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A simple dynamic model explains the diversity of island birds worldwide

Citation for published version:

Valente, L, Phillimore, A, Melo, M, Warren, BH, Clegg, SM, Havenstein, K, Tiedemann, R, Illera, JC, Thébaud, C, Aschenbach, T & Etienne, RS 2020, 'A simple dynamic model explains the diversity of island birds worldwide', Nature, vol. 579, pp. 92-96. https://doi.org/10.1038/s41586-020-2022-5

Digital Object Identifier (DOI):

10.1038/s41586-020-2022-5

Link: Link to publication record in Edinburgh Research Explorer

Document Version: Peer reviewed version

Published In: Nature

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1	A simple dynamic model explains island bird diversity worldwide
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Colonisation, speciation and extinction are dynamic processes that influence 29 global patterns of species richness¹⁻⁶. Island biogeography theory predicts that 30 the contribution of these processes to the build-up of species diversity depends 31 on area and isolation^{7,8}. Remarkably, there has been no robust global test of this 32 prediction⁹, because neither the appropriate data nor the analytical tools have 33 been available. Here, we address both deficiencies to reveal, for island birds, the 34 empirical shape of the general relationships that determine how colonisation, 35 extinction and speciation rates covary with island area and isolation. We compile 36 the first global molecular phylogenetic dataset of birds on islands, based on the 37 terrestrial avifaunas of 41 oceanic archipelagos worldwide (including 596 avian 38 taxa), and apply novel methodology to estimate the sensitivity of island-specific 39 rates of colonisation, speciation and extinction to island features (area, isolation). 40 Our model predicts, with high explanatory power, several global relationships: a 41 decline of colonisation with isolation, a decline of extinction with area, and an 42 increase of speciation with area and isolation. Combining the theoretical 43 foundations of island biogeography^{7,8} with the temporal information contained in 44 molecular phylogenies¹⁰ proves a powerful approach to unveil the fundamental 45 relationships that govern variation in biodiversity across the planet. 46

A key feature of global diversity is the tendency for some areas to harbour many more 48 species than others^{7,8}. Uncovering the drivers and regulators of spatial differences in 49 diversity of simple systems such as islands is a crucial step towards understanding the 50 global distribution of species richness. The two most prominent biodiversity patterns in 51 fragmented or isolated environments worldwide are the increase of species richness 52 with area and the decline of richness with isolation^{8,11-14}. In their theory of island 53 biogeography, MacArthur and Wilson proposed how the processes of colonisation and 54 extinction could explain these patterns^{7,8}. They argued that the rates of these processes 55 are determined by the geographic context: colonisation decreases with isolation and 56 extinction decreases with area^{7,8}. They also suggested that rates of formation of island 57 endemic species via *in situ* speciation increase with island isolation and area⁸. Despite 58 an abundance of studies over five decades supporting the general patterns predicted by 59 MacArthur and Wilson^{2,15-18}, tests of predictions regarding the dependence of the 60 underlying processes – colonisation, speciation and extinction – on island geographic 61 context (area and isolation) are few in number, and are either restricted in temporal, 62 geographic, or taxonomic scope^{5,19,20} or seek to infer speciation rates in the absence of 63 data on the relationships among species^{2,16}. As a result, there has been no robust and 64 powerful test of MacArthur and Wilson's predictions on a global scale, and the effect of 65 area and isolation on biogeographical processes acting on macro-evolutionary time 66 scales remains largely unexplored. 67

68 Here we expand on approaches that leverage the information in time-calibrated molecular phylogenies of insular species^{1,10,21,22} to determine how the processes of 69 colonisation, speciation and extinction are influenced by area and isolation. The 70 dynamic stochastic model DAISIE¹⁰ (Dynamic Assembly of Islands through Speciation, 71 Immigration, and Extinction) can accurately estimate maximum likelihood (ML) rates of 72 colonisation, extinction and speciation rates (CES rates) from branching times 73 (colonisation times and any in situ diversification events) and endemicity status of 74 species resulting from one or multiple independent colonisations of a given island 75 system (e.g. all native terrestrial birds on an archipelago)¹⁰. This method can also detect 76 the presence or absence of diversity-dependence in rates of colonisation and speciation, 77 by estimating a carrying capacity (upper bound to the number of species in an island 78 system). Here we extend DAISIE to estimate, for the first time, the hyperparameters that 79

control the shape of the relationships between CES rates and the area and isolation of
 islands worldwide.

Accurate estimation of fundamental island biogeographic relationships requires 82 suitable data from many archipelagos, but divergence-dated phylogenies of complete 83 communities on islands remain scarce. Hence, we produced new dated molecular 84 phylogenies for the terrestrial avifaunas of 41 archipelagos worldwide. By 'archipelago' 85 we refer to both true archipelagos (composed of multiple islands) and isolated insular 86 units consisting of single islands (e.g. Saint Helena). For each archipelago we compiled 87 avian taxon lists (excluding introduced, marine, migratory, and aquatic species, as well 88 as birds of prey, rails and nocturnal birds, see Methods) and collected physical data (Fig. 89 1, Supplementary Data 1 and 2). We use archipelagos as our insular unit, because the 90 high dispersal abilities of birds within archipelagos imply that for birds, archipelagos 91 can be considered equivalent to single islands for less dispersive taxa²³, and because 92 archipelagos constitute the most appropriate spatiotemporal unit for framing analyses 93 of biodiversity patterns at a large scale^{2,24,25}. We extracted colonisation and speciation 94 times for each archipelago from the phylogenetic trees, producing a 'global dataset' for 95 the 41 archipelagos, which includes each archipelago's complete extant avifauna, plus 96 all species known to have become extinct due to anthropogenic causes. The dataset 97 comprises 596 insular taxa from 491 species. The phylogenies revealed a total of 502 98 archipelago colonisation events and 26 independent in-situ 'radiations' (cases where 99 100 diversification has occurred within an archipelago) ranging in size from 2 to 33 species (the Hawaiian honeycreepers being the largest clade). The distribution of colonisation 101 times is summarised in Fig. 1 and the full dataset is given in Supplementary Data 1. 102

Our extension of the DAISIE framework allows us to estimate hyperparameters 103 that control the relationship between archipelago area and isolation and archipelago-104 specific local CES rates, i.e., rates of colonisation, cladogenesis (within-archipelago 105 speciation involving *in situ* lineage splitting), anagenesis (within-archipelago speciation 106 by divergence from the mainland without lineage splitting), natural extinction rates and 107 carrying capacity. We tested the hypothesis that area and distance from the nearest 108 mainland have an effect on the specific CES rates, and, where a significant effect was 109 identified, estimated its shape and scaling. We developed a set of a priori models 110 (Supplementary Table 1) where CES rates are power law functions of archipelago 111 features. Area has been proposed to have a positive effect on cladogenesis and carrying 112

capacity^{3,5,8}, and a negative effect on extinction rates^{8,26}. Archipelago isolation is
hypothesised to reduce colonisation rates⁷ and elevate anagenesis rates²⁷. Models
including or excluding diversity-dependence in rates of colonisation and cladogenesis¹⁰
(i.e. estimating a carrying capacity parameter) were compared. We also considered a set
of *post hoc* models with alternative shapes for the relationships (*post hoc* power and *post hoc* sigmoid models, see Methods, Supplementary Table 1).

We fitted a set of 28 candidate models to the global dataset using ML 119 (Supplementary Table 2). The shape of the relationship of CES rates with area and 120 distance for the two best models is shown in Fig. 2. Under the preferred *a priori* model 121 (lowest value of Bayesian Information Criterion (BIC); M14, eight parameters) 122 colonisation rates decline with archipelago isolation (exponent of the power law = -0.25123 (95% confidence interval = -0.17 - -0.34)) and extinction rate decreases with area 124 (scaling = -0.15 (-0.11 - 0.18)). Rates of cladogenesis increase with area (scaling = 0.26125 (0.13 - 0.37), whilst anagenesis increases with isolation (scaling = 0.42 (0.24 - 0.61)). 126 The preferred *post hoc* model (M19, eight parameters) was also the preferred model 127 overall and differs qualitatively from the preferred *a priori* model M14 only in the 128 cladogenesis function. In M14 cladogenesis is solely a function of area, whereas in M19 129 cladogenesis depends interactively and positively on both area and distance from the 130 nearest mainland, such that the cladogenesis-area relationship is steeper for more 131 isolated archipelagos (Fig. 2 and Extended Data Fig. 1). In addition, we found no 132 133 evidence for diversity-dependence, as the carrying capacity (K) was estimated to be much larger than the number of species on the island and models without a K134 parameter (no upper bound to diversity), such as M14 and M19, performed better than 135 models including one (Supplementary Table 2). We also tested whether the inclusion of 136 a combination of true archipelagos and single islands in our dataset could have affected 137 our results, for example if opportunities for allopatric speciation are higher when an 138 area is subdivided into multiple islands²⁸. We repeated analyses excluding single island 139 units and found that the same model (M19) is preferred with similar parameter 140 estimates. Hence, we discuss only the results for the main dataset (including both single 141 islands and true archipelagos). Our results are robust to uncertainty in colonisation and 142 branching times (see section 'Sensitivity to alternative divergence times and tree 143 topologies'). 144

