and endemic species richness.

	Indigenous				Natives				Endemics				
	Model 1	Model 2	Model 3	Variable importance	Model 1	Model 2	Model 3	Model 4	Variable importance	Model 1	Model 2	Model 3	Variable importance
Intercept plants	1.57 (0.16)	1.36 (0.15)	1.76 (0.26)	1	5.13 (0.58)	4.43 (0.48)	3.865 (0.57)	4.59 (0.72)	1	0.59 (0.23)	0.77 (0.28)	0.63 (0.54)	1
Contrast birds	-1.24 (0.04)	-1.238 (0.04)	-1.24 (0.04)	1	-1.36 (0.08)	-1.36 (0.08)	-1.36 (0.08)	-1.36 (0.08)	1	-1.21 (0.09)	-1.214 (0.09)	-1.88 (0.56)	1
Contrast snails	-0.66 (0.04)	-0.66 (0.04)	-0.66 (0.04)	1	-1.19 (0.08)	-1.19 (0.08)	-1.19 (0.08)	-1.19 (0.08)	1	-0.36 (0.09)	-0.36 (0.09)	-0.59 (0.56)	1
Contrast spiders	-	-	-	-	-	-	-	-	-	-0.63 (0.09)	-0.63 (0.09)	0.55 (0.56)	1
Area total	0.57 (0.05)	0.60 (0.05)	0.57 (0.05)	1	-	-	0.147 (0.09)	0.11 (0.09)	0.37	0.73 (0.08)	0.71 (0.08)	0.73 (0.08)	1
No. of islands	-0.91 (0.15)	-0.85 (0.17)	-0.75 (0.16)	1	-	-	-	-	0	-0.97 (0.26)	-1.02 (0.25)	-0.94 (0.27)	1
Isolation	-	-	-0.13 (0.07)	0.27	-0.81 (0.16)	-0.69 (0.16)	-0.66 (0.15)	-0.76 (0.16)	1	-	-	-0.01 (0.17)*	0.22
Age	-	-	-	0	-0.39 (0.22)	-	-	-0.32 (0.21)	0.461	-	-	-	-
Latitude	-0.01 (0.002)	-	-	0.44	-	-	-	-	0	-	-0.004 (0.004)	-	0.23
R ²	0.97	0.97	0.97		0.94	0.944	0.943	0.94		0.86	0.86	0.90	
No. parameters	8	7	8		7	6	7	8		8	9	12	
ΔAICc	0	0.83	0.94		0	0.09	0.67	1.68		0	1.72	1.80	
wAICc	0.44	0.29	0.27		0.32	0.309	0.230	0.140		0.547	0.23	0.22	

The most parsimonious are all models (e.g. Model 1, 2, etc.) with an AICc difference (ΔAICc) less than 2 and are the ones presented (see also Materials and Methods). Effect sizes (estimates of regression coefficients) are presented along with their standard errors (in parentheses). For the levels of the factor 'taxon', effect sizes for birds, snails and spiders are shown in the form of differences (contrasts) with respect to the absolute value for plants. The R² value, the number of parameters in the model, the AICc difference and the AICc weight (wAICc) are given for each model. The importance of each variable, i.e. the sum of wAICc of models including them, is also given. Results for birds, snails and plants are for 14 archipelagos and for endemic spiders for 12. Results for birds are with the three Atlantic archipelagos, i.e. the Azores, Madeira and the Canary Islands, but their exclusion gives identical results (see Table S3 in Appendix S3). See also Table S4 in Appendix S3 for the same analysis but with the extant number of islands for each archipelago instead of paleo-connections.

*Indicates taxon-variable interactions.

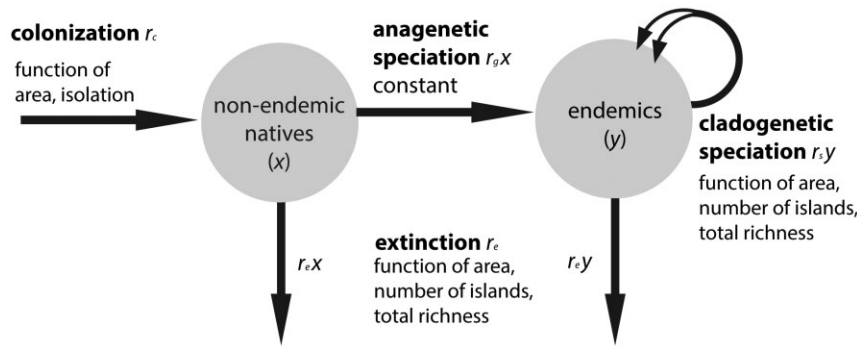


Figure 1 Conceptual diagram showing the basic rates controlling endemic richness and non-endemic native richness. In the model-based analysis, the rates (arrows) are allowed to be functions of predictor variables recovered as significant in the regression analysis.

$$r_c = c_0 A^{c_1} I^{c_2}$$

$$r_e = e_0 A^{e_1} N^{e_2} (x + y)^{e_3}$$

$$r_g = g_0$$

$$r_s = s_0 A^{s_1} N^{s_2} (x + y)^{s_3}$$

where A is area, I is isolation, N is number of islands and c_0, g_0, s_0 and e_0 are taxon-specific constants.

