

# Endemic island songbirds as windows into evolution in small effective population sizes

*The nearly neutral theory in light of passerine birds' genomes*

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## Abstract

Due to their limited ranges and inherent isolation, island species provide a unique opportunity to study the impact of non-adaptive forces on molecular evolution, especially how effective population size may influence the accumulation of deleterious mutations. By estimating nucleotide diversity at synonymous and non-synonymous sites for a large set of in-house and published whole genome sequences covering distant phylogenetic taxa (14 endemic insular and 11 mainland passerine bird species), we found support for contrasted patterns of molecular variation between endemic island and mainland species. Even after controlling for the phylogenetic inertia, we observed significantly lower nucleotide diversities, higher mutation loads and lower adaptive substitution rates in the endemic insular species. Overall, our results indicate that the smaller area available on islands likely constrains the upper bound of the effective population size, in such a way that demography has affected the ability of natural selection to efficiently remove weakly deleterious mutations. Endemic insular species' genomes therefore represent excellent windows into evolution in small effective population sizes and we discussed the implications for both evolutionary and conservation biology.

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**Keywords:** *insularity, neutral theory, molecular evolution, purifying selection, adaptive substitutions, background selection, census population sizes*

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## Introduction

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Endemic island species have long been of central interest for a diverse range of biologists. The study of island organisms stimulated the early study of speciation (Darwin 1859; Mayr 1942) and the development of the island biogeography theory (MacArthur & Wilson 1963; Warren et al. 2015; Valente et al. 2020). Oceanic islands also provide emblematic examples of adaptive changes in response to climate fluctuations (e.g. beak size evolution in the Darwin's finches, Grant and Grant 2006), and adaptive evolutionary radiations (Schluter 2000; Losos 2009; Moyle

et al. 2009; Grant and Grant 2019). Islands also represent areas of particular interests for conservation biologists, because of their high species richness (15-20% of the global species diversity for some plant and animal taxa; e.g. Johnson & Stattersfield 1990, Kreft et al. 2008) but also the more severe consequences of recent anthropogenic disturbances (80% of the reported extinctions since the year 1500, Ricketts et al. 2005). From a molecular evolution perspective, less clear is the long-term impact of genetic drift – and more broadly, non-adaptive processes – on island species evolution. Two main reasons explain why genetic drift can play so crucial a role in insular species. First, oceanic islands are often small and remote, thus providing opportunities for extreme founder effects in colonist populations and isolation impeding gene flow with the initial gene pool after colonization. Second, once a founding population is established, its growth will be mechanically limited by the island size – the smaller the island, the smaller the census population size.

The rate of evolutionary changes caused by genetic drift - which is captured by the effective population size ( $N_e$ ) - is expected to be central for endemic island species. These species are expected to be more prone to an increased inbreeding and enhanced effects of genetic drift (Frankham, 2002). Indeed, island species represent groups of species with lower census and, supposedly, lower  $N_e$  than mainland species. If this hypothesis is true and assuming the nearly neutral theory of molecular evolution (Ohta 1973; 1992), such a lower  $N_e$  is expected to limit the adaptive potential of the endemic island species, because smaller populations produce lower number of mutations per generation at the species level and therefore carry less alleles that may become beneficial after an environmental change (Lanfear et al. 2014, Nam et al 2017, Gossman et al 2012, Rousselle et al. 2020). This is particularly important, because island species are expected to face higher demographic and environmental stochasticity, in such a way that a low proportion of adaptive mutations may limit the capacity to buffer or trade off these variations. In addition,  $N_e$  is also expected to influence the efficacy of natural selection to purge deleterious alleles, with higher effectiveness in large populations assuming the nearly neutral models of genome evolution (Ohta 1973, Charlesworth 2009; Eyre-Walker et al. 2002a; Lanfear et al. 2014). To sum up, following the nearly neutral theory, slightly deleterious alleles are supposed to be accumulated at a higher rate in endemic island species as compared to their mainland relatives, leading to an increasing deleterious mutational burden over time. Taxon-specific evidences

supporting this important burden in island species start to accumulate (Loire et al. 2013 for Giant Galápagos tortoises; Rogers and Slatkin 2017 for woolly mammoths; Robinson et al. 2016; 2018 for island foxes and Kutschera et al. 2019 for crows), but the strength of such a general relationship between census and effective population sizes remains debated. For example, James et al. (2016) found no effect of island colonization on mitochondrial molecular evolution. By sequencing the transcriptome of a broad taxonomic sample of animals, Romiguier et al. (2014) found no relationship between range size and the level of genetic diversity. More recently, Díez-del-Molino et al. (2018) found a positive, albeit weak, correlation between census size and genetic diversity in birds and mammals. The absence of a general conclusion was one of our main motivations for this study.

Despite its central importance, estimating  $N_e$  is challenging in practice.  $N_e$  is indeed well-known to fluctuate rapidly over short periods of time, in such a way that the current size may not reflect past population size that has shaped present-day genetic diversity (e.g. Atkinson et al. 2008; Schiffels & Durbin, 2014 for humans). As a consequence, depending on the evolutionary time scale under consideration, the  $N_e$  can be very different. For example, the  $N_e$  that have determined the dynamics of ancestral alleles detected using between-species (divergence) data might be several orders of magnitude different from recent  $N_e$  estimates using present-day within-species (polymorphism) data (Eyre-Walker 2002b, Rousselle et al. 2018). In animals or plants, life-history traits such as body mass or longevity are known to provide useful proxies for long-term  $N_e$  (Popadin et al. 2007, Romiguier et al. 2014, Figuet et al. 2016, Chen et al. 2017). In birds, the same life-history traits also represent proxies of  $N_e$ , but only after taking into account some confounding effects, suggesting that the relationship between  $N_e$  and life-history is weaker for birds than for some other clades (Weber et al. 2014, Botero et al. 2017, Mugal et al. 2020). However, there is a fundamental limitation to the use of life-history traits as a proxy of  $N_e$ . The relationship between these two variables is only poorly described and only holds as a result of the correlations between body size and population density (White et al. 2007). Long-term  $N_e$  can be more directly estimated using population genomic variables. The level of genetic drift is inversely proportional to  $N_e$  (Kimura & Crow, 1964; Kimura et al. 1983) and consequently, the observed nucleotide diversity levels at synonymous sites ( $\pi_s$ ) can be used as a predictor of the long-term  $N_e$  (e.g. Romiguier et al. 2014).

The impact of  $N_e$  on the efficacy of positive selection still remains highly debated among evolutionary biologists (e.g. Kern & Hahn, 2018 and Jensen et al., 2019 for recent publications), despite the significant progress in both the methods and the fundamental knowledge gained in this field over the last decade (e.g. Nikolaev et al. 2007; Keightley and Eyre-Walker 2010; Gossmann et al. 2010; 2012; Galtier 2016; Chen et al. 2017; Rousselle et al. 2020). By combining information at both the within-species (polymorphism) and between-species (divergence) levels, the proportion of adaptive substitutions can be estimated based on the comparisons of the ratios of non-synonymous and synonymous mutations in the polymorphism and substitution data ( $\pi_N/\pi_S$  &  $d_N/d_S$ , McDonald Kreitman, 1991). Over the last decade, several methods inspired by the seminal work of McDonald and Kreitman (1991) have been developed to take into account short-term demographic variation and the presence of slightly deleterious mutations. These methods used the Site Frequency Spectra (SFS) at both synonymous and non-synonymous sites to estimate the Distribution of Fitness Effects (DFE) of non-synonymous mutations (Keightley & Eyre-Walker 2007; Eyre-Walker & Keightley 2009; Galtier 2016; Tataru et al. 2017; see also Moutinho et al. 2019 for a review). Using these methods, can we expect different proportions of adaptive mutations between island and mainland species? On the one hand, we can hypothesize that the population with the larger effective population sizes will also exhibit the higher rates of adaptive evolution, because of a greater number of *de novo* mutations produced per generation and because of the greater standing genetic variability available. On the other hand, assuming that the species with small  $N_e$  exhibits a higher proportion of deleterious mutations, therefore generating proteins that contribute to pulling away from their fitness optima, we can hypothesize the production of more frequent compensatory adaptive mutations in these species.