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A parametric bootstrap analysis of the two preferred models (M14 and M19)

demonstrated that the method is able to recover hyperparameters with high precision 146 and little bias (Extended Data Fig. 2). In order to test the significance of the 147 relationships between area, isolation and CES rates, we conducted a randomization test 148 on the global dataset with reshuffled areas and distances. This test estimated the 149 exponent hyperparameters as zero in most reshuffled cases (i.e. no effect of area or 150 isolation detected; Extended Data Fig. 3), confirming that it is the observed 151 relationships between diversity and archipelago characteristics that generate our 152 parameter estimates. 153

To assess model fit we simulated archipelago communities under the best model 154 (M19) and found that for most archipelagos the observed diversity metrics (numbers of 155 species, cladogenetic species and colonisations) were similar to the expected numbers, 156 with some exceptions: for example, diversity was underestimated for Comoros and São 157 Tomé & Príncipe (Fig. 3 and Extended Data Fig. 4). The ability of the model to explain 158 observed values (pseudo- R^2 for total species = 0.72, cladogenetic species = 0.52, 159 colonisers = 0.60) was very high considering the model includes only eight parameters 160 (at least 12 parameters would be needed if each rate depended on area and isolation, 161 and at least 164 parameters if each archipelago was allowed to have its own 162 parameters) and was able to explain multiple diversity metrics. This represents a very 163 large proportion of the explanatory power one would expect to obtain for data 164 generated under the preferred model (Extended Data Fig. 5). Simulations under the best 165 166 model reproduced the classic observed relationships between area, distance and diversity metrics (Fig. 4). 167

Our approach reveals the empirical shape of fundamental biogeographic 168 relationships that have hitherto largely evaded estimation. In agreement with recent 169 studies^{2,29}, we found strong evidence for a decline of rates of colonisation with isolation 170 and of rates of extinction with area, confirming two of the key assumptions of island 171 biogeography theory⁷. The colonisation-isolation effect was detected despite the decline 172 of avian richness with distance from the nearest mainland in our empirical data not 173 being as pronounced as in other less mobile taxa^{4,11}, revealing isolation to be a clear 174 determinant of probability of immigration and successful establishment of populations 175 even in a highly dispersive group such as birds. The extinction-area relationship has 176 been a fundamental empirical generalization in conservation theory (for example for 177 the design of protected areas³⁰) but this is the first time the shape of this dependence is 178

characterized at the global spatial scale and macro-evolutionary time scale.

We provide novel insights into the scaling of speciation with area and isolation. 180 Contrary to previous work on within-island speciation, which suggested the existence of 181 an area below which cladogenesis does not take place on single islands⁵, we do not find 182 evidence for such an area threshold at the archipelago level, and under our model 183 speciation is predicted to be non-zero even at small areas. In addition, our post hoc 184 finding that rates of cladogenesis increase through an interactive effect of both island 185 size and distance from the nearest mainland (Fig. 2 and Extended Data Fig. 1) provides a 186 mechanism that limits radiations to archipelagos that are both large and remote^{6,27}. 187 Why this interaction exists requires further investigation, but one possibility is that 188 unsaturated niche space provides greater opportunities for diversification⁶. In addition 189 to the effects of physical features on cladogenesis, we found that rates of anagenesis 190 increase with island isolation. While impressive insular radiations tend to receive the 191 most attention from evolutionary biologists (e.g. Darwin's finches or Hawaiian 192 honeycreepers), our phylogenies revealed that the majority of endemic birds in our 193 dataset in fact display an anagenetic pattern (at the time of human arrival 231 of 350 194 endemic species had no extant sister taxa on the archipelago and there were only 26 195 extant in situ radiations). The positive effect of archipelago isolation on rates of 196 anagenesis that we estimate suggests this fundamental but overlooked process is 197 impeded by high levels of movement between island and mainland populations. 198

A variety of global patterns of biodiversity have been described – from small 199 islands and lakes, up to biomes and continents - but the processes underpinning them 200 remain little explored. Our simulations using parameters estimated from data were able 201 to reproduce classic global patterns of island biogeography across 41 archipelagos (Fig. 202 4). This advances our understanding of macro-scale biology, by providing missing links 203 between local process, environment and global patterns. Over half a century since the 204 seminal work of MacArthur & Wilson⁷, we now have the data and tools to go beyond 205 statistical descriptions of diversity patterns, enabling us to quantify community-level 206 processes that have long been elusive. 207

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279 MAIN TEXT FIGURE LEGENDS

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Figure 1 – Archipelago and island bird colonisation time data. Circles show number 281 of species belonging to our focal group (both extinct and extant) found in each 282 archipelago (at the time of human arrival). Numbers on map correspond to numbers to 283 the left of the archipelago name. Numbers to the right of the archipelago name: number 284 of species from our focal assemblage on the archipelago | percentage of species sampled 285 in the phylogenetic trees. Even species not sampled in the trees are accounted for by 286 including them as missing species that could have colonised any time since emergence 287 of the archipelago. Colonisation times plot: grey horizontal lines - archipelago ages 288 (Extended Data Table 1). Violin plots (blue) show the kernel density of the distribution 289 of times of colonisation of bird species in each archipelago, obtained from the 290 phylogenetic trees. Thick black line inside violin - interquartile distance; thin black line -291 95% CI; black dot - median. Archipelagos with no violin plot or dots are cases for which 292 no species of our focal assemblage were present at the time of human arrival, or none 293 were sampled using molecular data. Birds from left to right: Seychelles sunbird, 294 Seychelles magpie robin, silvereye, Príncipe thrush, laurel pigeon, dodo (extinct), 295 Mauritius fody, red-moustached fruit dove (extinct), Galápagos warbler, Norfolk kaka 296 (extinct). Bird images used with permission from: Cláudia Baeta, Pedro Cascão, Martijn 297 Hammers, Julian Hume, Dubi Shapiro and Juan Varela. There are no *in-situ* radiations in 298 the Mascarenes (Mauritius, Reunion and Rodrigues) because we treat the islands as 299 separate entities (but see sensitivity analyses). 300

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Figure 2 - Estimated relationships between island area and isolation and local 302 island biogeography parameters. Isolation measured as distance to the nearest 303 mainland (D_m). Based on the maximum likelihood global hyperparameters of the best 304 models (equations describing the relationships given in Supplementary Table 1). 305 Darker lines - M14 model, lighter lines - M19 model. Under the M14 model, 306 cladogenesis rate depends only on area. Under the M19 model, cladogenesis rate 307 increases with both area and D_m , and thus lines for more (far, 5,000 km) and less (near, 308 50 km) isolated islands are shown. See also Extended Data Fig. 1 for the relationship of 309 cladogenesis with both area and distance under the M19 model. 310

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Figure 3 – Goodness of fit of the preferred model (M19). The map identifies whether the diversity metrics were well estimated (empirical value matches 95% confidence interval of simulations), underestimated (empirical value higher than 95% interval) or overestimated (empirical value lower than 95% interval). Intervals based on 1000 simulations of each archipelago (see Extended Data Fig. 4). Numbers indicating archipelagos on the map match those in Fig. 1.

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Figure 4 – Observed and predicted island diversity-area and island diversitydistance relationships. Grey vertical lines show the 95% confidence intervals across 1,000 datasets simulated for each of the 41 archipelagos assuming the M19 model. Blue points: mean values of the simulations; blue line – fitted line for the simulated data; red points – observed values in the empirical data; red line – fitted line for the empirical data; red shaded area is the 95% confidence interval of the predicted relationship for the empirical data.

326 Methods

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328 Archipelago selection

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We focus on oceanic islands, i.e. volcanic islands that have never been connected to any 330 other landmass in the past. We also include the Granitic Inner Seychelles, even though 331 these islands have a continental origin, because they have been separated from other 332 landmasses for a very long period of time (64 million years (Ma)³¹) and can be 333 considered quasi-oceanic, as all extant avian species originated in much more recent 334 times. The 41 archipelagos chosen are located in the Atlantic, Indian and Pacific oceans, 335 with latitudes between 45° North and South. Islands within these archipelagos are 336 separated by a maximum of 150 km. The sole exceptions are the Azores and Hawaii, two 337 very isolated systems where the distances between some islands exceed this value. The 338 shape files used to plot the maps of Figs. 1 and 3 were obtained from Weigelt, Jetz and 339 Kreft 2013³². 340

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342 Physical and geological data

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Full archipelago data is given in Supplementary Data 2 and Extended Data Table 1. We 344 obtained data on total contemporary landmass area for each archipelago. For our 345 isolation metric, we computed the minimum round earth distance to the nearest 346 mainland (D_m) in km using Google Earth. We considered 'nearest mainland' to be the 347 nearest probable source of colonists (but see 'Sensitivity to archipelago selection and 348 isolation metrics' section for different isolation metrics). This is the nearest continent 349 except for island groups that were closer to Madagascar, New Guinea or New Zealand 350 than to the continent, in which case we assigned these large continent-like islands as the 351 mainland. This is supported by our phylogenetic data – for example, many Indian Ocean 352 island taxa have closest relatives on Madagascar rather than mainland Africa. 353

Island palaeo-areas and past archipelago configurations have been shown to be better predictors of endemic insular diversity than contemporary area^{15,33}. In contrast, island total native and non-endemic richness is better predicted by present island characteristics^{15,33}. With insufficient data on island ontogeny being available (i.e. describing empirical area trajectories from island birth to present) we therefore analysed contemporary area and isolation as currently the most appropriate units forour dataset.