An exponent of zero corresponds to no effect of the predictor variable, > 0 and < 1 to sublinear scaling, 1 to linear scaling, and > 1 to superlinear scaling. The $(x + y)$ term allows per-species speciation and extinction rates potentially to vary with total richness on the archipelago (i.e. diversity can be ‘regulated’). To facilitate our analysis, we assumed that archipelagic numbers of native and endemic species are at a steady state ($dx/dt = dy/dt = 0$). This can be enforced by assuming that the per-species extinction rate increases with total richness, and per-species cladogenesis decreases with increasing richness ($e_3 > 0, s_3 < 0$). The steady state is reached when inputs from colonization plus cladogenetic speciation balance the losses due to extinction, i.e. equilibrium (MacArthur & Wilson, 1967; but see Heaney, 2000). The possibility that per-species extinction rates might increase as richness increases is a commonly held view going back to original work on the theory of island biogeography (e.g. MacArthur & Wilson, 1967), and the diversity dependence of speciation rates has been supported by numerous analyses consistent with diversity dependence of speciation in phylogenetic studies (e.g. Phillimore & Price, 2008; see also Rabosky, 2009, 2013; Valente *et al.*, 2014). Even so, the strength of the regulation of both speciation and extinction rates can freely vary in the analysis and, if justified by the data, one rate might show strong regulation while the other does not.

We used the deterministic rate of change equations to find the expected steady-state richness values, and assumed that deviations from the steady state of the log-transformed species numbers are Gaussian. This is similar to the error model used in most regression analyses of species–area curves. An alternative would be to allow the biogeographic rate parameters to determine the fluctuations around the steady state in a fully stochastic model (e.g. as a continuous-time Markov chain). We ultimately decided against the latter for reasons of computational effi-

ciency, and furthermore it is not clear that all the deviations from expected values should be endogenous and linked to the rate parameters (i.e. exogenous factors such as biogeographic history or variation in sampling and taxonomic treatment could account for deviations from expected richness).

To calculate the likelihood of a parameter set given the data, we solved the rate-of-change equations at equilibrium:

$$0 = c_0 A^{c_1} I^{c_2} - g_0 \hat{x} - \epsilon_0 A^{e_1} N^{e_2} (\hat{x} + \hat{y})^{e_3} \hat{x}$$

$$0 = g_0 \hat{x} + s_0 A^{s_1} N^{s_2} (\hat{x} + \hat{y})^{s_3} \hat{y} - \epsilon_0 A^{e_1} N^{e_2} (\hat{x} + \hat{y})^{e_3} \hat{y}$$

for the steady-state richness values, \hat{x} and \hat{y} . The likelihood of a given parameter set, given vectors of observed species numbers (\mathbf{x}, \mathbf{y}) , is then:

$$\mathcal{L}(\theta | \mathbf{x}, \mathbf{y}) = \prod_{i,j} \left[\frac{1}{\sigma_j \sqrt{2\pi}} \exp\left(-\frac{(\ln(x_{\text{obs},ij}) - \ln(\hat{x}_{ij}))^2}{2\sigma_j^2}\right) \right]$$

$$\left[\frac{1}{\sigma_j \sqrt{2\pi}} \exp\left(-\frac{(\ln(y_{\text{obs},ij}) - \ln(\hat{y}_{ij}))^2}{2\sigma_j^2}\right) \right]$$

where i indexes over all islands, j indexes over taxonomic groups and sigma (σ) is the error term, or the standard deviation of the x and y values from the expectation. We fitted the model to the data in two stages. In the first, the datasets for the three different taxonomic groups were analysed separately (spiders were not included since data were only available for endemic species), and all parameters were optimized for each taxonomic group. In the second, we fitted the combined dataset; scaling exponents were fixed across taxonomic groups, but different constants (i.e. intercepts) were fitted for each taxonomic group. If these individual rates varied across taxonomic groups, and/or scaled with island characteristics such as area, isolation and number of islands, this would drive the overall scaling pattern of native and endemic richness we observe among islands.

Bayesian parameter inference

We took a Bayesian approach to characterize the range of parameters that fit the data well. Given the limitations of the size

of the dataset relative to the number of parameters in even a simple process-based island biogeography model, we aimed to assess the identifiability of the model parameters and evaluate which rates can be constrained by the data and which cannot. Species richness scaling might be more sensitive to colonization scaling than to speciation scaling, for example, and thus a particular richness scaling can constrain only those parameters. One cannot estimate the scaling of all rates with all potential predictor variables without additional data types (e.g. phylogenies for all taxa on all archipelagos) that are not available at the present time. A main challenge in this kind of analysis is overfitting: the number of parameters explodes when all effects are allowed to vary. We constrained intercepts to vary between 10^{-15} and 10^{10} and exponents to vary between -2 and 2 . The only exception was the scaling of speciation with species richness, which is zero or negative ($-2, 0$) and the scaling of extinction with species richness, which is zero or positive ($0, 2$). The prior is a non-informative uniform distribution over the parameter space.

We used a Markov chain Monte Carlo (MCMC) algorithm to sample parameter combinations from the posterior. We used the 'mcmcrun()' function in the MCMC Toolbox in Matlab (Haario *et al.*, 2006) to implement the sampling (code available from EPE). For each analysis, we ran 10 independent MCMC runs, and combined the post-burn-in posterior samples of all runs to give the final posterior. Burn-in was evaluated for each run individually by examining the stationarity of model parameters and consistency of sampled parameter estimates across runs. Overall convergence was assessed using consistency of posterior distributions between runs.