Passerine birds represent an excellent group of species to investigate the effect of insularity on evolutionary dynamic for a series of reasons. First, a lot of genomic resources are available for passerine birds. Since the first genome in 2010 (Warren et al. 2010), a lot of passerine bird species have recently joined the list (e.g. Ellegren et al. 2012; Zhang et al. 2012; Cornetti et al. 2015; Laine et al. 2016; Lundberg et al. 2017; Leroy et al. 2019). Beyond genome assemblies, several excellent population genomics studies have focused on passerine birds and released the data in the public domain (e.g. Ellegren et al. 2012; Lamichhaney et al. 2015 and Lundberg et al. 2017 among others). Second, the large community of bird-enthusiasts, including ornithologists but also numerous voluntary birdwatchers, provides excellent monitoring of the

species presence, distribution and abundance. Third, probably thanks to their good dispersal capability, a quite large diversity of birds is endemic to islands (17% of the world's bird species, Johnson & Stattersfield 1990). It is important to mention that endemic islands birds have however experienced a dramatic species loss over the last five centuries (Ricketts et al. 2005). Islands habitats are indeed more threatened habitats than continents, especially given the deleterious consequences of human activities and disturbances (Steadman 1995). Fourth, both the chromosome architectures and the recombination landscapes are stable in birds (Ellegren 2010; Singhal et al. 2015). Unlike mammals, bird genomes lack the *PRDM9* gene (Baker et al. 2017). In mammals, this *PRDM9* gene indeed explains the rapid turnover in recombination hotspots (Baudat et al. 2010; Parvanov et al. 2010). As a consequence, the correlation between the recombination rates and the G+C content is stronger than in mammal species for instance. This relationship is due to the GC-biased gene conversion (gBGC), a recombination-associated segregation bias that favors G and C over A and T alleles (for details, see Duret & Galtier, 2009). In birds, the genomic GC content, or more specifically GC-content at third codon position (hereafter GC3) for coding regions, are therefore a good predictor of the local recombination rate. For all these reasons, we have therefore chosen to focus on Passerida -- a clade of passerines -- species to evaluate the molecular features of insularity. Passerida is a clade of songbirds that originated 20 to 30 Million years ago (Baker et al. 2004; Prum et al. 2015, Oliveros et al. 2019) and represents the most species-rich avian clade. In this study, we however considered species with relatively similar body-mass, longevity and clutch-size to take into account some confounding factors (see Materials and Methods).

In a nutshell, we combined in-house (n=89) and published (n=206) whole genome sequences of 25 passerine bird species, including 14 endemic insular species, to estimate parameters of molecular evolution. More precisely, we investigated the following questions regarding the use of endemic insular species for evaluating predictions from the nearly neutral theory: (i) Do endemic insular species exhibit genomic signatures consistent with evolution under lower long-term effective sizes as compared to widely-distributed mainland species? (ii) Are high deleterious mutation loads a common feature of insularity? (iii) Is there any covariation between the deleterious mutation loads and heterogeneities in recombination rate? (iv) Is the proportion of adaptive substitutions significantly lower or higher for the insular species? (v) How informative these statistics are in a conservation biology context?

# Results

## Genomic signatures of lower long-term $N_e$ in island species

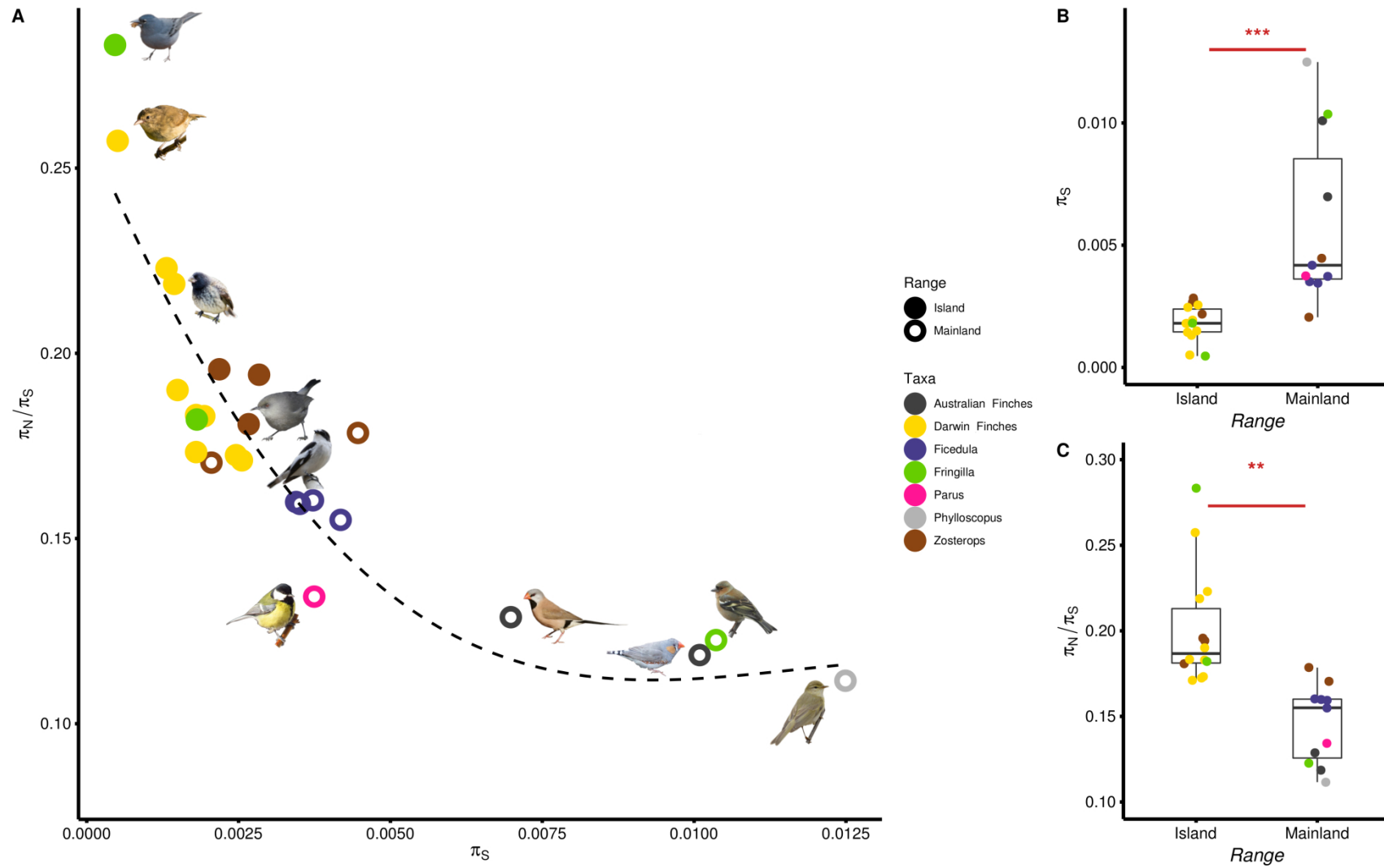
Using a combination of in-house (n=89) and published (n=206) whole genome sequences of a set of 25 focal species consisting of 14 insular and 11 mainland species (Table 1, Fig. S1), we estimated the nucleotide diversity at synonymous ( $\pi_S$ ) and at non-synonymous sites ( $\pi_N$ ) for each species at 6,499 orthologous genes on average (range: 5,018-7,514, among 8,253 orthogroups).  $\pi_S$  varies from 0.07% in the Tenerife blue chaffinch (*Fringilla teydea*) to 1.26% in the willow warbler (*Phylloscopus trochilus*) representing a 17-fold difference (Table S1). These species also have the highest and lowest  $\pi_N/\pi_S$  with 0.285 and 0.120 for the Tenerife blue chaffinch and the willow warbler respectively (Table S1).

We recovered a strong negative correlation between the ratio of non-synonymous to synonymous nucleotide diversities ( $\pi_N/\pi_S$ ) and  $\pi_S$  (Fig.1A, Spearman's  $\rho = -0.879$ , p-value= $2.2 \times 10^{-06}$ , see also Figs. S2-S4). Endemic insular species exhibit significantly less mean  $\pi_S$  than mainland species with only one mainland species (*Zosterops pallidus*) overlapping with the island species (mean  $\pi_S = 0.59\%$  and  $0.18\%$  for mainland and island species respectively, Fig.1 B; Welch t-test = -3.82; p-value=0.003). Similarly, species endemic to island have a  $\pi_N/\pi_S$  40% higher on average than mainland species (Fig. 1 C; mean  $\pi_N/\pi_S = 0.145$  and  $0.201$  for mainland and island species respectively, Welch t-test = 4.86; p-value= $6.8 \times 10^{-5}$ ).

**Table 1: Sequencing data used in this study.** The abbreviation in parenthesis following Darwin's finches names indicate the island of origin (C=Coco, E=Española, L=San Cristobal, P=Pinta, S=Santiago, W=Wolf, Z=Santa Cruz).