We conducted an extensive survey of the literature and consulted geologists to 361 obtain archipelago geological ages (Extended Data Table 1), treating the age of the 362 oldest currently emerged island as an upper bound for colonisation. Islands may have 363 been submerged and emerged multiple times and we consider the age of the last known 364 emergence. For the Aldabra Group we used an age older than the published estimate. 365 The current estimated age of re-emergence of Aldabra is 0.125 Ma³⁴, but nine out of 12 366 Aldabra colonisation events in our dataset are older, suggesting the archipelago was not 367 fully submerged prior to this and may have been available for colonisation for a longer 368 period. Therefore, for Aldabra we used an older upper bound of 1 Ma for colonisation, 369 although we acknowledge that the mitochondrial markers used for dating may not 370 provide sufficient resolution at the shallow temporal scale of the published age. For 371 Hawaii, the colonisation times we obtained for more than half of the colonisation events 372 were older than the age of the current high islands that is often used as a maximum age 373 374 for colonisation (~ 5 Ma). Therefore, instead of this age, we used the much older estimate of 29.8 Ma of the Kure Atoll³⁵ to account for currently submerged or very low-375 lying Hawaiian islands that could have received colonists in the past. For Bermuda and 376 Marianas, we could not find age estimates in the literature, and we therefore consulted 377 geologists to obtain these (P. Hearty, R. Stern and M. Reagan, pers. comm., Extended 378 379 Data Table 1).

380

381 Island avifaunas

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Our sampling focused on native resident terrestrial birds and we considered only birds 383 that colonise by chance events (e.g. hurricanes, rafts). We thus excluded marine and 384 migratory species, because they are capable of actively colonising an island at a much 385 higher rate. We focused on songbird-like and pigeon-like birds, which constitute the 386 majority of terrestrial (land-dwelling) birds on islands. Following a precedent set by 387 previous work^{10,27,36}, we included only species from the same trophic level (in the spirit 388 of MacArthur and Wilson's model): we excluded aquatic birds, birds of prey, rails (many 389 are flightless or semi-aquatic) and nightjars (nocturnal). We also excluded introduced 390 and vagrant species. Including species such as rails and owls (which are components of 391

many island avifaunas) would have led to a higher estimate of the product of colonisation rate and mainland pool size due to a larger mainland pool, and potentially to higher estimated rates of anagenesis (many owl or rail species are island endemics with no close relatives on the islands).

For the focal avian groups, we compiled complete taxon lists for each of the 41 396 archipelagos based on recent checklists from Avibase (http://avibase.bsc-eoc.org), 397 which we cross-checked with the online version of the Handbook of the Birds of the 398 *World* (HBW³⁷). We followed HBW's nomenclature and species assignations, except for 399 12 cases where our phylogenetic data disagree with HBW's scheme (noted in the 400 column 'Taxonomy' of Supplementary Data 1). For example, in eleven cases 401 phylogenetic trees support raising endemic island subspecies to species status (we 402 sampled multiple samples per island taxon and outgroup, and the island individuals 403 form a reciprocally monophyletic well-supported clade), and for these taxa we decided 404 405 it was more appropriate to use a phylogenetic species concept so as not to underestimate endemicity and rates of speciation (Supplementary Data 1). We re-ran 406 DAISIE analyses using HBW's classification and found that the ML parameters are very 407 similar and thus we report only the results using the scheme based on the phylogenies 408 produced for this study. 409

For each bird species found on each archipelago we aimed to sample sequence data for individuals on the archipelago and the closest relatives outside the archipelago (outgroup taxa). Our sampling success per archipelago is shown on Fig. 1 and Extended Data Table 1.

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416 **Extinct species**

We do not count extinctions with anthropogenic causes as impacting the natural background rate of extinction. Therefore, we explicitly include species where there is strong evidence that they have been extirpated by humans. We treat taxa extirpated on an archipelago by humans as though they had survived in that archipelago until the present following the approach of Valente, Etienne and Dávalos 2017³⁸.

We identified anthropogenic extinctions based on published data³⁹⁻⁴⁶ and personal comments (Josep Antoni Alcover and Juan Carlos Rando on unpublished Macaronesian taxa; Ferran Sayol and Søren Faurby). We include the species present on

the islands that belong to our archipelago definition as in Supplementary Data 2. We 425 excluded largely hypothetical accounts or pre-Holocene fossils that greatly pre-date 426 human arrival. Our dataset accounts for 153 taxa that were present upon first human 427 contact and have gone extinct since, probably because of human activities including 428 human introduction of invasive species. To our knowledge 71 of these taxa have 429 previously been sequenced using ancient DNA or belong to clades present in our trees, 430 and we were thus able to include them in the phylogenetic analyses as regular data (n = 1)431 54), or as missing species by adding them as unsampled species to a designated clade (n 432 = 17). For the remaining 82 extinct taxa, sequences were not available and we were 433 unable to obtain samples and to allocate them to clades. We assume that these taxa 434 represent extinct independent colonisations and we included them in the analyses using 435 the "Endemic MaxAge" and "Non endemic MaxAge" options in DAISIE, which assume 436 that they have colonised at any given time since the birth of the archipelago (but before 437 any in situ cladogenesis event). As an example, our dataset includes the 27 species of 438 Hawaiian birds belonging to our focal group that are known to have gone extinct since 439 human colonisation. Eight of these species were included using DNA data, 17 were 440 added as missing species to their clades (14 honeycreepers and 3 *Myadestes*) and two 441 were added using the Endemic_MaxAge option in DAISIE (Corvus impluviatus and 442 Corvus viriosus). 443

444

445 Sequence data: GenBank

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We conducted an extensive search of GenBank for available DNA sequences from the 447 596 island bird taxa fitting our sampling criteria and from multiple outgroup taxa, using 448 software Geneious 11⁴⁷. The molecular markers chosen varied from species to species, 449 depending on which marker was typically sequenced for the taxon in question, the 450 commonest being cytochrome b (cyt-b). In total, we downloaded 3155 sequences from 451 GenBank. For some taxa, sequences from both archipelago and close relatives from 452 outside the archipelago were already available from detailed phylogenetic or 453 phylogeographic analyses. In some cases, a target species had been sampled, but only 454 from populations outside the archipelago. In other cases, the species on the archipelago 455 had been sampled, but the sampling of the relatives outside of the archipelago was 456 lacking or only from distant regions, which meant a suitable outgroup was not available 457

on GenBank. Finally, for some species there were no previous sequences available on
GenBank. GenBank accession numbers and geographical origin for the downloaded
sequences are provided on the DNA matrices and maximum clade credibility trees
(uploaded to Mendeley Data).

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463 Sequence data: new samples

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Sequences available on GenBank covered only 54% (269/502) of the total independent 465 colonisation events. We improved the sampling by obtaining new sequences for many 466 island taxa (n = 174 taxa) and from their close relatives from continental regions (n = 174 taxa) and from their close relatives from continental regions (n = 174 taxa) and from their close relatives from the continental regions (n = 174 taxa) and from the close relatives from the continental regions (n = 174 taxa) and from the close relatives from the close 467 78). We obtained new samples from three sources: field trips, research collections and 468 colleagues who kindly contributed field samples. New samples were obtained during 469 field trips conducted by M.M. (Gulf of Guinea and African continent); B.H.W. and C.T. 470 (Comoros, Mauritius Isl., Rodrigues, Seychelles); S.M.C. (New Caledonia); J.C.I. 471 (Macaronesia, Europe and Africa) and L.V. (New Caledonia), between 1999 and 2017. 472 473 Samples of individuals were captured using mist-nets or spring traps baited with larvae. Blood samples were taken by brachial venipuncture, diluted in ethanol or Queen's lysis 474 buffer in a microfuge tube. Birds were released at the point of capture. Aldabra Group 475 samples were obtained from research collections of the Seychelles Islands Foundation. 476 Museum samples from several Galápagos and Comoros specimens were obtained on 477 478 loan from respectively, the California Academy of Sciences and the Natural History Museum London. Additional samples from various localities (Aldabra Islands, Iberian 479 Peninsula, Madagascar and Senegal) were kindly provided by collaborators, as indicated 480 in Supplementary Table 3. Sample information and GenBank accession numbers for all 481 new specimens are provided in Supplementary Table 3. 482

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DNA was extracted from blood, feathers and museum toe-pad samples using QIAGEN DNeasy Blood and Tissue kits (QIAGEN, USA). For museum samples, we used a dedicated ancient DNA lab facility at the University of Potsdam to avoid contamination. The cyt-*b* region (1100 base pairs) was amplified using the primers shown in Extended Data Table 2. DNA from historical museum samples was degraded and cyt-*b* could not be amplified as a single fragment. We thus designed internal primers to sequence different overlapping fragments in a stepwise fashion (Extended Data Table 2).