RESULTS

Multiple regressions – linear mixed effect models

According to the AICc-based model selection procedure, more than a single parsimonious model was selected for indigenous, native and endemic species. Considering only the variables with $wAICc = 1$, i.e. the variables included in all the best models and thus having maximum relative importance (i.e. empirical support; Burnham & Anderson, 2002), we selected total area (positive effect) and number of islands (negative effect) for indigenous and endemic species. For native species, isolation has a negative effect (Table 2). Latitude, isolation and the geological age of the archipelagos had lower, or no, importance across models (Table 2). The models explained most of the variation in species richness for the three distributional categories ($R^2 \geq 0.92$ for all cases). Total area and number of islands were not significantly correlated ($P > 0.10$).

No taxon \times variable interactions were retained in the most parsimonious models, apart for isolation in a single case (see Table 2), indicating indistinguishable slopes across taxa for the relationships between species richness, total area and number of islands for indigenous and endemic species. All the best models for indigenous, native and endemic species

richness revealed significantly different intercepts for all taxa (Table 2).

The indistinguishable slopes indicate that indigenous species richness scales to the same power with total archipelagic island area for three different taxa across archipelagos scattered across the globe (Fig. 2). The same is true for endemic species, and also regardless of the number of islands considered, i.e. taking into account palaeo-connections or using number of extant islands (Appendix S3). The slopes of the species–total area relationship in logarithmic space were 0.57 ± 0.04 SE for indigenous species and 0.72 ± 0.06 SE for endemic species, both of which are within the range of values proposed for the inter-provincial species–area relationship for continental landmasses (Rosenzweig, 1995) and higher than the typical slopes for the classic species–area relationship across individual islands (Rosenzweig, 1995; Ricklefs & Bermingham, 2007; Triantis *et al.*, 2012). In contrast to the homogeneous slopes, the intercepts of the indigenous and endemic species–area relationships differed considerably among the taxa (Fig. 2, Table 2). Considering extant islands for each archipelago (Table S1 in appendix S1), without taking into account past connections of islands, gave similar results (Table S4 in Appendix S3).

To test the absence of slope differences among taxa for the relationships between species richness and total area and number of islands, we also used average area, which summarizes the positive effect of total area and the negative effect of the number of islands. The results remained unchanged (Fig. S4 in Appendix S1, Table S5 in Appendix S3).

Model-based inference

In the independent models for each taxon, the posterior distributions generally agreed across taxonomic groups and with the model combining all taxa (Fig. 3), with exponents of similar sign and magnitude (or similar levels of uncertainty across the parameter space). To determine that the models produced reasonable results, we compared the model predictions from 1000 randomly chosen parameter sets from the posterior distributions with the empirical values for native and endemic richness, and they agreed well (Fig. S6 in Appendix S1) and also matched the empirical exponents of the richness–area scaling (Fig. S7 in Appendix S1).

The combined model was generally the most informative with regard to conclusions about colonization and extinction, with the speciation parameters being less identifiable (Fig. 3). Colonization exhibits a negative relationship with isolation for all taxonomic groups. The extinction rate shows a negative relationship with area and a positive scaling with number of islands. Moreover, results for extinction provide support for diversity dependence, with extinction rates showing a positive relationship with species richness. The posterior distribution for *in situ* cladogenetic speciation rate, for which we did not consider isolation, is generally flat, indicating that the data are consistent with a range of cladogenetic speciation scalings and the parameters are not identifiable with these data, with the exception that for the combined data there seems to be (weak) evidence that