	Species	Clade	Range	Individuals	Data
1	<i>Certhidea olivacea</i> (S)	Darwin's finches	Island	5	Lamichhaney <i>et al.</i> 2015
2	<i>Certhidea fusca</i> (E)	Darwin's finches	Island	10	Lamichhaney <i>et al.</i> 2015
3	<i>Certhidea fusca</i> (L)	Darwin's finches	Island	10	Lamichhaney <i>et al.</i> 2015
4	<i>Platypiza crassirostris</i> (Z)	Darwin's finches	Island	5	Lamichhaney <i>et al.</i> 2015
5	<i>Camarhynchus pallidus</i> (Z)	Darwin's finches	Island	5	Lamichhaney <i>et al.</i> 2015
6	<i>Pinaroloxias inornata</i> (C)	Darwin's finches	Island	8	Lamichhaney <i>et al.</i> 2015
7	<i>Geospiza difficilis</i> (P)	Darwin's finches	Island	10	Lamichhaney <i>et al.</i> 2015
8	<i>Geospiza septentrionalis</i> (W)	Darwin's finches	Island	8	Lamichhaney <i>et al.</i> 2015
9	<i>Geospiza conirostris</i> (E)	Darwin's finches	Island	10	Lamichhaney <i>et al.</i> 2015
10	<i>Ficedula albicollis</i>	Ficedula flycatchers	Mainland	20	Ellegren <i>et al.</i> 2012
11	<i>Ficedula hypoleuca</i>	Ficedula flycatchers	Mainland	20	Ellegren <i>et al.</i> 2012
12	<i>Ficedula speculigera</i>	Ficedula flycatchers	Mainland	20	Ellegren <i>et al.</i> 2012
13	<i>Ficedula semitorquata</i>	Ficedula flycatchers	Mainland	20	Ellegren <i>et al.</i> 2012
				6	Bourgeois <i>et al.</i> 2017
14	<i>Zosterops borbonicus</i>	White-eyes	Island	1	Leroy <i>et al.</i> 2019
				18	This study
15	<i>Zosterops olivaceus</i>	White-eyes	Island	15	This study
16	<i>Zosterops mauritanus</i>	White-eyes	Island	9	This study
17	<i>Zosterops pallidus</i>	White-eyes	Mainland	2	This study
18	<i>Zosterops virens</i>	White-eyes	Mainland	11	This study
19	<i>Fringilla coelebs</i>	Chaffinches	Mainland	15	This study
20	<i>Fringilla coelebs palmae</i>	Chaffinches	Island	9	This study
21	<i>Fringilla teydea</i>	Chaffinches	Island	10	This study
22	<i>Taeniopygia guttata castanotis</i>	Estrildidae	Mainland	19	Singhal <i>et al.</i> 2015
23	<i>Poephila acuticauda acuticauda</i>	Estrildidae	Mainland	10	Singhal <i>et al.</i> 2015
24	<i>Parus major</i>	Paridae	Mainland	10	Corcoran <i>et al.</i> 2017
25	<i>Phylloscopus trochilus</i>	Phylloscopidae	Mainland	9	Lundberg <i>et al.</i> 2017





**Fig.1: Island species as models for evolution in small effective population sizes.** A. Local polynomial regression (LOESS with  $\text{span}=1.25$ ) between the ratio of nonsynonymous to synonymous nucleotide diversity ( $\pi_N/\pi_S$ ) and the levels of nucleotide diversity ( $\pi_S$ ), used as an indicator of effective population sizes. See also Fig. S3 for a log-transformed regression. B & C. Variation in nucleotide diversity ( $\pi_S$ , B) and  $\pi_N/\pi_S$  between the island endemic and the mainland species (C). Photo credits: F. Desmoulins, E. Giacone, G. Lasley, Lianaj, Y. Lyubchenko, B. Nabholz, J.D. Reynolds, K.Samodurov, A.Sarkisyan (iNaturalist.org); M. Gabrielli (personal communication).

## 198 **Linking census, effective population size and conservation status**

199 Current census population size and the size of the geographical range size in square  
200 kilometers also significantly correlate with both  $\pi_S$  and  $\pi_N/\pi_S$  (Fig.2). Median estimates for the  
201 current census population sizes (only available for 13 of our 25 focal species) are positively  
202 correlated with  $\pi_S$  and negatively with  $\pi_N/\pi_S$  ( $R^2 = 0.63$ , p-value=0.001 and  $R^2 = 0.66$ , p-value= $6.9 \times$   
203  $10^{-04}$ , respectively; Fig. 2 A & B). Similarly, geographic range size is also significantly correlated  
204 with log<sub>10</sub>-transformed  $\pi_S$  and  $\pi_N/\pi_S$  (Fig.2 C & D, table 2,  $R^2 = 0.54$ , p-value= $3.1 \times 10^{-05}$  and  $R^2 =$   
205  $0.51$ , p-value= $6.1 \times 10^{-05}$ , respectively). In contrast, IUCN red list assessment has a weak effect on  
206  $\pi_S$  and  $\pi_N/\pi_S$  (Table 2,  $R^2 = 0.20$ , p-value=0.03 and  $R^2 = 0.18$ , p-value=0.05, respectively). Census  
207 population size, geographical range, IUCN assessment and insularity (*i.e.* island/mainland status)  
208 are all cross-correlated. We use linear model analyses to evaluate the relative contribution of  
209 insularity in comparison to the other explanatory variables to  $\pi_S$  and  $\pi_N/\pi_S$  variations (Table 2).  
210 Following the minimum AIC criteria, the best model is census size or range size for explaining  $\pi_S$   
211 but models considering insularity exhibit close AIC values ( $\Delta AIC < 2$ ). Insularity is, however, the  
212 best model to explain  $\pi_N/\pi_S$  (Table 2). Linear models combining insularity and range size or census  
213 size always have lower AIC than models with one explanatory variable (Table 2) demonstrating  
214 that the effect of insularity is indeed linked to population size variation and that there is no  
215 intersect effect of insularity *per se*. In contrast, IUCN assessment has a weak effect, and this  
216 effect completely disappears in combination with insularity (Table 2). Even after explicitly  
217 controlling our statistical analyses for the phylogenetic relationship between species using a  
218 Phylogenetic Generalized Least Squared (PGLS) method with a Brownian motion model of  
219 character evolution (Grafen 1989; Paradis 2012), the significant effect of the island/mainland  
220 status, range size and census size on  $\pi_S$  and  $\pi_N/\pi_S$  remain highly significant (Table 2). Our dataset  
221 indeed contains species with at least 4 independent mainland-island transitions (Fig. S1), thus  
222 ensuring sufficient statistical power to support our interpretations.

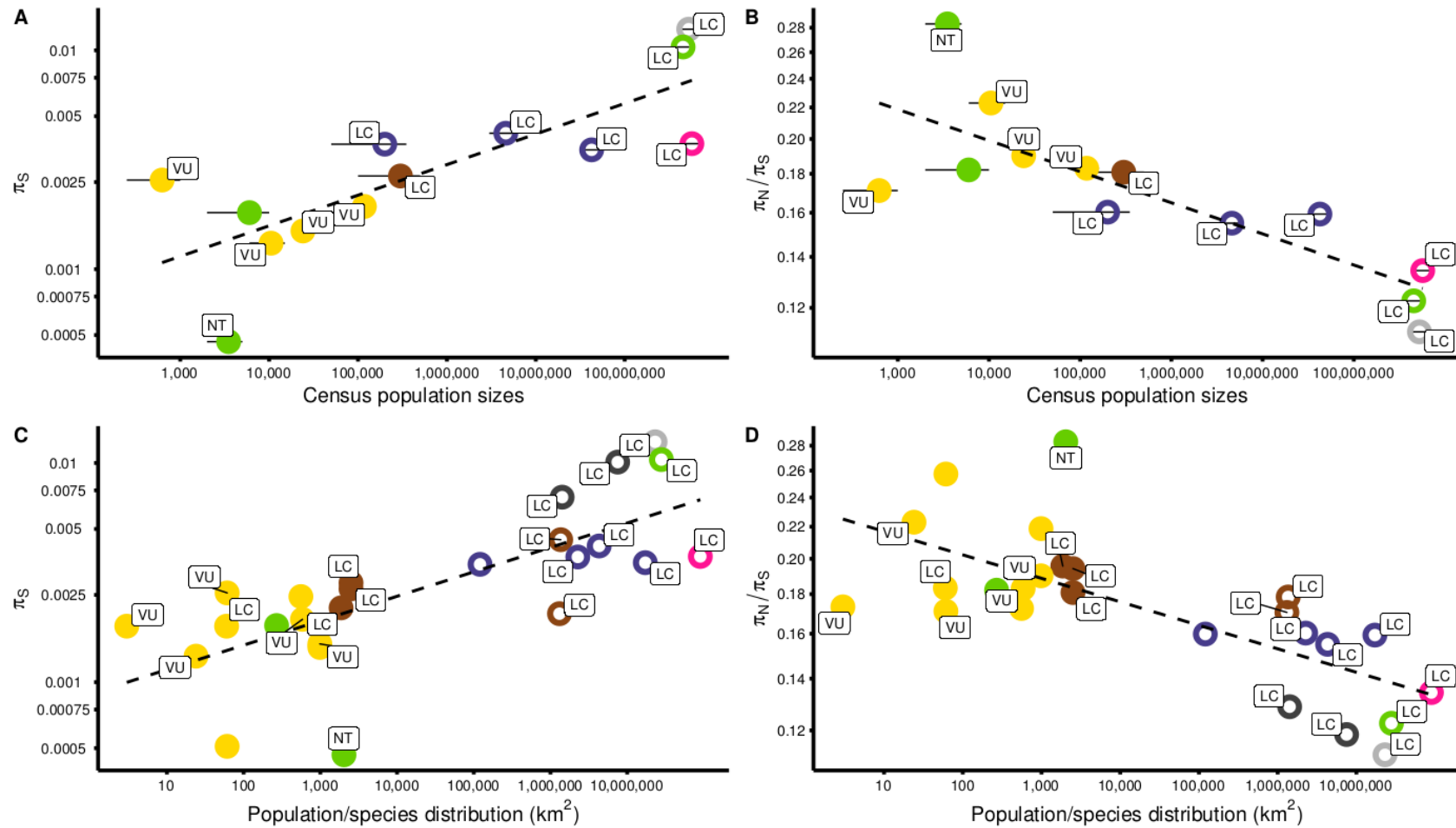
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**Table 2: Linear models and Phylogenetic Generalized Least Squared (PGLS) method-based tests performed.**