490

Polymerase chain reactions (PCR) were set up in 25 µL total volumes including 5

 μ L of buffer Bioline MyTaq, 1 μ L (10 mM) of each primer, and 0.12 μ L MyTaq 491 polymerase. PCRs were performed with the following thermocycler conditions: initial 492 denaturation at 95° C for 1 min followed by 35 cycles of denaturation at 95° C for 20 s, 493 with an annealing temperature of 48° C for 20 s, and extension at 72° C for 15 s min and 494 a final extension at 72° C for 10 min. Amplified products were purified using 495 Exonuclease I and Antarctic Phosphatase, and sequenced at the University of Potsdam 496 (Unit of Evolutionary Biology/Systematic Zoology) on an ABI PRISM 3130xl sequencer 497 (Applied Biosystems) using the BigDye Terminator v3.1 Cycle Sequencing Kit (Applied 498 Biosystems). We used Geneious 11 to edit chromatograms and align sequences. 499

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502 Phylogenetic analyses

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To estimate times of colonisation and speciation for each archipelago we produced new 504 divergence dated phylogenies or compiled published dated trees, to yield a total of 91 505 independent phylogenies (maximum clade credibility trees and posterior distribution 506 deposited in Mendeley for all new trees produced for this study; the 11 previously 507 published trees are available upon request). Information on all alignments and trees, 508 including molecular markers, source of data, calibration method and substitution model 509 are given in Extended Data Tables 3 and 4 and Supplementary Table 4. The majority of 510 alignments/phylogenies focus on a single genus, but some include multiple closely 511 related genera or higher order clades (family, order) depending on the diversity and 512 level of sampling of the relevant group (taxonomic scope indicated in Extended Data 513 Tables 3 and 4). Most alignments include taxa from a variety of archipelagos. 514 Alignments were based on a variety of markers, according to which marker had been 515 mostly sequenced for a given group. 516

For the new dating analyses conducted for this study, we created 80 separate alignments for different groups using a combination of sequences from GenBank (n = 3155) and new sequences (n = 252) produced for this study. In some cases, we obtained DNA alignments directly from authors of previous studies and these are credited in Extended Data Table 3. Phylogenetic divergence dating analyses were performed in BEAST 2^{48} . For each alignment we performed substitution model selection in jModeltest⁴⁹ using the Bayesian information criterion. We used rates of molecular

evolution for avian mitochondrial sequences, which have been shown to evolve in a 524 clock-like fashion at an average rate of $\sim 2\%$ per Ma⁵⁰. Molecular rate calibrations can be 525 problematic for ancient clades, due to high levels of heterotachy in birds⁵¹. In addition, 526 mitochondrial DNA saturates after about 10 to 20 million years, and genetic distances of 527 more than 20% may provide limited information regarding dating⁵². Therefore, we only 528 used molecular rate dating to extract node ages for branching events at the tips of the 529 trees, at the species or population level (oldest colonisation time in our dataset is 15.3 530 Ma, but most are much younger). Rates of evolution were obtained from the literature 531 and varied between different markers and taxonomic group (Supplementary Table 4). 532 We applied the avian mitochondrial rates estimated from cyt-b by Weir and Schluter⁵⁰ 533 (but see 'Sensitivity to alternative divergence times and tree topologies' section for 534 different rates). 535

We applied a Bayesian uncorrelated lognormal relaxed clock model. For each analysis, we ran two independent chains of between 10 and 40 million generations, with a birth-death tree prior. We assessed convergence of chains and appropriate burnins with Tracer, combined runs using LogCombiner, and produced maximum clade credibility trees with mean node heights in Tree Annotator. We produced a total of 80 maximum clade credibility trees.

For 11 groups (Extended Data Table 4), well-sampled and rigorously-dated phylogenies were already available from recent publications, all of which conducted Bayesian divergence dating using a variety of calibration methods, including fossils and molecular rates. We obtained maximum clade credibility trees from these studies from online repositories or directly from the authors (Extended Data Table 4).

547

548 **Colonisation and branching times**

549 The nodes selected in the dated trees for estimates of colonisation and branching times are given for each taxon in Supplementary Data 1. Our node selection approach was as 550 follows. For cases in which samples representing species or populations from 551 archipelagos formed a monophyletic clade consisting exclusively of archipelago 552 individuals, we used the stem age of this clade as colonisation time. For cases in which 553 only one individual of the archipelago was sampled, we used the length of the tip 554 leading to that individual, which is equivalent to the stem age. For cases in which the 555 archipelago individuals were embedded in a clade containing mainland individuals of 556

the same species, i.e. paraphyly or polyphyly; we assumed (based on morphological 557 characteristics) that this is due to incomplete lineage sorting of the insular and 558 mainland lineages, and we therefore used the MRCA of the archipelago individuals, or 559 the crown node when the MRCA coincides with the crown. For these later cases using 560 the stem would most likely have been an overestimation of the colonisation time, as we 561 assume that colonisation happens from the mainland to the archipelago. For such cases 562 we applied the ages using the "MaxAge" option in DAISIE, which integrates over the 563 possible colonisation times between the present and the upper bound. A robustness test 564 of our results to node choice is given in section "Sensitivity to alternative branching 565 times and tree topologies". 566

For a total of 19 endemic taxa we could not obtain sequences, but we could 567 allocate them to a specific island clade (e.g. Hawaiian honeycreepers and solitaires). 568 These were added as missing species to that clade. For 96 non-endemic taxa we could 569 not obtain sequences of individuals from the archipelago, but we could obtain 570 sequences from the same species from different regions. For these cases we used the 571 572 crown or the stem age of the species as an upper bound for the age of the colonisation event, using the "Non_endemic_MaxAge" option in DAISIE. Finally, for 124 taxa (20.8 %) 573 no sequences of individuals from the archipelago were available on GenBank and we 574 were not able to obtain samples for sequencing from the species or from close relatives. 575 We assumed these cases constituted independent colonisations that could have taken 576 577 place any time since the origin of the archipelago and the present, and applied the "Non_endemic_MaxAge" and "Endemic_MaxAge" options in DAISIE with a maximum age 578 equal to the archipelago age. DAISIE makes use of this information⁵³. 579

580

581 Global dataset characteristics

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Data points from taxa of the same archipelago were assembled into 41 archipelagospecific datasets. These 41 datasets were in turn assembled into a single dataset (D1) which was analysed with DAISIE (D1 DAISIE R object, available in Mendeley Data <u>https://doi.org/10.17632/sy58zbv3s2.2</u>). This dataset (information Supplementary Data 1) has a total of 596 taxa (independent colonisation events plus species within radiations), covering 491 species from 203 different genera and eight orders. All taxa were included in the analyses: those which we sampled in phylogenies, but also those for which sequences or phylogenies could not be obtained and which were included
 following the approaches described in the *colonisation and branching time* methods. A
 summary of diversity and sampling per archipelago is given in Extended Data Table 1.

593

594 Sampling completeness

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In total, we produced new sequences from 252 new individuals, comprising 90 different 596 species from 45 different genera, covering an additional 110 colonisation events that 597 had never before been sampled (i.e. populations from islands where the species had not 598 been sampled before). For at least 12 of these 90 species, we found no previous 599 sequences on GenBank, including island endemics from Comoros, Galápagos, Rodrigues 600 and São Tomé (Supplementary Table 5). The new sequences from 252 individuals 601 increase the molecular sampling for extant colonisation events from 60% (223/373) to 602 89% (332/373). If we include historically extinct colonisations, we increased the 603 molecular sampling from the existing 54% (269/502) of colonisation events to 75%604 605 (379/502). We also substantially increased molecular sampling of continental relatives, adding 78 new individuals from the continent or islands surrounding our archipelagos, 606 covering 43 different species. The percentage of taxa sampled in phylogenies varied 607 widely between archipelagos (Extended Data Table 1 and Fig. 1). For eight archipelagos 608 (Bermuda, Fernando de Noronha, Pitcairn, Rapa Nui, Rodrigues, Saint Helena, Society 609 610 Islands and Tonga) less than 50% of the species were sampled in phylogenies, and thus the majority of the species for these island groups were added with maximum ages and 611 endemicity status. For 13 archipelagos, which accounted for more than a third of the 612 total species, over 90% of the species were sampled in phylogenies. 613

614

615 DAISIE

616

We used the method DAISIE¹⁰ (Dynamic Assembly of Islands through Speciation, Immigration, and Extinction) to estimate rates of species accumulation (colonisation, speciation and extinction) on the archipelagos. The model assumes that after the origin of an island, species can colonise from a mainland pool. Once a species has colonised, it may remain similar to its mainland ancestor (non-endemic species), become endemic through anagenetic speciation (new endemic species is formed without lineage splitting on the island), split into new species via cladogenetic speciation and/or go extinct. A
 carrying capacity (i.e. maximum number of species each colonist lineage can attain) is
 implemented, such that rates of cladogenesis and colonisation decline with increasing
 number of species in the colonising clade.