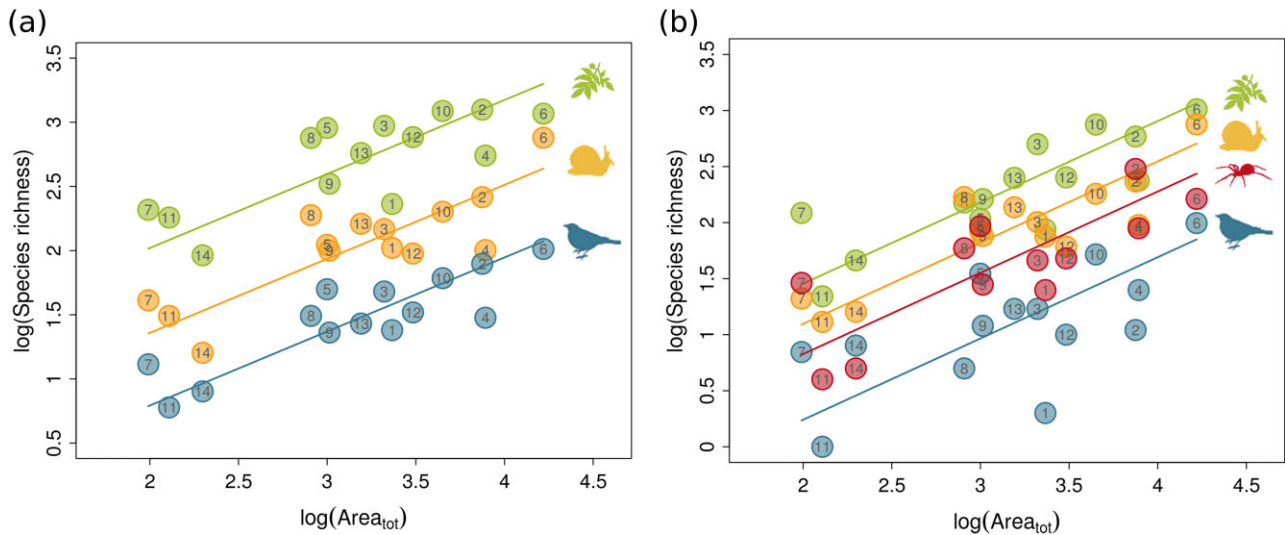


Figure 2 Individual relationships between archipelagic total area and taxon richness. (a) Indigenous and (b) archipelagic endemic species for land birds (🐦), vascular plants (🌿), land snails (🐌) and spiders (🕸). The lines are the predictions from mixed effect models (Table 2), the effects of variables other than total area and interactions are included, using the mean value across archipelagos and wAICc-averaged coefficients. The slopes for each taxon are parallel (see Table 2). For birds we include the three main Atlantic archipelagos, i.e. the Azores, Madeira and the Canary Islands; however their exclusion provides similar results (Table S3, Fig. S3). Archipelagos are: (1) Comoros, (2) Galápagos, (3) Hawaii, (4) Juan Fernández, (5) Mariana Northern, (6) Marquesas, (7) Mascarenes, (8) Samoa, (9) Gulf of Guinea Islands, (10) Society Islands, (11) Tristan da Cunha, (12) Azores, (13) Canary Islands, (14) Madeira.

the speciation rate scales positively with area. To test that our conclusions about colonization and extinction are not affected by uncertainty in the speciation parameters, we performed an additional analysis fixing a positive relationship of speciation

with area, and found that the other inferred relationships were stable even given this constraint (see Fig. S8 in Appendix S1). The alternative model, in which endemics and non-endemics can undergo cladogenetic speciation, produced similar results

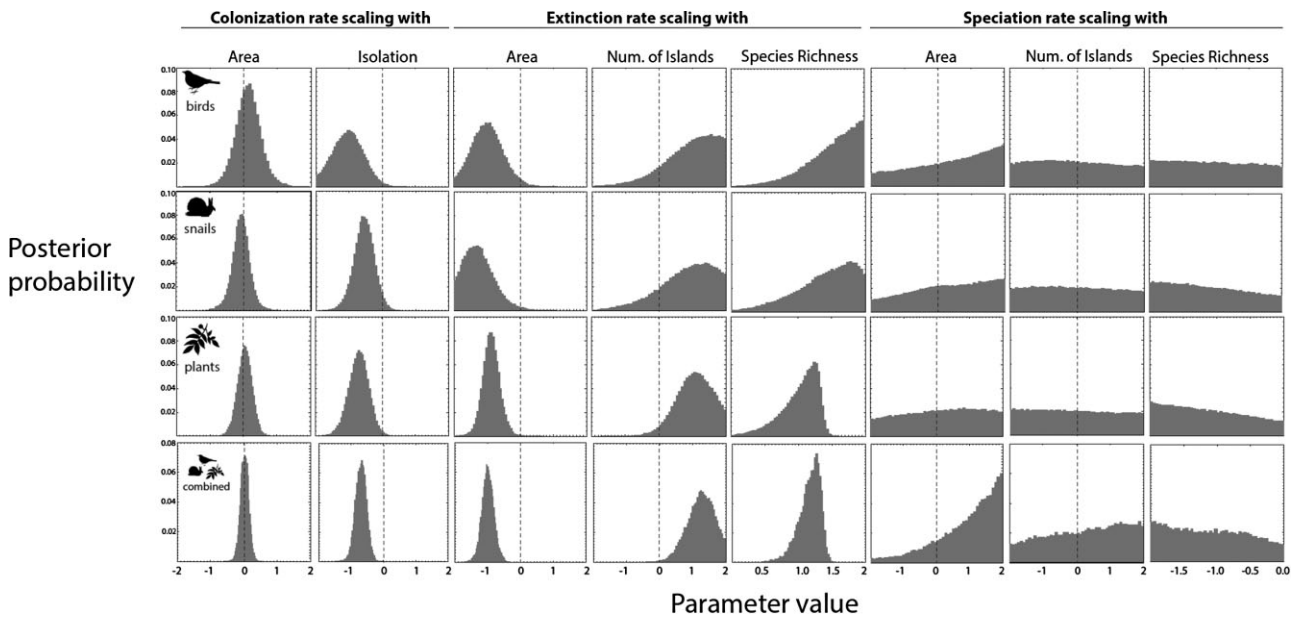


Figure 3 Sampled posterior distributions of scaling exponent parameters of different biogeographic rates and predictor variables. These are the posterior distributions of each parameter across the whole parameter space, not at a fixed values of other parameters. Each parameter reflects the scaling relationship between one rate (e.g. extinction) and one predictor variable (e.g. area). The top three rows reflect models fitting of each taxonomic group individually. The bottom row assumes common exponent parameters across taxonomic groups, but allows group-specific intercepts.