To perform accurate comparisons, model selections were performed using the maximum number of species (N) for which all information was available (i.e. 13 species with census size estimates, 21 species with IUCN status, and 25 with both insularity status & range sizes). IUCN status was considered as a binary variable “Threatened”, when the status is “Vulnerable”, or “Non-threatened” otherwise (i.e. “Least Concerned or Near-threatened”). For each comparison, the best model is indicated in bold. For the models with PGLS, the p-value of the model was obtained from a Likelihood ratio test using the *anova.lme* function (*nlme* package, Pinheiro et al. 2020).

Type	Model	N	R <sup>2</sup>	p-value model	p-value variable 1	p-value variable 2	AIC	ΔAIC
Linear	$\log_{10}(\pi_S) \sim \text{Insularity}$	13	0.577	0.00259			5.01	1.72
Linear	<b><math>\log_{10}(\pi_S) \sim \log_{10}(\text{census\_size})</math></b>	13	0.630	<b>0.00121</b>			<b>3.29</b>	<b>0</b>
Linear	$\log_{10}(\pi_S) \sim \log_{10}(\text{census\_size}) + \text{Insularity}$	13	<b>0.656</b>	0.00485	0.162	0.405	4.34	1.05
Linear	$\log_{10}(\pi_S) \sim \text{IUCN}$	21	0.133	0.104			16.33	13.62
Linear	<b><math>\log_{10}(\pi_S) \sim \text{Insularity}</math></b>	21	0.547	<b>0.000128</b>			<b>2.71</b>	<b>0</b>
Linear	$\log_{10}(\pi_S) \sim \text{IUCN} + \text{Insularity}$	21	<b>0.548</b>	0.000787	0.83	0.000727	4.66	1.95
Linear	$\log_{10}(\pi_S) \sim \text{Insularity}$	25	0.528	0.000039			4.45	0.52
Linear	<b><math>\log_{10}(\pi_S) \sim \log_{10}(\text{range\_km}^2)</math></b>	25	0.537	<b>0.000031</b>			<b>3.93</b>	<b>0</b>
Linear	$\log(\pi_S) \sim \text{Insularity} + \log(\text{range\_km}^2)$	25	<b>0.552</b>	0.000145	0.398	0.282	5.1	1.17
Linear	$\log_{10}(\pi_N/\pi_S) \sim \text{Insularity}$	13	0.577	0.00258			-27.68	3.01
Linear	<b><math>\log_{10}(\pi_N/\pi_S) \sim \log_{10}(\text{census\_size})</math></b>	13	0.665	<b>0.00068</b>			<b>-30.69</b>	<b>0</b>
Linear	$\log_{10}(\pi_N/\pi_S) \sim \log_{10}(\text{census\_size}) + \text{Insularity}$	13	<b>0.681</b>	0.00331	0.102	0.496	-29.33	1.36
Linear	<b><math>\log_{10}(\pi_N/\pi_S) \sim \text{Insularity}</math></b>	<b>21</b>	<b>0.552</b>	<b>0.000113</b>			<b>-49.67</b>	<b>0</b>
Linear	$\log_{10}(\pi_N/\pi_S) \sim \text{IUCN}$	21	0.067	0.256			-34.26	15.41
Linear	$\log_{10}(\pi_N/\pi_S) \sim \text{IUCN} + \text{Insularity}$	21	<b>0.578</b>	0.000421	0.305	0.00019	-48.93	0.74
Linear	<b><math>\log_{10}(\pi_N/\pi_S) \sim \text{Insularity}</math></b>	25	0.529	<b>0.000038</b>			<b>-59.4</b>	<b>0</b>

Linear	$\log_{10}(\pi_N/\pi_S) \sim \log_{10}(\text{range\_km}^2)$	25	0.513	0.000057			-58.36	1.04
Linear	$\log_{10}(\pi_N/\pi_S) \sim \text{Insularity} + \log_{10}(\text{range\_km}^2)$	25	<b>0.541</b>	0.00019	0.254	0.461	-58.03	1.37
PGLS	<b><math>\log_{10}(\pi_S) \sim \text{Insularity}</math></b>	<b>13</b>		<b>0.0001</b>	<b>0.0005</b>		<b>8</b>	<b>0</b>
PGLS	$\log_{10}(\pi_S) \sim \log_{10}(\text{census\_size})$	13		0.0153	0.029		17.25	9.25
PGLS	$\log_{10}(\pi_S) \sim \log_{10}(\text{census\_size}) + \text{Insularity}$	13		0.0003	0.365	0.0058	8.87	0.87
PGLS	$\log_{10}(\pi_S) \sim \text{Insularity}$	25		0.0053	0.0081		20.38	2.16
PGLS	<b><math>\log_{10}(\pi_S) \sim \log_{10}(\text{range\_km}^2)</math></b>	<b>25</b>		<b>0.0016</b>	<b>0.0028</b>		<b>18.22</b>	<b>0</b>
PGLS	$\log_{10}(\pi_S) \sim \log_{10}(\text{range\_km}^2) + \text{Insularity}$	25		0.0049	0.117	0.458	19.51	1.29
PGLS	<b><math>\log_{10}(\pi_S) \sim \text{Insularity}</math></b>	<b>21</b>		<b>0.0066</b>	<b>0.0107</b>		<b>9.42</b>	<b>0</b>
PGLS	$\log_{10}(\pi_S) \sim \text{IUCN}$	21		0.942	0.946		16.8	7.38
PGLS	$\log_{10}(\pi_S) \sim \text{IUCN} + \text{Insularity}$	21		0.024	0.804	0.0127	11.35	1.93
PGLS	<b><math>\log_{10}(\pi_N/\pi_S) \sim \text{Insularity}</math></b>	<b>13</b>		<b>0.0006</b>	<b>0.0019</b>		<b>-22.93</b>	<b>0</b>
PGLS	$\log_{10}(\pi_N/\pi_S) \sim \log_{10}(\text{census\_size})$	13		0.021	0.0378		-16.38	6.55
PGLS	$\log_{10}(\pi_N/\pi_S) \sim \log_{10}(\text{census\_size}) + \text{Insularity}$	13		0.0021	0.557	0.0233	-21.4	1.53
PGLS	<b><math>\log_{10}(\pi_N/\pi_S) \sim \text{Insularity}</math></b>	<b>25</b>		<b>0.0046</b>	<b>0.0071</b>		<b>-46.96</b>	<b>0</b>
PGLS	$\log_{10}(\pi_N/\pi_S) \sim \log_{10}(\text{range\_km}^2)$	25		0.0089	0.0131		-45.76	1.2
PGLS	$\log_{10}(\pi_N/\pi_S) \sim \log_{10}(\text{range\_km}^2) + \text{Insularity}$	25		0.0121	0.41	0.191	-45.75	1.21
PGLS	<b><math>\log_{10}(\pi_N/\pi_S) \sim \text{Insularity}</math></b>	<b>21</b>		<b>0.0027</b>	<b>0.0049</b>		<b>-45.98</b>	<b>0</b>
PGLS	$\log_{10}(\pi_N/\pi_S) \sim \text{IUCN}$	21		0.667	0.686		-37.17	8.81
PGLS	$\log_{10}(\pi_N/\pi_S) \sim \text{IUCN} + \text{Insularity}$	21		0.0085	0.496	0.0052	-44.53	1.45