The only effect of anagenesis under DAISIE is that the colonising species becomes 627 endemic, because further anagenesis events on the endemic species do not leave a 628 signature in the data. However, the rate of anagenesis is not systematically 629 underestimated. Suppose the rate was higher; it would then follow that colonising 630 species would also become endemic faster, and we would see more endemic species. 631 Thus, the number of endemic species determines the rate of anagenesis, and DAISIE 632 estimates the true rate of anagenesis without systematic bias. Further anagenesis 633 events do not have an effect on the state variables, and hence do not enter the equations 634 anymore. 635

In its parameterization of extinction, DAISIE accounts for the fact that there may 636 have been several lineages that were present on the insular system in the past but 637 which went completely extinct due to natural causes, leaving no extant descendants. 638 Simulations have shown that the rate of natural extinction is usually well estimated in 639 DAISIE (Methods section *Measuring precision and accuracy* and ref. ⁵³). Studies on 640 phylogenies of single clades suggest that phylogenetic data on only extant species 641 provide less information on extinction than on speciation (or rather diversification 642 643 rates⁵⁴). However, there is information-content in such data⁵⁵, especially when diversification dynamics are diversity-dependent⁵⁶. Moreover, here we use colonisation 644 times in addition to phylogenetic branching times to estimate extinction rates, and we 645 are estimating hyperparameters that theory suggests correlate with extinction (i.e. 646 area). Finally, we use data from many independent colonisations, which increases the 647 power of our statistical method considerably, and decreases the bias, as ML is known to 648 asymptotically provide unbiased estimates. 649

650

651 Estimating global hyperparameters

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Our aim is to examine the dependencies of the parameters that govern species assembly (colonisation, extinction, cladogenesis, anagenesis (CES rates), and carrying capacity) on the features of archipelagos (area, isolation). We developed a new method to estimate global hyperparameters that control the relationship between two key archipelago features (area and isolation) and archipelago-specific (local) CES rates. One can estimate directly from the global dataset the shape of the relationship between isolation and colonisation rate that maximizes the likelihood for the entire dataset.

Our method finds the hyperparameters that maximize the likelihood of the entire 660 dataset, i.e. the sum of the log likelihoods for each archipelago. We tested the hypothesis 661 that area and distance from the nearest mainland have an effect on CES rates 662 (cladogenesis, anagenesis, extinction and colonisation). If an effect was identified we 663 also estimated the scaling of the effect. We developed a set of *a priori* models where CES 664 rates are affected by archipelago features as is often assumed in the island 665 biogeography literature (Supplementary Table 1). For the *a priori* models, we 666 considered that CES rates are determined by a power function of area or distance. In the 667 power function, par = par_0I^h , where par is the CES rate (e.g. local rate of colonisation), 668 par_0 is the initial value of the biogeographical rate (e.g. global initial rate of 669 colonisation), I is the physical variable (area or distance) and h is the strength of the 670 relationship. The exponent h can be negative or positive depending on the nature of the 671 relationship. par_0 and h are the hyperparameters. If the exponent h is estimated as zero, 672 there is no relationship between I and the parameter. By including or excluding *h* from 673 the different relationships we can compare different models with the effects switched 674 on or off (Supplementary Table 1, e.g. in model M1 all relationships are estimated, but in 675 676 model M2 the exponent of the relationship between anagenesis and distance is fixed to zero and thus anagenesis does not vary with distance). 677

In addition to the *a priori* models, we considered a set of *post hoc* models with 678 alternative shapes of relationships. We fitted two types of *post hoc* models: power 679 models and sigmoid models (Supplementary Table 1). In the *post hoc* power models we 680 modelled all parameters as in the *a priori* models, except for cladogenesis: we allowed 681 cladogenesis to be dependent both on area and distance. The reason for this is that we 682 found that the predicted number of cladogenetic species under the *a priori* models were 683 not as high as observed, so we examined whether including a positive effect of distance 684 would improve the fit. We described the relationship between area, distance and 685 cladogenesis using different functions - one model where there is an additive effect of 686 area and distance (M15); and three models (M16, M17, M18) where the effect of area 687 and distance is interactive. In addition, we fitted a model identical to M16 but with one 688

parameter less (M19). The reason for this was that this parameter (y) was being estimated as zero in M16

In the *post hoc* sigmoid models, we allowed the relationship between distance and 691 a given parameter to follow a sigmoid rather a power function. The rationale for this 692 was that we wanted to investigate whether for birds the effect of distance on a 693 parameter only starts to operate after a certain distance from the mainland, as below 694 certain geographical distances archipelagos are within easy reach for many bird species 695 by flight so that at these distances the island behaves almost as part of the mainland 696 from a bird's perspective. We fitted nine different sigmoid models (Supplementary 697 Table 1), allowing cladogenesis, anagenesis and colonisation to vary with distance 698 following a sigmoid function. The sigmoid function we used has an additional parameter 699 in comparison to power functions. 700

In total we fitted 28 candidate models (14 a priori, 14 post hoc) to the global 701 dataset using ML. We fitted each model using 20 initial sets of random starting 702 parameters to reduce the risk of being trapped in local likelihood suboptima. We used 703 the age of each archipelago (Extended Data Table 1) as the maximum age for 704 colonisation. We assumed a global mainland species pool M of 1000 species. The 705 product of *M* and the intrinsic rate of colonisation $(\sqrt{\gamma_0})$ is constant as long as *M* is large 706 enough (larger than the number of island species), and thus the chosen value of M does 707 not affect the results. 708

709 To decide which information criterion to use to select between different models we compared the performance of the BIC and the Akaike information criterion (AIC). 710 We simulated 1,000 datasets each with models M9 and M19 and then fitted the M9, 711 M14, M17 and M19 models to each of these datasets using two initial sets of starting 712 parameters for each optimisation. We found that for datasets simulated using M9 an 713 incorrect model was preferred using AIC in 10.4 % of cases, but only in 0.11 % of cases 714 when using BIC. For datasets simulated using M19 an incorrect model was preferred 715 12.8 % of cases using AIC and 11.1 % of cases using BIC. We thus compared models 716 using BIC, as this model has lower error rates. 717

An alternative approach to estimating hyperparameters would be to calculate CES rates and their uncertainty independently for each archipelago and to then conduct a meta-analysis of the resulting data, including archipelago area and isolation as predictors. However, errors in parameter estimates will vary, particularly because some

archipelagos have small sample sizes (only a few extant colonisation events, or none at 722 all, e.g. Chagos) and are thus much less informative about underlying process⁵³. Thus, 723 maximizing the likelihood of all data sets together by estimating the hyperparameters 724 (which is precisely our aim) is preferable. For completeness, we present CES rates 725 estimated independently for each archipelago in Supplementary Table 6, excluding 726 archipelagos with fewer than six species and for which we sampled less than 60% of the 727 species in the phylogenies. However, as argued above we do not advocate using these 728 parameter estimates for further analyses because the number of taxa for some of these 729 archipelagos is still low and by excluding archipelagos with fewer than six taxa we 730 cannot capture the lower part of the relationship between area/isolation and CES rates. 731

All DAISIE analyses were run using parallel computation on the high-performance 732 computer clusters of the University of Groningen (Peregrine cluster) and the Museum 733 für Naturkunde Berlin. The new version of the R package DAISIE is available on Github. 734

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Randomization analysis 736

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We conducted a randomization analysis to evaluate whether there is significant signal 738 of a relationship between area and distance and local CES rates in our global dataset. 739 We produced 1,000 datasets with the same phylogenetic data and archipelago ages as 740 the global dataset, but randomly reshuffled archipelago area and D_m in each dataset. We 741 742 then fitted the best *post hoc* model to each of these 1,000 randomized datasets. If the ML estimates of exponent hyperparameters (i.e. the strength of the relationship) in the 743 randomized datasets were non-zero this would indicate that the method is finding 744 evidence for a relationship even if there is none. If, on the other hand, non-zero 745 hyperparameters are estimated in the real data but not in the randomized datasets, this 746 would mean that there is information in the data regarding the putative relationships. 747

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The randomization analysis showed that in global datasets with reshuffled areas and distances the exponent hyperparameters are estimated as zero in most cases, 749 whereas in the empirical global dataset they are not (Extended Data Fig. 3). 750

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A posteriori simulations 752

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We simulated 1,000 phylogenetic global datasets (41 archipelagos each) with the ML 754

hyperparameters of the best *a priori* (M14) and *post hoc* models (M19). We first calculated the local CES rates for each archipelago based on their area and isolation and the hyperparameters for the model, and then used these CES rates as the parameters for the simulations using the DAISIE R package. The simulated data were used to measure bias and accuracy of the method, goodness of fit and the ability of our method to recover observed island biogeographic diversity patterns (see below).

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762 763

Measuring precision and accuracy of method

DAISIE estimates CES rates with high precision and little bias^{10,53}. We conducted 764 parametric bootstrap analyses to assess whether the ability to estimate 765 hyperparameters from global datasets is also good (Extended Data Fig. 2), and to obtain 766 confidence intervals on parameter estimates (Extended Data Table 5). We used DAISIE 767 to estimate hyperparameters from the M14 and M19 simulated datasets (1,000 768 replicates each). We measured precision and accuracy by comparing the distribution of 769 parameters estimated from the 1,000 simulated data set with the real parameters used 770 771 to simulate the same datasets. To check whether ML optimisations of the simulated global datasets converge to the same point in parameter space, we first performed a test 772 on a subset of the simulated data. We ran optimisations with 10 random sets of initial 773 starting values for each of 10 simulated datasets. All optimisations converged to the 774 same likelihood and a very similar hyperparameter set; therefore, we are confident we 775 found the global optimum for each simulated global dataset, even for models with many 776 parameters. 777

778

779 Measuring goodness of fit

780

We measured how well the preferred models fitted the data using different approaches. First, we examined whether our models successfully reproduce the diversity patterns found on individual archipelagos. We calculated the total number of species, cladogenetic species and independent colonisations in each archipelago for each of the 1,000 simulated datasets. We then plotted these metrics versus the observed values in the empirical data (Extended Data Fig. 4 and Fig. 3). Our preferred models have a slight tendency to overpredict species richness when there are a few species and ⁷⁸⁸ underpredict it when there are many. We do not have a clear explanation for this. This ⁷⁸⁹ slight deviation does not seem to be due to an additional area- or distance-dependence, ⁷⁹⁰ so an explanation should be sought in other factors that we did not model. We note that ⁷⁹¹ the fact that all three plots show this tendency rather than just one is to be expected ⁷⁹² because the three metrics of species richness are not entirely independent, with total ⁷⁹³ species richness being the sum of the other two.