(see Fig. S5 in Appendix S1), with slightly more evidence in the combined model that speciation rates are also dependent on the number of islands and area.

DISCUSSION

The strikingly high proportion of variation in species richness that is explained by area and number of islands and the consistent scaling of species richness with area across taxa are puzzling. Within isolated archipelagos, diversity is mainly regulated by loss of species and the production of new species. These processes depend on the dynamic nature of oceanic archipelagos, expressed by the continuous change in number and size of islands, as well as attributes of organisms, including body size, generation time and dispersal ability (Ricklefs & Bermingham, 2007; Whittaker *et al.*, 2008; Pigolotti & Cencini, 2009; Kisel & Barraclough, 2010). Colonization from outside typically plays a secondary role (e.g. Whittaker & Fernández-Palacios, 2007). Moreover, within- and between-island mechanisms of speciation generally operate only above certain thresholds of island size and archipelago configuration, which differ among taxa according to dispersal abilities and which typically produce change in the slope of the species–area relationship (Rosenzweig, 1995; Losos & Schluter, 2000; Triantis *et al.*, 2008). Accordingly, we expected the relationships between species richness and area of the four taxa to have different slopes owing to their disparate ecological requirements, dispersal abilities, typical population sizes, average spatial extent of intraspecific gene flow and propensity to form new species (e.g. related to generation time; see Patiño *et al.*, (2014) for single islands). However, even when the effects of colonization from outside archipelagos are removed by considering only endemic species, the slope of the species–area relationship increases, but remains homogeneous, for all the taxa (Fig. 2, Table 2). Additionally, the varying degrees of completeness of the archipelagos' faunas and the fossil records (e.g. Cameron *et al.*, 2013), and the possible inconsistencies among taxonomists working across widely scattered oceanic archipelagos, add to the underlying stochasticity and thus further support the robustness of our findings.

Emergent ecological correlates of species diversity might result from different combinations of underlying evolutionary mechanisms (Rahbek, 2005). For example, the positive scaling of species richness with area could reflect several independent mechanisms, including a decrease in extinction rate and increases in speciation and colonization rates with area; but more certainly from several of these mechanisms, possibly acting with different strengths depending on spatial scale and/or a taxon's life-history traits (Rosenzweig, 1995; Triantis *et al.*, 2012). Our process-based model provided support for empirically and theoretically established island patterns: colonization is strongly linked to isolation; cladogenetic speciation and extinction are both correlated with area (MacArthur & Wilson, 1967; Rosenzweig, 1995; Whittaker & Fernández-Palacios, 2007; Pigolotti & Cencini, 2009; Kisel *et al.*, 2011; Rosindell & Phillimore, 2011; Cabral *et al.*, 2014). The model also

highlighted the strong dependency of extinction rate on archipelagic parameters, while dependences are less clear for speciation rate. While higher levels of archipelagic fragmentation might elevate diversification and allow more species to coexist within the same total area, LMMs showed a negative relationship of species richness with the number of islands (Table 2). Our process-based model provides support for the notion that fragmentation, at this scale, reduces the area of the component islands below a size that can maintain viable populations, thereby increasing extinction rates (e.g. Kisel *et al.*, 2011), without an obvious negative effect on the generation of archipelagic endemics (Losos & Schluter, 2000). In general, however, speciation scaling may be difficult to infer with the current data, and so incorporation of phylogenetic information would greatly assist in inferring the drivers of speciation rates.

Obvious differences in the ecology of the taxa considered here suggest that different contributions of colonization, extinction and speciation (within-island and within-archipelago) to species richness for any particular archipelago were to be expected for each taxon, as indicated by the varying levels of endemism (see Results and Table 1). Highly mobile organisms such as birds do not undergo species formation on small islands, although they can form new species on archipelagos of small islands (e.g. Grant & Grant, 2008). For such taxa, the number of endemic species in an archipelago is rarely the sum of single-island endemic species, in contrast to highly sedentary species such as snails, that disperse infrequently between islands. According to our processed-based model, the posterior distributions of scaling exponents for the different biogeographic rates, i.e. colonization, extinction and speciation, across the three different taxonomic groups were mostly similar (Fig. 3). For example, the scaling exponent of extinction with area is similar for the three taxa considered, and so extinction presented the same relationship to area regardless of taxon, i.e. negative. Hence, despite the varying contribution of each of these processes to the species richness of each taxon, the processes exhibit similar scaling relationships with isolation, area and number of islands. Similar scaling of the underlying processes that influence species richness provides a strong explanation for the parallel scaling of species richness itself (see Model-based inference in Appendix S3 and Figs S6 & S7 in Appendix S1).

The homogeneous slopes for indigenous and endemic species imply that speciation and extinction rates for each taxon respond similarly to changes in archipelagic area. Differences between the taxa in the intercepts of the LMMs suggest differences in the capacity of island archipelagos to support species ecologically at unit area (Triantis *et al.*, 2012), and these intercepts match the likely ecological space required by species of each of the taxa (i.e. plants < snails < spiders < birds) (Öckinger *et al.*, 2010; Triantis *et al.*, 2012; Fig. 2). Clearly, the results of our model are limited by the selection of the predictor variables and the sample size (although minimized by the use of LMMs, cf. Cabral *et al.*, 2014; Patiño *et al.*, 2014). The results could also be further supported, or challenged, by additional data that were not available in this study, such as the phylogenetic structure of communities, which provide information on the effect of area

and other variables on the persistence times of island populations and rates of lineage splitting (e.g. Cadena *et al.*, 2005). However, even without such information our results agree with existing island theory and provide insight into the striking pattern described herein.