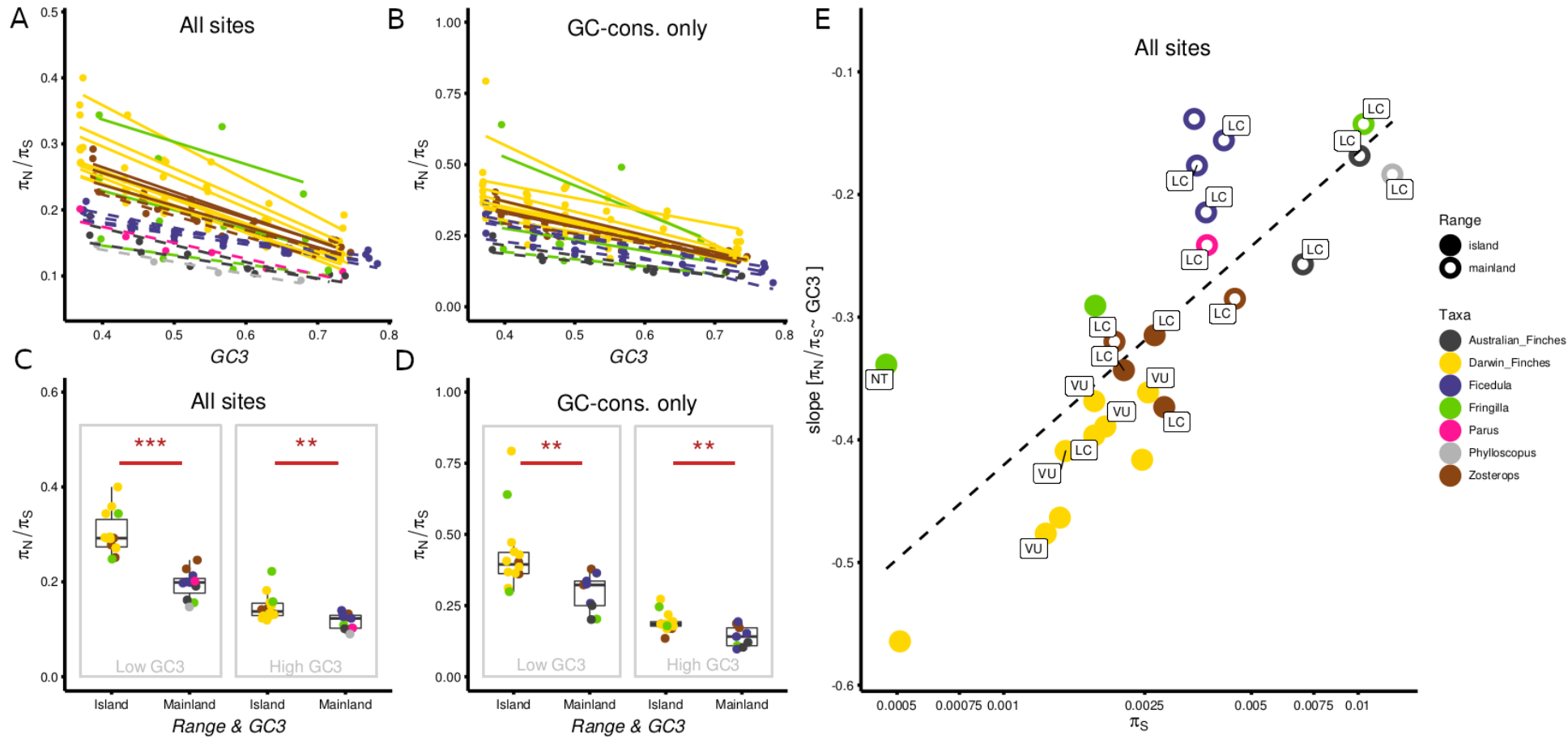


**Fig. 2: Ecological-evolutionary correlations based on the variables investigated in this study.**  $\pi_S$  and  $\pi_N/\pi_S$  are used as proxies of  $N_e$  and the efficacy of natural selection to remove deleterious variants and are correlated with both the median estimates for the current census population sizes (A & B) and the geographical range sizes (C & D). Both ecological and evolutionary parameters are log-transformed. Filled and open dots represent the island and the mainland species, respectively (Fig. 1 for details). Only the 13 species with estimates of the current census population sizes are included for the panels A & B (with ranges shown with a thin black line). Where known, the IUCN conservation status of the investigated species is indicated (LC = least concerned, NT = near threatened, VU = vulnerable).

## **Within-genome variation in the efficacy of purifying selection**

In birds, the GC-content at third codon position (GC3) is known to be highly correlated with recombination rates (Hillier et al. 2004; Backström et al. 2010; Singhal et al. 2015). After ordering genes according to GC3 (see Materials and Methods), we observed negative slopes between ( $\pi_N/\pi_S$ ) and GC3 (Fig. 3.A, B). These results were equally supported when using the estimates of  $\pi_N/\pi_S$  based on the method described in Rousselle and collaborators (2020) which considers the site frequency spectra at both non-synonymous and synonymous (method 2 in Fig. S5). The relationship between  $\pi_N/\pi_S$  and GC3 is therefore supported, and consistent with an increased efficiency of natural selection with higher local recombination rate.

By comparing sets of genes exhibiting the lowest and highest GC3 (2Mb of total concatenated sequences used for the computations), we reported significant differences in mean  $\pi_N/\pi_S$  between island and mainland species in both GC-rich ( $t=3.316$ ,  $p=0.003$ ) and GC-poor ( $t=7.300$ ,  $p=2.1 \times 10^{-07}$ ) regions, but with more marked  $\pi_N/\pi_S$  differences in genes exhibiting low GC3 ( $\Delta \text{mean}_{\text{insular vs. mainland}} = 0.107$ , 95%CI: 0.077-0.194) than in those exhibiting high GC3 ( $\Delta \text{mean} = 0.029$ , 95%CI: 0.011-0.048) (Fig. 3C). The genetic advantage of recombination, which is captured by the slope between the  $\pi_N/\pi_S$  and the GC3 is itself highly correlated with the levels of nucleotide diversity (Fig. 3E), consistent with a growing importance of the local recombination rate for the efficiency of purifying selection with respect to the effective population size, which is especially important for the endemic island species (filled dots, Fig. 3E). The stronger effect of recombination for island species is captured by the significant interaction between GC3 and insularity in the linear model :  $\pi_N/\pi_S \sim \text{GC3} + \text{insularity} + \text{GC3}:\text{insularity}$  ( $R^2 = 0.72$ ,  $p_{\text{model}} < 2.2 \times 10^{-16}$ ,  $p_{\text{GC3}} < 2.2 \times 10^{-16}$ ,  $p_{\text{insularity}} = 2.8 \times 10^{-10}$ ,  $p_{\text{interaction}} = 1.33 \times 10^{-05}$ ).



**Fig.3: Efficiency of natural selection of insular and mainland passerine bird species to purge potentially deleterious mutations depending on the local recombination rates.** A & B. Linear regressions between  $\pi_N/\pi_S$  ratios and the third codon position (GC3), a proxy of local recombination rates for all sites (A) and GC-conservative sites only (B). C & D. Observed variation in  $\pi_N/\pi_S$  ratios at GC-rich and GC-poor genes between island and mainland species for all sites (C) and for GC-conservative sites only (D). E. Slopes of the linear regressions between  $\pi_N/\pi_S$  and GC3 (i.e. as shown in the panel A) depending on the synonymous nucleotide diversity. The dotted line indicates the best-fitting linear model.

To take into account the potential impact of gBGC on the estimations of  $\pi_S$  and  $\pi_N$ , we used the same approach than in Rousselle et al. 2018 and computed  $\pi_N/\pi_S$  based on the Site Frequency Spectrum (SFS) computed for GC-conservative sites only (i.e., A $\leftrightarrow$ T and G $\leftrightarrow$ C mutations), a part of the total variants which is unaffected by gBGC. Interestingly, we also recovered significant differences in  $\pi_N/\pi_S$  between island and mainland species at GC-conservative sites (Fig. 3D, see also Fig. S6), albeit slightly less significant (GC-poor:  $\Delta\text{mean}=0.138$  with 95%CI: 0.050-0.225,  $t=3.284$ ,  $p=0.004$ ; GC-rich:  $\Delta\text{mean}=0.050$  with 95%CI: 0.018-0.082,  $t=3.284$ ,  $p=0.005$ ). The evolutionary advantage of genetic recombination is higher in the island (GC-conservative  $\pi_N/\pi_S$ ,  $\Delta\text{mean}_{\text{GC-rich vs. GC-poor}}=0.239$ ) than in the mainland species ( $\Delta\text{mean}=0.151$ ), which seems to imply that recombination more efficiently limits the mutation burden of insular species. However, this advantage is still insufficient to entirely compensate for the higher fixation of deleterious alleles due to the increased drift effects in species in small populations, as typically represented by the island passerine species used in our study.