Second, we examined whether the models successfully predict the empirical 794 relationships between area, distance and diversity metrics (total species, cladogenetic 795 species, and number of independent colonisations). We fitted generalised linear models 796 (GLM) for each diversity metric, with quasipoisson family errors and log area (or 797 distance) as predictors. We then repeated this across 1000 independent sets of 798 simulated data for the 41 archipelagos and compared the mean of slopes and intercepts 799 for archipelago area and archipelago isolation to the equivalent estimates for the 800 801 empirical data (Fig. 4).

Third, we estimated the pseudo- R^2 of the best model (M19) as a measure of the 802 model's explanatory power. We simulated two independent sets of 10,000 global 803 datasets under M19 model (Set 1 and Set 2). We calculated the mean total number of 804 species, number of cladogenetic species and colonisations for each archipelago across 805 all datasets from Set 1. For each diversity metric we calculated a pseudo-R² (pseudo-R²-806 observed) where the total sum of squares was obtained from the empirical data and the 807 808 residual sum of squares was obtained as the difference between empirical values and expected values (i.e. the simulation means). As the model is inherently stochastic, even 809 if the model is an accurate and complete reflection of the underlying processes then the 810 pseudo- R^2 would tend to be < 1. To estimate the distribution of pseudo- R^2 expected 811 under the model we treated the set 2 simulations as data and estimated the pseudo-R² 812 for each (pseudo-R²-simulated). We then calculated the ratio of the pseudo-R²-observed 813 values over the 10,000 pseudo-R²-simulated values. A ratio approaching 1 would 814 indicate that the model is explaining the observed data as well as the average dataset 815 simulated under this process (Extended Data Fig. 5). 816

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818 Sensitivity to alternative divergence times and tree topologies

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Despite having sampled many new individuals from islands worldwide, given the wide

geographical scale of our study we still rely on sequence data for thousands of 821 individuals submitted to GenBank over the years. Whenever multi-loci analyses 822 including our focal taxa were available we used them, but these are rare (Extended Data 823 Table 4). Therefore, the majority of our phylogenies are based on a small number of 824 genes, and most on a single gene, cyt-b, which is the most widely sequenced 825 mitochondrial marker in birds. Although some studies on island birds have shown that 826 colonisation and diversification times derived from mitochondrial trees often do not 827 differ much from those obtained using multiple loci (e.g. ⁵⁷), it is possible that for some 828 cases the scaling and topologies of the trees might have been more accurate had we 829 used multiple loci⁵⁸. This is particularly relevant for recent island colonists, given 830 incomplete lineage sorting⁵⁹. An additional shortcoming of relying on published 831 sequence data is that many of our DNA alignments often have substantial sections with 832 missing data (e.g. because only one small section of the gene could be sequenced and 833 was uploaded to GenBank), which has been shown to lead to biases in branch lengths 834 and topology⁶⁰. While future studies using phylogenomic approaches may address these 835 issues, obtaining tissue samples for all these taxa will remain an obstacle for a long time. 836

Although DAISIE does not directly use topological information (only divergence 837 times are used), it is possible that the true **topology** for a clade may differ from that of 838 the gene tree we have estimated and this could have an impact on our results by a) 839 affecting colonisation and branching times (addressed in the paragraph below); or b) by 840 841 altering the number of colonisation events. Alternative topologies may have led to an increase or decrease in colonisation events – for instance, some species that appear to 842 have colonised an archipelago only once may have colonised multiple times and if these 843 re-colonisations are recent they may go undetected when using one or few loci. As with 844 any phylogenetic study, we cannot rule out this possibility, but we assume that recent 845 re-colonisation of the archipelagos in our dataset by the same taxon is rare, as these are 846 all oceanic and isolated. For archipelago lineages with cladogenesis (26 out of 502 847 lineages), alternative topologies could include non-monophyly of island radiations, with 848 the corollary being that they would be the result of multiple colonisation events. 849 However, this seems improbable for these isolated and well-studied radiations, for 850 which morphological evidence (e.g. HBW³⁷) is consistent with their monophyly as 851 supported by existing molecular data. 852

Regarding scaling of divergence times, we assessed how uncertainty in our 853 estimated node ages could influence our results by running an analysis of 100 datasets. 854 For each dataset we sampled the node ages (i.e. colonisation and branching times) at 855 random from a uniform distribution centred on the posterior mean for that node in the 856 BEAST tree and extending twice the length of the highest posterior density (HPD) 857 interval. For example, for a node with a 95% HPD interval of 2-3 Ma in our trees, the 858 uniform distribution was set to between 1.5 and 3.5 Ma. The HPD interval will capture 859 uncertainty under the selected phylogenetic and substitution models for the loci we 860 used, but we conduct our sensitivity analysis over a broader interval to accommodate 861 the potential that the selected models and gene trees are inadequate. For cases where 862 using this approach meant that the lower bound of the uniform distribution was lower 863 than 0, we assigned a value of 0.00001 Ma to the lower bound. We fitted the nine best 864 models to the 100 datasets using five initial starting parameters for each model (total 865 4,500 optimisations). We found that parameter estimates across the 100 datasets do 866 not differ strongly from those in the main dataset (Supplementary Table 7). 867 Importantly, model selection was unaffected, with the M19 model being selected for all 868 100 datasets. This is because a lot of the information used for model selection is coming 869 from the other sources of information DAISIE uses (island age, number of species, 870 endemicity status) rather than colonisation/branching times. 871

The ML parameters of the M19 model and the resulting area and isolation dependencies for datasets D1 to D6 - discussed below - are shown in Extended Data Fig. 6 and the DAISIE R objects including these alternative datasets are available in Mendeley Data (https://doi.org/10.17632/sy58zbv3s2.2).

To account for uncertainty in rates of molecular evolution, we repeated all 876 BEAST dating analyses for markers that were not cyt-*b* using 1) the Weir and Schluter⁵⁰ 877 cyt-b rate (dataset D1, equal to main dataset) and 2) marker-specific rates estimated by 878 Lerner et al.⁴¹, which are also widely used in the literature (dataset D2). Although the 879 trees dated using the Lerner et al. rates provide younger ages, we found that the DAISIE 880 results were very similar using either approach (same model preferred and similar 881 parameters). Therefore, in the main text we only discuss the results of analyses of D1, 882 i.e. applying Weir and Schluter's cyt-*b* rate to all markers. 883

For some taxa we did not use the stem age as the estimate of colonisation time, and instead used alternative nodes (see 'Colonisation and branching times' section). To test whether our **choice of nodes** affects our main conclusions, we recoded all such taxa by extracting the stem ages and used these ages as an upper bound for colonisation (DAISIE MaxAge option). We fitted all 28 models to this new dataset (D3) and found that the M19 model is preferred and that the parameters and area/isolation relationships vary only slightly from those of the main analysis. We therefore conclude that our results are robust to the node selection approach.

If extinction has been high on the mainland, or if we failed to sample the 892 closest relatives of the island taxa, this could lead to an overestimation of colonisation 893 times when using the stem age as the precise time of colonisation. To investigate how 894 this could have influenced our results, we ran analyses of datasets where we allow 895 colonisation to have happened at any time since the stem age (i.e. the time of divergence 896 from the nearest relative of the taxon on the mainland). For this we use the DAISIE 897 options Endemic_MaxAge or NonEndemic_MaxAge, which integrate over all possible 898 ages between the given maximum age and the present (or the first branching event 899 within the archipelago for cases where cladogenesis has occurred). We repeated this 900 901 analysis coding all stem ages as maximum ages (D4), or coding only the 25% older stem ages as maximum ages (to account for the fact that on older stems there is potential for 902 there to be more bias) (D5). We also ran analyses on 100 datasets (D6) for which we 903 assigned precise younger ages by randomly selecting a value between the stem age and 904 the present (or crown age for cladogenetic groups). For all these datasets (D4-D6) we 905 906 found that the same model (M19) is preferred, but the initial values of the biogeographical rates (cladogenesis, extinction, colonisation and anagenesis) are 907 estimated to be higher than in the main dataset. Importantly, the exponent 908 hyperparameters are similar to those in the main dataset, meaning that the shape of the 909 relationships between parameters and area/isolation is not much affected (Extended 910 Data Fig. 6). The only exception is perhaps anagenesis, for which the relationships vary 911 more markedly - with isolated islands achieving very high rates for this parameter -, 912 but still agreeing with our main conclusions. Anagenesis is in general the most difficult 913 parameter to estimate⁵³. Thus, our conclusions are robust to the colonisation times 914 potentially being younger than those in our main dataset. 915