The biotas of each of the oceanic archipelagos are independently derived from different continental faunas and floras, and within-archipelago diversification is independent across archipelagos (Whittaker & Fernández-Palacios, 2007; Gillespie & Clague, 2009). We found that these systems have independently converged on the same relationship between diversity and ecological space, expressed by the total archipelagic area and the way it is apportioned among islands. The generality of the diversity scaling described here suggests that it is rooted in fundamental limits to the processes establishing diversity (Hutchinson, 1959; Ricklefs, 2004; Phillimore & Price, 2008; Rabosky, 2009; Economo *et al.*, 2015). This tightly constrained scaling of diversity, and its underlying processes, can be considered as an emergent property of the endogenous dynamics of oceanic archipelagos, explaining to a large extent the high variation in species richness obtained with the best multiple regression models for indigenous and endemic species (Table 2). The relative roles of evolutionary and ecological processes, as well as the geological histories of archipelagos, in producing consistent diversity–area outcomes for different taxa remain to be determined by additional data, including the phylogenetic structure of island biotas. A possible explanation for a common species–area scaling could be related to the idea of the existence of a (taxon-specific) theoretical maximum species diversity, i.e. carrying capacity (Brown *et al.*, 2001; Whittaker *et al.*, 2008; Rabosky, 2013), which canalizes the filling of available ecological space by diversity-dependent biogeographic processes (Phillimore & Price, 2008; Rabosky, 2013). However, our results presuppose that the taxa disperse well enough to colonize the archipelagos in the first place. Other groups, such as non-flying mammals and lizards, would probably not show similar diversity scaling on the same 14 archipelagos because they have failed to colonize many of the more isolated ones, for example Hawaii, the Azores and Tristan da Cunha (Gillespie & Clague, 2009).

Overall, the modelling framework presented here represents a novel direction in linking biodiversity data to biogeographic rates. However, as with any new inquiry, we should proceed with caution when interpreting results. First, our assumptions could be violated; for example, archipelagos may not be at steady-state richness (e.g. Whittaker & Fernández-Palacios, 2007). A future direction would be to compare a model based on an equilibrium assumption with predictions of a non-equilibrium model in a model selection framework. Second, the current implementation is highly computationally intensive, making the evaluation of its performance across the vast parameter space difficult, and there may be regions of parameter space that confound inference. The development of more computationally efficient, and ideally fully stochastic, implementations would be a promising direction for further work and an improvement upon the analyses presented here.

Our approach could be extended to allow a broader array of assumptions and predictor variables, but additional data and different types of data (e.g. phylogenies) would probably be needed to improve the power of the existing model and fit more complex models. The development of fully stochastic implementations would be a promising direction for further work, building upon the analyses described here. The potential of a process-based model is that it can, in principle, generate predictions for different biological patterns (e.g. richness, phylogenetic structure, population structure) that arise from the same underlying ecological and evolutionary dynamics. Key challenges to such a model, especially for islands, would be to consider the possible effect on speciation and extinction of a species richness carrying capacity of an island (e.g. Whittaker *et al.*, 2008; Valente *et al.*, 2014), untangle the different temporal scales on which colonization and speciation act, and take into consideration that isolated, large archipelagos may never reach an equilibrium but rather their state is described by a dynamic disequilibrium (e.g. Heaney, 2000). In this context, oceanic archipelagos, seen as distinct spatial and evolutionary units, present unique opportunities for synthetic analysis in biogeography and ecology.

ACKNOWLEDGEMENTS

This work was supported a Fundação para a Ciência e a Tecnologia (FCT) Fellowship (SFRH/BPD/44306/2008) to K.A.T. and partially by PTDC/BIA-BIC/119255/2010. F.G. was supported by the 'Range Shift' project (PTDC/AAC-AMB/098163/2008) from FCT, co-financed by the European Social Fund. R.E.R. acknowledges support from the Curators of the University of Missouri. E.P.E. acknowledges the support of NSF (DEB-1145989) and OIST. We thank J. Price, A. Boyer, J. M. Fernández-Palacios, P. A. V. Borges, H. Schaeffer, A. Oikonomou, Ch. Barboutis, S. Sfenthourakis, R. Preece, R. H. Cowie, J. C. Rando, M. Carine, P. Cardoso and A. F. Martins for assistance with the data; Jim Brown, Rosemary Gillespie, Larry Heaney and Michael Borregaard for discussions during the 6th International Biogeography Society Conference; R. J. Ladle, L. Heaney, B. C. Emerson, R. J. Whittaker and, especially, J. Chase and J. Brown for critical comments on previous versions of the manuscript. A. Phillimore, P. Weigelt, two anonymous referees and the editors provided extensive and insightful comments and suggestions.