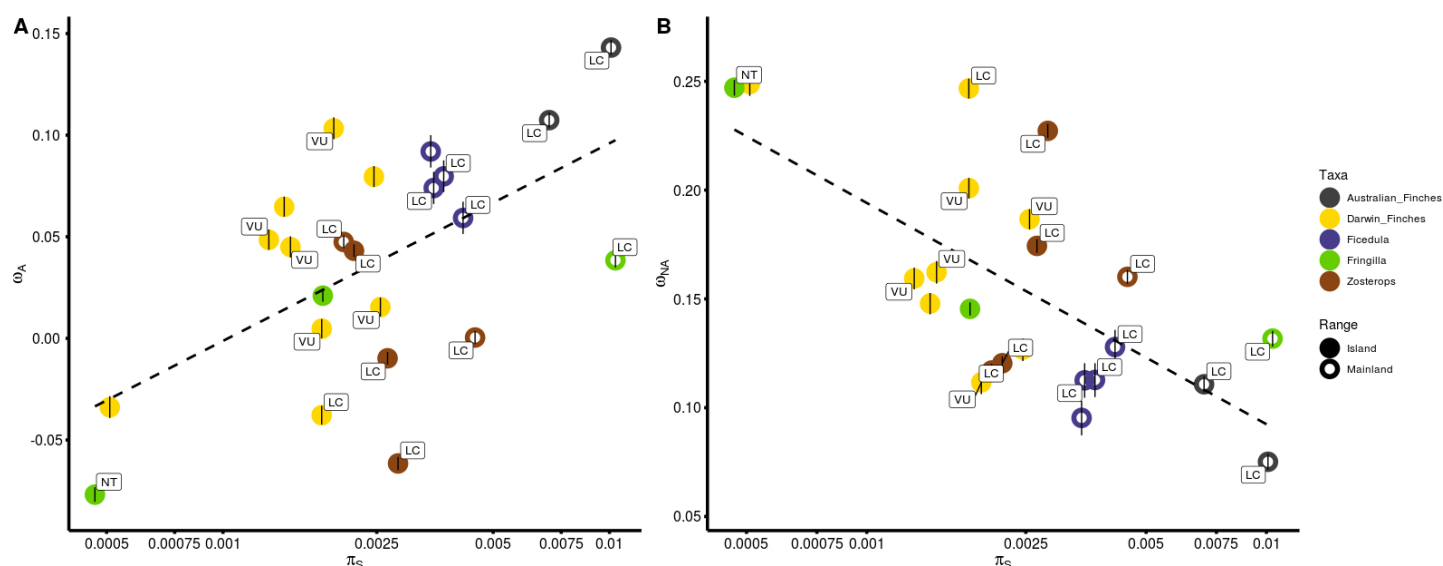
## **The double penalty of insularity: fewer advantageous & more weakly deleterious substitutions**

For each species, we derived the expected ratio of non-adaptive non-synonymous to synonymous substitutions between species (ie,  $\omega_{NA}$ : the rate of non-adaptive substitution; Galtier 2016) using the unfolded SFS at both synonymous and nonsynonymous sites. We then compared this expectation to the observed  $d_N/d_S$  ( $\omega$ ) as estimated based on the sequence divergence with an outgroup species (for a list, see Table S3), obtaining the proportion of amino-acid substitutions that result from positive selection ( $\alpha$ ). After scaling this proportion by the observed number of synonymous substitutions, we estimated  $\omega_A$ , the rate of adaptive substitutions relative to neutral divergence. In other words, we used the following equations:  $d_N/d_S = \omega_A + \omega_{NA}$  and  $\alpha = d_A/d_N$ , where  $d_A$  is the adaptive non-synonymous divergence.

Interestingly, we observed significant differences in the rate of adaptive ( $\omega_A$ ) or non-adaptive ( $\omega_{NA}$ ) substitutions between insular and mainland species as well as significant correlations between both variables and  $\pi_S$ , the predictor of long-term  $N_e$  (Fig. 4; see Sup Fig S7 for  $\alpha$  estimates). In detail,  $\omega_A$  is significantly positively correlated with the log-transformed  $\pi_S$  (slope=0.097,  $R^2=0.300$ ,  $p\text{-value}=0.004$ ; Fig. 4A) and negatively with the log-transformed  $\pi_N/\pi_S$



(slope=-0.403,  $R^2=0.389$ , p-value= $9 \times 10^{-4}$ ; Fig. S8). Reciprocally,  $\omega_{NA}$  is significantly negatively correlated with  $\pi_S$  (slope=-0.102,  $R^2=0.414$ , p-value= $6 \times 10^{-4}$ ; Fig. 4B) and positively with  $\pi_N/\pi_S$  (slope=0.385,  $R^2=0.437$ , p-value= $4 \times 10^{-4}$ ; Fig. S8). Mainland and island species exhibits significant differences for both  $\omega_A$  ( $\Delta\text{mean}=0.057$ ,  $t=-2.8407$ , p-value=0.010) and  $\omega_{NA}$  ( $\Delta\text{mean}=0.063$ ,  $t=4.16$ , p-value= $5 \times 10^{-4}$ ).



**Fig. 4: Proportion of adaptive (A) and non-adaptive (B) substitutions along the neutral genetic diversity gradient ( $\pi_S$ ) as estimated after comparing the observed and the expected  $d_N/d_S$  under near neutrality assuming the polymorphism data using the DFE-a method ( $\alpha$  shown Fig. S7; with  $\omega_A = \alpha(d_N/d_S)$  &  $\omega_{NA} = (1-\alpha)(d_N/d_S)$ ). Estimates were performed using all sites and the GammaExpo model. Error bars (purple line) represent the 95% confidence intervals of each estimate under this model. Where known, the IUCN conservation status of the investigated species is indicated (LC = least concerned, NT = near threatened, VU = vulnerable).**

## Discussion

The analysis or reanalysis of full-genome resequencing data allowed us to find consistent results of lower nucleotide diversities, higher mutation loads and lower adaptive substitution rates in endemic insular species as compared to mainland ones. These results provide important insights for both evolutionary and conservation biology, and we will develop these two lines of reasoning one after the other.

## 297 **Lower $N_e$ in endemic island species**

298 The smaller land area available on islands was supposed to constrain the upper bound of  
 299 both census and effective population sizes of endemic insular species, in such a way that this  
 300 demography has affected the ability of purifying selection to remove weakly deleterious  
 301 mutations. Overall, our results are consistent with this general hypothesis. For all parameters,  
 302 including  $\pi_S$  and  $\pi_N/\pi_S$ , our results were consistent with significant differences associated with  
 303 insularity (mainland/island), even after explicitly taking into account the phylogenetic signal. The  
 304 evidence of low  $N_e$  in insular species is important for evolutionary biology for a series of reasons,  
 305 (i) even if  $\pi_S$  and  $\pi_N/\pi_S$  are convenient and informative summary statistics in population  
 306 genomics, there is a fundamental statistical problem of non-independence of using  $\pi_S$  as both a  
 307 proxy of  $N_e$  and as the denominator of  $\pi_N/\pi_S$  for estimating the efficacy of selection (Piganeau  
 308 and Eyre-Walker 2009), (ii) leaving aside this statistical issue, there is a conceptual problem of  
 309 circularity in using population genomics variables to estimate  $N_e$  for testing the impact of this  $N_e$   
 310 on population genomics variables, (iii) very different scenarios can led to a similar level of  
 311 nucleotide diversity at synonymous sites, in particular assuming that selected and neutral alleles  
 312 have different dynamics (Mugal et al. 2013, 2020), (iv) finally, the use of life-history traits is  
 313 unsatisfactory as well because these proxies of  $N_e$  are indirect and only rely on the correlations  
 314 between body size and population density (White et al. 2007).

315 Our results supporting a lower  $N_e$  in the insular species is consistent with several previous  
 316 taxon-specific investigations (Loire et al. 2013 for Giant Galápagos tortoises; Rogers and Slatkin  
 317 2017 for woolly mammoths; Robinson et al. 2016; 2018 for island foxes). In contrast, James et al.  
 318 (2016) found that island and mainland species have a similar level of genetic diversity. James and  
 319 collaborators' study was however almost exclusively focused on organelle DNA sequences. The  
 320 difference can probably be explained by the importance of combining information from several  
 321 independent loci (*i.e.* thousands of orthologous genes in our study). More recently, Kutschera et  
 322 al. 2019 found consistent results in corvids. Using whole-genome sequencing data of seven  
 323 species within the genus *Corvus*, they found support for higher mutation loads in the insular  
 324 species, as compared to their mainland relatives. However, the low sample size (only two island  
 325 species), as well as the emphasis on a single taxon, limits its degree of generalization (Kutschera  
 326 et al. 2019). Using a larger set of island and mainland species from several distant taxa across the  
 327 passerida phylogeny, our conclusion reached another level of significance. Indeed, the difference

between mainland and island species remains highly significant, even after a conservative control for the phylogenetic relationship among the species, thus allowing us to fully exclude the hypothesis of an impact of the shared ancestry on the conclusions (Kutschera et al. 2019). In plants, a four-taxa comparison with both endemic island and widespread relatives for each taxa found significant lower nucleotide diversities and higher mutation loads on the island endemic species, and therefore provide support for the endemic island status in plants (Hamabata et al. 2019). Therefore, it is very likely that the endemic island status provides a good proxy of  $N_e$ , not only for specific animal clades such as birds or mammals, but for terrestrial species as a whole. More broadly, this result opens up new opportunities for using island species as models to understand the impact of  $N_e$  on genome evolution, including genome size, or natural selection on non-coding genomic regions.