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917 Sensitivity to archipelago selection and isolation metrics

The results of the following sensitivity analyses are presented in Supplementary Data 3 919 and the DAISIE R objects including these alternative datasets are available in Mendeley 920 Data (https://doi.org/10.17632/sy58zbv3s2.2). 921

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To test whether the **inclusion of both true archipelagos and single islands** in our dataset could affect the results, we repeated analyses excluding single island units 923 and found that the same model is preferred. The estimated initial rate of cladogenesis 924 (λ^{c_0}) is higher if we exclude single islands, but this parameter is not different from a 925 distribution of parameters estimated from datasets generated using a stratified-random 926 sampling of both archipelagos and single islands. 927

Alternative isolation metrics to D_m have been shown to explain varying and 928 often higher amounts of variation in species richness on islands⁶¹. We tested two 929 alternative metrics: distance to the nearest larger or equivalent-sized landmass (D_b) , 930 and the mean between D_m and D_b (metrics given in Supplementary Data 2). We found 931 that the same DAISIE model with very similar parameters was preferred in both cases, 932 and thus we used only the D_m metric, as this is more similar to the original model of 933 MacArthur & Wilson. 934

The Mascarenes (Mauritius Isl., Reunion and Rodrigues) are often treated as a 935 single biogeographical unit in analyses. We chose to analyse them as independent units 936 because a) the distance between islands is much greater than our threshold for 937 archipelago definition (more than 500 km between Mauritius Isl. and Rodrigues; more 938 939 than 170 km between Reunion and Mauritius Isl.); b) only two species of our target group are shared between the islands (*Terpsiphone bourbonnensis* found in Mauritius 940 Isl. and Reunion; and Psittacula eques found in Mauritius Isl. and extirpated from 941 Reunion), suggesting low connectivity; c) while there are three clades whose branching 942 events took place within the Mascarenes (*Coracina, Pezophaps/Raphus* and *Zosterops*), 943 the remaining species result from independent colonisations suggesting that the three 944 islands behave mostly as three different biogeographical units. We nevertheless ran an 945 analysis treating the islands as a single archipelagic unit and found that the same model 946 was preferred and with similar parameter estimates, and we therefore discuss only the 947 results treating them as separate. 948

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1155

1156 **Data availability**

New sequence data produced for this study have been deposited in GenBank with the 1157 accession codes: MH307408- MH307656. The following datasets have been deposited in 1158 DNA (https://doi.org/10.17632/vf95364vx6.1), Mendeley: alignments 1159 new phylogenetic trees produced for this study (<u>https://doi.org/10.17632/p6hm5w8s3b.2</u>) 1160 and DAISIE R objects (https://doi.org/10.17632/sy58zbv3s2.2). The 11 previously 1161 published trees are available upon request. 1162

1163

1164 **Code availability**

The custom computer code used for this study is freely available in the DAISIE R package (https://github.com/rsetienne/DAISIE).

1167

1168 Acknowledgments

We thank the skilled guides and field assistants who helped with sample collection in the field; and the ornithologists and collection curators who were kind enough to reply to requests for material. For support/advice: Thomas von Rintelen, Kristina von Rintelen, Christine Zorn. For comments on the manuscript: Alex Pigot. For providing samples or DNA sequences: Nancy Bunbury (Seychelles Islands Foundation) who

organized sample loans of Aldabra island; Janske van de Crommenacke, Jim 1174 Groombridge, Hazel Jackson. For sharing data on extinct species: Josep Antoni Alcover, 1175 Juan Carlos Rando, Ferran Sayol, Søren Faurby. For permission to use photographs or 1176 1177 illustrations: Cláudia Baeta, Martijn Hammers, Julian Hume, Dubi Shapiro, Juan Varela, Pedro Cascão. For expertise on island geological ages: Paul Hearty, Robert Stern, Mark 1178 Reagan. For providing phylogenetic data: Alice Cibois, Jimmy McGuire, Heather Lerner, 1179 Petter Marki, Borja Milá, Guillermo Friis, Jérôme Fuchs, John P. Dumbacher, Ore Carmi. 1180 For map data: Patrick Weigelt. For permission to obtain new samples or to access 1181 existing samples, and for logistic support: São Tomé e Príncipe: Arlindo Carvalho and 1182 the Department of the Environment; Equatorial Guinea: José Obiang, Noélia Calvo, the 1183 Universidad Nacional de Guinea Ecuatorial for Bioko and Annobón samples; Seychelles: 1184 the Ministry of Environment, Energy and Climate Change of the Republic of Seychelles, 1185 the Seychelles Bureau of Standards, BirdLife Seychelles, Seychelles Islands Foundation; 1186 Comoros: Centre National de Documentation et de Recherche Scientifique (Grande 1187 Comore & Anjouan), Action Comores, Direction de l'Agriculture et de la Foret (Mayotte); 1188 1189 Madagascar: Ministere des Eaux et Forets (Madagascar), the Madagascar Institute pour la Conservation des Ecosystemes Tropicaux; Mauritius: Mauritius National Parks and 1190 Conservation Service, Mauritius Wildlife Foundation; New Caledonia: Olivier Hébert, 1191 Waifite Waheoneme, Nicholas Clark, the Direction de L'Environment (South Province), 1192 Direction du Développement Economique (Loyalty Islands Province), local chiefs and 1193 1194 landowners; Morocco: Moroccan Environment Ministry; Cape Verde: Cape Verde Agriculture and Environment Ministry; Cameroon: Francis Njie and the Limbe Botanical 1195 and Zoological Garden; Gabon: Station de Recherche de l'IRET at Ipassa-Makokou; 1196 Angola: Fernanda Lages (ISCED-Huíla); Spain: the regional governments of Andalucía 1197 and the Canary Islands; Portugal: regional governments of Madeira and the Azores. 1198 Museum samples: Department of Ornithology and Mammalogy of the California 1199 Academy of Sciences (Laura Wilkinson & Maureen Flannery) for loaning Galápagos 1200 samples; Natural History Museum at Tring (Mark Adams) for loaning Comoros samples; 1201 the Stuttgart State Museum of Natural History for loaning stonechat samples from 1202 Madagascar. Sebastian Block assisted with cluster analyses at the Museum für 1203 Naturkunde. The Center for Information Technology of the University of Groningen 1204 provided support and access to the Peregrine high-performance computing cluster. 1205

L.V. was funded by the German Science Foundation (DFG Research grant VA 1206 1102/1-1), the Alexander von Humboldt Foundation, the Brandenburg Postdoc Prize 1207 2015 and by a VIDI grant from the Netherlands Organisation for Scientific Research 1208 (NWO); R.S.E. by a NWO VICI grant; M.M. by the Portuguese Science and Technology 1209 Foundation (Post-doctoral grant: SFRH/BPD/100614/2014); S.M.C by the National 1210 Geographic Society (CRE grant # 9383-13); J.C.I. by the Spanish Ministry of Science, 1211 Innovation and Universities (Ref.: PGC2018-097575-B-I00) and by a GRUPIN research 1212 grant from the Regional Government of Asturias (Ref.: IDI/2018/000151); C.T. by the 1213 'Laboratoire d'Excellence' TULIP (ANR-10-LABX-41). 1214

1215

1216 Author contributions

L.V., A.B.P. and R.S.E. designed the study, developed the analytical framework and performed statistical analyses. L.V. compiled the data, conducted most of the analyses and wrote the first draft. R.S.E. developed the likelihood method. A.B.P. and R.S.E. contributed substantially to the writing. M.M., B.H.W., S.M.C., J.C.I. and C.T. provided expertise on islands birds and collected bird tissue samples, as well as molecular and/or phylogenetic data. K.H. and J.C.I. performed laboratory work. R.T. contributed to molecular analyses. T.A. performed analyses. All authors commented on the draft.

1224

1225 **Supplementary information** is available for this paper.

1226

1227 **Competing interests** The authors declare no competing interests.

1228

1229 **Correspondence and requests for materials** should be addressed to L.V.

1230 Extended Data Figure Legends

1231Extended Data Figure 1 | Variation of cladogenesis with isolation and area.1232Contour plot showing how the local rate of cladogenesis varies with area and distance1233from the nearest mainland (D_m) assuming the ML global hyperparameters of the M191234model (equations describing the relationships given in Supplementary Table 1).1235Numbers correspond to the archipelago numbers from Fig. 1, and show the local1236cladogenesis rates for each of the archipelagos in our dataset. Area in log scale.

1237

Extended Data Figure 2 | Bootstrap precision estimates of the parameters of the M19 model. Parametric bootstrap analysis fitting the M19 model to 1,000 global data sets simulated with ML parameters of the M19 model. Plots are frequency histograms of estimated parameters. Black lines show the median estimated values across all simulations and the blue lines the simulated values. Dashed lines show 2.5 – 97.5 percentiles. Parameters explained in Supplementary Table 1. Bootstrap parameter estimates for the M14 model are shown in Extended Data Table 5.