REFERENCES

- Brown, J.H., Ernest, S.K.M., Parody, J.M. & Haskell, J.P. (2001) Regulation of diversity: maintenance of species richness in changing environments. *Oecologia*, **126**, 321–332.
- Bunnefeld, N. & Phillimore, A.B. (2012) Island, archipelago and taxon effects: mixed models as a means of dealing with the imperfect design of nature's experiments. *Ecography*, **35**, 15–22.
- Burnham, K.P. & Anderson, D.R. (2002) *Model selection and inference: a practical information-theoretic approach*. Springer-Verlag, New York.

- Cabral, J.S., Weigelt, P., Kissling, W.D. & Kreft, H. (2014) Biogeographic, climatic and spatial drivers differentially affect alpha, beta and gamma diversities on oceanic archipelagos. *Proceedings of the Royal Society B: Biological Sciences*, **281**, 20133246.
- Cadena, C.D., Ricklefs, R.E., Jiménez, I. & Bermingham, E. (2005) Is speciation driven by species diversity? *Nature*, **438**, E1–E2.
- Cameron, R.A.D., Triantis, K.A., Parent, C.E., Guilhaumon, F., Alonso, M.R., Ibanez, M., Martins, A.M.F., Ladle, R.J. & Whittaker, R.J. (2013) Snails on oceanic islands: testing the general dynamic model of oceanic island biogeography using linear mixed effect models. *Journal of Biogeography*, **40**, 117–130.
- Carracedo, J.C. & Tilling, R.I. (2003) *Geología y volcanología de islas oceánicas. Canarias – Hawaii*. Caja Canarias, Gobierno de Canarias, Santa Cruz de Tenerife.
- Chen, X. & He, F. (2009) Speciation and endemism under the model of island biogeography. *Ecology*, **90**, 39–45.
- Cornell, H.V. (2013) Is regional species diversity bounded or unbounded? *Biological Reviews*, **88**, 140–165.
- Economio, E.P., Klimov, P., Sarnat, E., Guénard, B., Lecroq, B. & Knowles, L.L. (2015) Global phylogenetic structure of the hyperdiverse ant genus *Pheidole* reveals the repeated evolution of macroecological patterns. *Proceedings of the Royal Society B: Biological Sciences*, **282**, 20141416.
- Gillespie, R.G. & Clague, D. (eds) (2009) *Encyclopedia of islands*. University California Press, Berkeley, CA.
- Givnish, T.J., Millam, K.C., Mast, A.R., Paterson, T.B., Theim, T.J., Hipp, A.L., Henss, J.M., Smith, J.F., Wood, K.R. & Sytsma, K.J. (2009) Origin, adaptive radiation, and diversification of the Hawaiian lobeliads (Asterales: Campanulaceae). *Proceedings of the Royal Society B: Biological Sciences*, **276**, 407–416.
- Grant, P.R. & Grant, B.R. (2008) *How and why species multiply: the radiation of Darwin's Finches*. Princeton University Press, Princeton, NJ.
- Gravel, D., Massol, F., Canard, E., Mouillot, D. & Mouquet, N. (2011) Trophic theory of island biogeography. *Ecology Letters*, **14**, 1010–1016.
- Haario, H., Laine, M., Mira, A. & Saksman, E. (2006) DRAM: efficient adaptive MCMC. *Statistical Computing*, **16**, 339–354.
- Hanski, I. (1999) *Metapopulation ecology*. Oxford University Press, Oxford.
- Heaney, L.R. (2000) Dynamic disequilibrium: a long-term, large-scale perspective on the equilibrium model of island biogeography. *Global Ecology and Biogeography*, **9**, 59–74.
- Hutchinson, G.E. (1959) Homage to Santa Rosalia or why are there so many kinds of animals? *The American Naturalist*, **93**, 145–159.
- Kisel, Y. & Barraclough, T.G. (2010) Speciation has a spatial scale that depends on levels of gene flow. *The American Naturalist*, **175**, 316–334.
- Kisel, Y., McInnes, L., Toomey, N.H. & Orme, C.D.L. (2011) How diversification rates and diversity limits combine to create large-scale species–area relationships. *Philosophical Transactions of the Royal Society B: Biological Sciences*, **366**, 2514–2525.
- Kvålseth, T.O. (1985) Cautionary note about R^2 . *American Statistician*, **39**, 279–285.
- Losos, J.B. & Schluter, D. (2000) Analysis of an evolutionary species–area relationship. *Nature*, **408**, 847–850.
- MacArthur, R.H. (1972) *Geographical ecology*. Harper and Rowe, New York.
- MacArthur, R.H. & Wilson, E.O. (1967) *The theory of island biogeography*. Princeton University Press, Princeton, NJ.
- Öckinger, E., Schweiger, O., Crist, T.O., Debinski, D.M., Krauss, J., Kuussaari, M., Petersen, J.D., Pöyry, J., Settele, J., Summerville, K.S. & Bommarco, R. (2010) Life-history traits predict species responses to habitat area and isolation: a cross-continental synthesis. *Ecology Letters*, **13**, 969–979.
- Olson, S.L. & James, H.F. (1984) The role of Polynesians in the extinction of the avifauna of the Hawaiian Islands. *Quaternary extinctions: a prehistoric revolution* (ed. by P.S. Martin and R.G. Klein), pp. 768–780. University of Arizona Press, Tucson, AZ.
- Patiño, J., Weigelt, P., Guilhaumon, F., Kreft, H., Triantis, K.A., Naranjo-Cigala, A., Sólomos, P. & Vanderpoorten, A. (2014) Differences in species–area relationships among the major lineages of land plants: a macroecological perspective. *Global Ecology and Biogeography*, **23**, 1275–1283.
- Phillimore, A.B. & Price, T.D. (2008) Density-dependent cladogenesis in birds. *PLoS Biology*, **6**, e71.
- Pigolotti, S. & Cencini, M. (2009) Speciation-rate dependence in species–area relationships. *Journal of Theoretical Biology*, **260**, 83–89.
- Price, J.P. & Clague, D.A. (2002) How old is the Hawaiian biota? Geology and phylogeny suggest recent divergence. *Proceedings of the Royal Society B: Biological Sciences*, **269**, 2429–2435.
- Rabosky, D.L. (2009) Ecological limits and diversification rate: alternative paradigms to explain the variation in species richness among clades and regions. *Ecology Letters*, **12**, 735–743.
- Rabosky, D.L. (2013) Diversity-dependence, ecological speciation, and the role of competition in macroevolution. *Annual Review of Ecology, Evolution, and Systematics*, **44**, 481–502.
- Rabosky, D.L. & Glor, R.E. (2010) Equilibrium speciation dynamics in a model adaptive radiation of island lizards. *Proceedings of the National Academy of Sciences USA*, **51**, 22178–22183.
- Rahbek, C. (2005) The role of spatial scale and the perception of large-scale species-richness patterns. *Ecology Letters*, **8**, 224–239.
- Ricklefs, R.E. (1987) Community diversity: relative roles of local and regional processes. *Science*, **235**, 167–171.
- Ricklefs, R.E. (2004) A comprehensive framework for global patterns in biodiversity. *Ecology Letters*, **7**, 1–15.
- Ricklefs, R.E. (2007) History and diversity: explorations at the intersection of ecology and evolution. *The American Naturalist*, **170**, S56–S70.