339

## 340 **Bird population genomics data has ridden to the rescue of the nearly neutral** 341 **theory**

342 Fifty years after the introduction of the neutral theory of molecular evolution by Motoo  
343 Kimura (1968) and by Jack Lester King and Thomas Hughes Jukes (1969), the neutralist–  
344 selectionist debate has recently been reinvigorated (Kern & Hahn, 2018). In their standpoint, the  
345 latter authors reaffirmed the importance of the linked positive and negative selection in such a  
346 way that they reached the conclusion that “*the neutral theory has been overwhelmingly*  
347 *rejected*”. Kern & Hahn’s point of view generated a series of counter-claims supporting the  
348 central role of the neutral theory for explaining empirical patterns of molecular evolution,  
349 without denying the substantial contribution of linked selection (e.g. Chen et al. 2020; Jensen et  
350 al. 2019). Following the latter authors, our investigations are consistent with the high explanatory  
351 power of the neutral theory, or more precisely the nearly theory (Ohta, 1992). As expected, we  
352 indeed found that the proportion of deleterious mutations scaled negatively with the nucleotide  
353 diversity. Similarly, we found support for lower proportion of adaptive substitutions and higher  
354 proportion of non-adaptive substitutions, in species with lower effective population sizes.  
355 Interestingly, for all the species we investigated, the proportion of adaptive amino-acid  
356 replacement substitutions never reached 50% nor 25%, but only 14.3% for the species exhibiting  
357 the highest  $\omega_A$ , which is somewhat incongruent with a pervasive effect of natural selection in the

evolution of protein sequences, as discussed by Kern & Hahn based on the patterns observed in fruit flies (Kern & Hahn 2018; see also Jensen et al. 2019). Similar low values for  $\omega_A$  (<20%) were observed by Rousselle *et al.* 2020 based on the analysis of 50 species from ten distant groups of animals, including two bird clades: fowls and passerines. To sum up, using our panel of songbird species and based on the empirical patterns of molecular evolution we were able to recover, the nearly neutral theory holds true. It must however be noted that, through GC3 - a proxy of the local recombination rates in birds (e.g. Bolívar et al. 2016) - we also found unambiguous evidence for the contribution of linked selection, influencing both the local levels of nucleotide diversity and loads of deleterious mutation, and consistent with heterogeneous landscapes of  $N_e$  throughout genomes (Rousselle et al. 2019). Interestingly, our analysis reveals an intriguing interaction between recombination rate and the efficacy of selection. Indeed, the effect of GC3 on  $\pi_N/\pi_S$  significantly differs between species exhibiting a low  $N_e$  and species with a high  $N_e$ , with a stronger effect of recombination in insular (low  $N_e$ ) species. We have however shown that this interaction is independent of the process of gBGC (Fig. 3B & D). This effect could be the result of a different distribution of the fitness effects of mutations between low  $N_e$  and high  $N_e$  species. The effect of linked selection is known to be particularly strong for mildly selected mutations (Comeron et al. 2008). We can therefore hypothesize that low  $N_e$  species contribute to a DFE shifted toward more mildly selected mutations as compared to the DFE of high  $N_e$  species, and consequently produce more numerous footprints of natural selection in their genomes.

## **Ecological-evolutionary ties and perspectives**

The importance of population genomics for ecological research has long remained quite limited (Hugues et al. 2008), but there is now a growing interest for conservation biology studies on the interplay between ecology and evolution. At a macroevolutionary scale, it has become clear that the determinants of genetic diversity are mostly ecologically-driven, as shown by the strong correlations between some key life-history traits and both polymorphisms levels and mutation loads, namely propagule sizes and longevity, respectively. These ecological-evolutionary correlations have been independently recovered in animals and plants (Romiguier et al. 2014; Chen et al. 2017; Plomion et al. 2018). Interestingly, Romiguier et al. 2014 failed to find an effect of geographical range that might be due to the limitation of investigating these patterns in a limited number of species per taxon over a large diversity of taxa. Using a dataset of

639 mammalian species, James & Eyre-Walker (2020) however found support for significant correlations between the levels of mitochondrial diversity and the range size, in addition to some life-history traits such as the mass-specific metabolic rates and the age of sexual maturity, a trait which is itself strongly correlated with longevity. At a microevolutionary scale, the intensity of these ecological-evolutionary links remains more debated. Importantly, our analysis was designed to focus on a group of species exhibiting roughly similar body mass and longevity to allow us to investigate some other ecological factors, because the avian genetic diversity is known to be associated with body mass (Eo et al. 2011). It is important to mention that we also checked a posteriori the absence of correlation between body mass and the levels of genetic diversity (Spearman's  $Rho = -0.11$ ,  $p\text{-val} = 0.58$ ). Consistent with our earlier discussion about insularity, we found that the nucleotide diversity scaled positively with the species range, which therefore represents a gradual transition from species restricted to small islands to species widely distributed over continents. In addition, census population sizes were found to be another excellent explanatory ecological variable, explaining both the levels of nucleotide diversity and the mutation loads. However, neither a relationship between levels of nucleotide diversity and census size nor between the levels of diversity and range size were recovered when endemic island or mainland species were analyzed separately ( $p\text{-value} > 0.10$  but see Díez-del-Molino et al. 2018 and Brüniche-Olsen et al. 2019). At this smaller scale, recent demographic variations may have affected the expected relationship between  $N_e$  and census size and similarly,  $N_e$  and range size. It should be noted however that our dataset was not ideal for testing the variation among island or mainland species separately, especially given that no mainland species with small census population sizes or range sizes were included in this study.

How informative population genomics variables can be to evaluate the conservative status of the endangered species? Even if our dataset was not initially designed to tackle this issue, with only five threatened (vulnerable) species, all among the 9 Darwin's Finch species, we found no obvious contrast between these species and the other non-threatened endemic insular species, neither for the levels of nucleotide diversity nor for their efficacy of natural selection. This result came as a surprise given that a general outcome is that threatened animal species generally exhibit lower genetic diversity than non-threatened species, including for birds (at least at microsatellite loci, Doyle et al. 2015; Willoughby et al. 2016 for a review). It must however be noted that more recent investigation using whole-genome data of a single individual from 78 mammal species was unable to recover this difference, at least without explicitly taking into

account the animals' diets (Brüniche-Olsen et al. 2018). Similarly, Díez-del-Molino et al. (2018) were unable to recover a significant effect of the IUCN assessment on the levels of nucleotide diversity using a large set of both birds and mammals. Comparing some critically endangered plant species from 4 Eastern Asian taxa and some of their widespread relatives, Hamabata et al. (2019) however found lower genetic diversity and higher mutation loads in the critically endangered plant species, suggesting a long-term evolution in small population size and a long-term reduction in the efficacy of selection. Using divergence data from 76 species of teleostean fishes, Rolland & Romiguier (2019) similarly found that the IUCN red list index provides an equally significant correlation with the non-synonymous over synonymous substitutions than body length, suggesting that the proportion of deleterious substitutions is strongly associated with vulnerability in fishes. In birds, Brüniche-Olsen et al. (2019) recently investigated this patterns on 17 species of tanagers from 27 islands, including a large majority of Darwin's Finches from the Galapagos and the Coco Islands (8 of the 9 Darwin's finches are also used in our study), but also two species from the Barbados. The authors found that the threatened species of tanagers based on the IUCN status (considering all together 'critically endangered', 'endangered' or 'vulnerable') exhibited lower genetic diversity than non-endangered species (*i.e.* 'near-threatened' and 'least concerned') and that both body size and island area are good predictors for the levels of genomic diversity, but the authors did not explicitly evaluate the efficacy of purifying selection. As a consequence, we still have a long way to go towards precisely describing the strengths of these ecological-evolution ties and how this information is in line with the current species conservation status.

## **Unresolved issue to tackle: long-term survival of highly loaded species**

IUCN provides threat categorizations to guide short-term (*i.e.* a century or less) conservation efforts based on the ecological information available, e.g. the evolution of the census population sizes or the current geographical ranges. This is especially important given the large scientific consensus for a climate emergency and the need for an effective action for preserving biodiversity and ecosystems (Ripple et al. 2020). Beyond all these recent human-induced threats, a still open and non-trivial issue is to evaluate if population genomics can provide summary statistics informative enough to extend these objectives over a long-term timeframe. Small island populations indeed exhibit some intrinsic features that may contribute to an increased maladaptation and accelerate, and virtually lead to extinction, alone or in

association with the ecological reasons. We can in particular question how the mutation load evolves over time and if the deleterious burden can result in the extinction of the species. This hypothesis holds true only if these deleterious mutations are not opposed by compensatory or beneficial mutations (Lynch & Gabriel, 1990). This is especially important to mention given that the 4 threatened species (vulnerable status) are not these species exhibiting the lower proportion of adaptive substitutions (mean  $\omega_A=0.053$  as compared to 0.036 for the 15 species with a Least Concerned status, respectively). As a consequence, we can question if these high mutation loads can really contribute to extinction alone, following the general hypothesis that these mutations are accumulated faster in genomes than they are removed from the population by selection (Kondrashov 1988). Using a macro-evolutionary approach, Warren et al. (2018) found that the probability of being endangered for an island species increased with the time spent on an island (*i.e.* the time since the island was colonized). Similarly, Valente et al. (2020) found support for a negative relation between island sizes and extinction rates, confirming the general expectation of the island biogeography theory. These results are interpreted as the consequence of age-dependent processes such as ecological specialisation or coevolutionary balance between immigrant and resident species, but the accumulation of deleterious mutations might explain, in whole or in part, this phenomenon as well. Rogers & Slatkin (2017) proposed that, after a tipping point, this mutational meltdown may contribute to the ultimate steps in the road to extinction, as suspected based on one of the fossil remains of one the latest mammoth specimens (Palkopoulou et al. 2015). Such a study also illustrates the importance of paleogenomics in future conservation studies, in order to trace the temporal erosion of nucleotide diversity and accumulation of deleterious mutations, especially for the endangered species (Díez-del-Molino et al. 2018). With the currently available affordable sequencing strategies, the cost-benefit balance is increasingly favorable to get a reference genome assembly and reduced- or whole-genome resequencing data for ancient and/or modern samples of a given species (Brandies et al. 2019). As shown in this study, endemic island species represent species of high interest to evaluate, in a conservation biology context, the long-term consequences of such limited effective population sizes.