1245

Extended Data Figure 3 | Randomization analysis of the M19 model. Distribution of 1246 global hyperparameters estimated from each of 1,000 datasets with the same 1247 phylogenetic data as our main global dataset but randomly reshuffling archipelago area 1248 and isolation among the 41 archipelagos in the data. Grey histograms show DAISIE ML 1249 parameter estimates for the M19 model. Red arrow shows the estimated parameter 1250 from the real data. In the majority of cases, the hyperparameters describing the 1251 exponent of the power models (x, α , β and d_0) are estimated as zero in the reshuffled 1252 datasets, which is not the case in the real data (red). Parameters explained in 1253 Supplementary Table 1. 1254

1255

Extended Data Figure 4 | Goodness of fit of the preferred model (M19). Plots show
 observed total number of species, cladogenetic species and colonisations versus those
 simulated under the model. Median and 95% percentiles shown for 1000 simulations of
 each archipelago. Selected archipelagos are highlighted in colour. Dashed line is y=x. See
 also Fig. 3.

1261

Extended Data Figure 5 | Ratio of pseudo-R²-observed over pseudo-R²-simulated.
 Based on 10,000 datasets simulated under M19 model. A ratio centred on 1 would
 indicate that the model explains the observed data as well as it is able to explain the
 average dataset simulated under the ML parameters.

1266

Extended Data Figure 6 | Sensitivity to colonisation and branching times. a, ML 1267 parameter estimates of the M19 model (preferred model) for datasets differing in 1268 colonisation and branching times. D6 represents 100 datasets, therefore, the 2.5 and 1269 97.5 percentiles are shown. Parameter symbols as in Supplementary Table 1. b, 1270 Estimated relationships between island area and isolation and local island 1271 biogeography parameters for each dataset. Under the M19 model, cladogenesis rate 1272 increases with both area and isolation, and thus plots for more (far, 5,000 km) and less 1273 (near, 50 km) isolated islands are shown. 1274

1275 **Extended Data table titles and footnotes**

1276

1277 Extended Data Table 1 | Archipelago characteristics and references for the island

geological ages. More data in Supplementary Data 2. For archipelagos closer to
Madagascar, New Guinea or New Zealand than to the continent, we use those islands as
the mainland.

1281 Footnote:

- ^{* 34} proposed an age of 0.125 Ma, but we used older age, see Methods.
- 1283 † At least 2 Ma; Paul Hearty pers. comm.
- 1284 **‡** Robert Stern & Mark K. Reagan pers. comm.

1285

1286

1287 **Extended Data Table 2 | Primer sequences used in this study.**

1288 1289

1290 Extended Data Table 3 | The 80 alignments used in the new phylogenetic analyses.

Main source of sequences is GenBank or the new sequences produced for this study, except for the cases noted in the table, where a matrix was directly obtained from a specific study. Details on molecular rates and molecular models applied to each alignment in Supplementary Table 4.

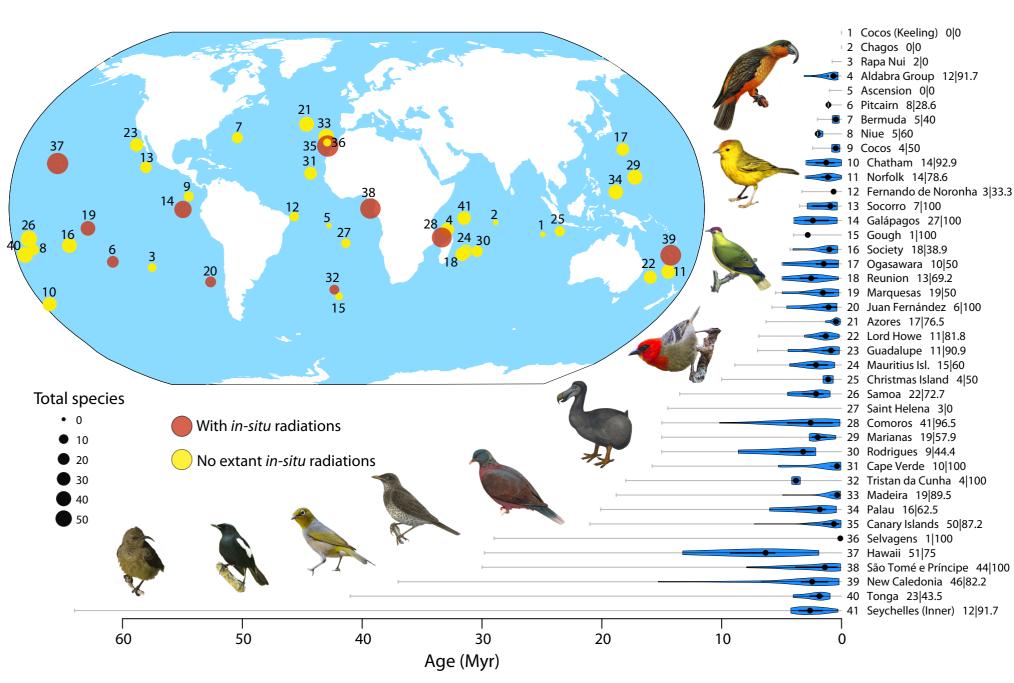
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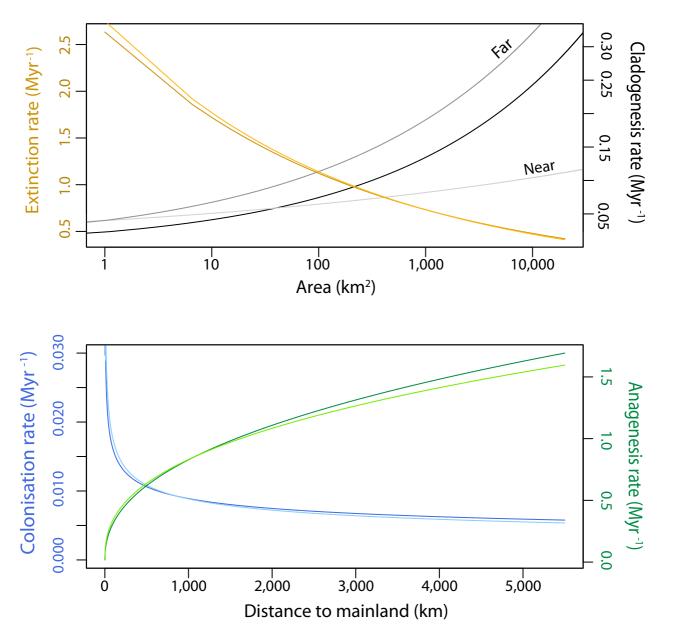
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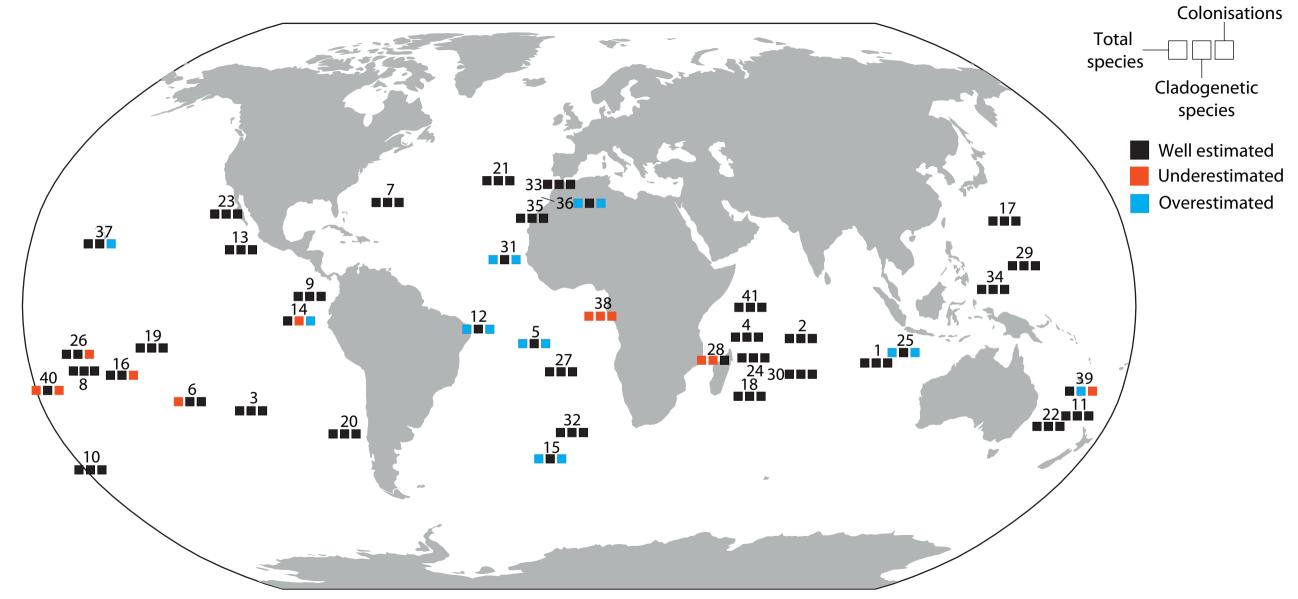
1297 **Extended Data Table 4 | Previously published dated trees used.**

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- 1299

Extended Data Table 5 | Bootstrap of M14 and M19 models. ML estimates and 95%
 confidence intervals of the parameters of the two best models. Confidence intervals
 obtained from the bootstrap analyses. Parameter symbols explained in Supplementary
 Table 1.







Total species

Cladogenetic species

Colonisations