- Ricklefs, R.E. & Bermingham, E. (2007) Evolutionary radiations of passerine birds in archipelagos. *The American Naturalist*, **169**, 285–297.
- Rosenzweig, M.L. (1995) *Species diversity in space and time*. Cambridge University Press, New York.
- Rosindell, J. & Phillimore, A.B. (2011) A unified model of island biogeography sheds light on the zone of radiation. *Ecology Letters*, **14**, 552–560.
- Sánchez-Marco, A. (2010) New data and an overview of the past avifaunas from the Canary Islands. *Ardeola*, **57**, 13–40.
- Steadman, D.W. (2006) *Extinction and biogeography of tropical Pacific birds*. University Chicago Press, Chicago.
- Stuessy, T.F., Jakubowsky, G., Salguero Gómez, R., Pfosser, M., Schlüter, P.M., Fer, T., Sun, B.-Y. & Kato, H. (2006) Anagenetic evolution in island plants. *Journal of Biogeography*, **33**, 1259–1265.
- Triantis, K.A., Mylonas, M. & Whittaker, R.J. (2008) Evolutionary species–area curves as revealed by single-island endemics: insights for the interprovincial species–area relationship. *Ecography*, **31**, 401–407.
- Triantis, K.A., Guilhaumon, F. & Whittaker, R.J. (2012) The island species–area relationship: biology and statistics. *Journal of Biogeography*, **39**, 215–232.
- Valente, L.M., Etienne, R.S. & Phillimore, A.B. (2014) The effects of island ontogeny on species diversity and phylogeny. *Proceedings of the Royal Society B: Biological Sciences*, **281**, 20133227.
- Weigelt, P., Jetz, W. & Kreft, H. (2013) Bioclimatic and physical characterization of the world's islands. *Proceedings of the National Academy of Sciences USA*, **110**, 15307–15312.
- Whittaker, R.J. & Fernández-Palacios, J.M. (2007) *Island biogeography: ecology, evolution, and conservation*, 2nd edn. Oxford University Press, Oxford.
- Whittaker, R.J., Triantis, K.A. & Ladle, R.J. (2008) A general dynamic theory of oceanic island biogeography. *Journal of Biogeography*, **35**, 977–994.

- Zuur, A., Ieno, E.N., Walker, N., Saveiliev, A.A. & Smith, G.M. (2009) *Mixed effects models and extensions in ecology with R*. Springer, New York.

SUPPORTING INFORMATION

Additional supporting information may be found in the online version of this article at the publisher's web-site.

Appendix S1 Supplementary figures.

Appendix S2 Properties of the 14 oceanic archipelagos.

Appendix S3 Methods, additional analyses and results.

Appendix S4 Supplementary references.

BIOSKETCHES

Kostas Triantis has a long-term fascination with the biogeography of islands and also works on conservation biogeography and macroecology. He is currently working on diversity patterns of oceanic archipelagos.

Evan Economo has interests in documenting, understanding and conserving biodiversity at a variety of scales.

François Guilhaumon is interested in understanding the distribution of different aspects of terrestrial and marine diversity, with the goal of informing regional and global conservation efforts.

Robert Ricklefs has been interested since graduate school in the distribution of birds on islands, particularly in the West Indies, and is currently working on the host and geographic distribution of malarial parasites of island birds.

Editor: Shai Meiri