## 480 Materials and Methods

481

482 In this study, we both reanalyzed publicly available and generated our own sequencing  
 483 data. By generating the data, our objective was to target taxa containing both island and  
 484 mainland relatives (chaffinches and white-eyes) in order to increase our statistical power. More  
 485 broadly, our comparison is only based on species with relatively similar body-mass, longevity and  
 486 clutch-size. This control was introduced to reduce the risk of some confounding factors that could  
 487 correlate with  $N_e$  (Romiguier et al. 2014) in order to be able to truly assess the effect of  
 488 insularity. All scripts and programs used are available at the following Open Science Framework  
 489 repository: [https://osf.io/uw6mb/?view\\_only=a887417cbc91429dac0bbdc1705e2f2b](https://osf.io/uw6mb/?view_only=a887417cbc91429dac0bbdc1705e2f2b).

490

### 491 Re-use of publicly available data for passerine birds

492

493 We collected publicly available raw sequencing data on SRA from a large range of studies  
 494 (Table S2). Because the phylogenetic relationship between these species are not fully resolved  
 495 (Lamichhaney et al. 2015; 2016; Zink & Vázquez-Miranda 2019), we first evaluate the net  
 496 divergence between all pairs of species to delimitate 9 groups of species with a net divergence  
 497 ( $D_A$ ) > 0.1% (see Sup Note 1). Within each group, we selected a single population based on the  
 498 number of individuals sequenced publicly available (Lamichhaney et al. 2015).

499

### 500 Newly generated whole genome sequencing data (*Zosterops* and *Fringilla* 501 *species*)

502

503 All *Zosterops* and *Fringilla* individuals were captured using mist nets. With the notable  
 504 exception of African *Zosterops* species (*Z. pallidus* and *Z. virens*, see below), we collected blood  
 505 samples for each bird by venipuncture of the brachial vein and then stored in absolute ethanol at  
 506 -20°C until DNA extraction. For African species, *Z. pallidus* and *Z. virens* individuals, DNA was



507 extracted from liver, muscle or blood. For these samples, the vouchers are stored at the Museum  
508 National d'Histoire Naturelle (MNHN), Paris, France and a tissue duplicate is deposited in the  
509 National Museum Bloemfontein (South Africa). For all *Zosterops* and *Fringilla* samples, total  
510 genomic DNA were extracted using the DNeasy Blood and Tissue kit (QIAGEN, Valencia, CA)  
511 following the manufacturer instructions. Library preparation (1.0 µg DNA used per sample) and  
512 Illumina high-throughput sequencing using a paired-end 150 bp (PE150) strategy were performed  
513 at Novogene (Cambridge, UK) to a minimum sequencing yield of 18 Gb per sample. Details on  
514 samples are available in Table S2.

515

## 516 **Newly generated genome assembly (*Fringilla coelebs*)**

517 DNA from muscle tissues of one *Fringilla* individual captured in Madrid, Spain was  
518 sequenced to obtain a draft genome of this species. For this individual, two libraries were  
519 prepared, a Chicago library and a Dovetail HiC library (Dovetail Genomics, Scotts Valley, CA), as  
520 described previously in Putnam et al. (2016) and Lieberman-Aiden et al. (2009), respectively.  
521 Both libraries were sequenced on an Illumina HiSeqX platform to produce over 100 million 2x151  
522 bp reads. A *de novo* assembly was constructed using Meraculous (v. 2.2.2.5 diploid\_mode 1,  
523 Chapman et al. 2011) with a kmer size of 73 and using 397.3 million paired-end reads (totaling  
524 1,204 Gbp). Reads were trimmed for quality, sequencing adapters, and mate-pair adapters using  
525 Trimmomatic (Bolger, Lohse & Usadel, 2014). The input *de novo* assembly, shotgun reads,  
526 Chicago library reads, and Dovetail HiC library reads were used as input data for HiRise, a  
527 software pipeline designed specifically for using proximity ligation data to scaffold genome  
528 assemblies (Putnam et al. 2016). The HiRise assembly total length was 907.37 Mb obtaining 3,240  
529 scaffolds and a final N50 of 68.85 Mb. The read coverage was 274X.

530

## 531 **Species distribution and IUCN red list status**

532 Species range sizes were obtained using from BirdLife (<http://datazone.birdlife.org/>) or,  
533 when not available, estimated based on the information shown on the IUCN-red list webpage  
534 using CalcMaps (<https://www.calcmaps.com/map-area/>). For endemic island species with  
535 populations on different islands (*C. fusca*) or on both island and mainland (*F. coelebs*), we

considered the total island area as a maximum bound for the population range. The conservation status shown by the IUCN red list are at the species level and not at below-species level. As a consequence, we considered either a missing information for both populations (e.g. *Certhidea fusca* E and *C. fusca* L) or the status only for the most widely distributed species (e.g. the Least concerned (LC) status for the population with a large continental distribution rather than for the island one (e.g. *F. coelebs palmae*, the population only found in the small La Palma island). For *Ficedula speculigera*, a species with a  $D_A > 0.002$  (Sup. Note 1) and recognized as a distinct species from *F. hypoleuca*, no information is yet available in the IUCN red list database.

## **Gene models & orthology prediction**

We used one reference genome for all species belonging to the same clade (Table 1). For Darwin's finches, we use the genome of a medium ground-finch individual (*Geospiza fortis*; assembly GeoFor\_1.0; GCF\_000277835; Parker et al. 2012). For *Ficedula* flycatcher, we used the genome and gene models of the collared flycatcher (*Ficedula albicollis*, GCF\_000247815; assembly FicAlb\_1.4; Ellegren et al. 2012). For the Zosterops, we used the genome of the Reunion grey white-eye (*Zosterops borbonicus*; GCA\_007252995; assembly ZoBo\_15179\_v2.0; Leroy et al. 2019). For the Estrildidae, we used the assembly of the zebra finch (*Taeniopygia guttata*; GCF\_000151805; assembly taeGut3.2.4; Warren et al. 2010). For the chaffinches, we use the newly generated assembly of *Fringilla coelebs* (see above). Finally, we used the genome of the willow warbler (*Phylloscopus trochilus*; GCA\_002305835; assembly ASM230583v1; Lundberg et al. 2017) and of the great tit (*Parus major*; GCF\_001522545.2; assembly Parus\_major1.1; Laine et al. 2016). For the willow warbler (*Phylloscopus trochilus*) and the chaffinch (*Fringilla coelebs*), we performed a protein homology detection and intron resolution using genBlastG (She et al. 2011; <http://genome.sfu.ca/genblast/download.html>) with the following options “-p genblastg -c 0.8 -r 3.0 -gff -e 1e-10”. The proteins of the collared flycatcher (*Ficedula albicollis*, assembly FicAlb\_1.4; Ellegren et al. 2012) were used as reference.

To analyse the same orthologous sequences in all species, we used the set of 8253 orthologs identified in Jarvis et al. 2014. Then, we add the sequence of our species to this set of

orthogroups using the method described in Scornavacca et al. 2019. Briefly, each orthogroup was used to build an HMM profile using the HMMER toolkit (Eddy 2011). Then, for each new sequence, *hmmsearch* was used on the HMM database to get the best hits among the orthogroups. For each orthogroup, the most similar sequences for each species were then detected via *hmmsearch*. Outputs from *hmmsearch* and *hmmsearch* were considered as accurate if the first hit score was substantially better than the second best one (in order to limit the risk of paralogy), following a best-reciprocal-hit approach when the results of both programs were compared (Scornavacca et al. 2019).

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## 574 **Variant identification**

We used Trimmomatic (v.0.33; Bolger et al. 2014) to remove adapters, stringently trim and filter reads using the following set of parameters: LEADING:3 TRAILING:3 SLIDINGWINDOW:4:15 MINLEN:50. All trimmed reads were then mapped against the reference genome for each clade (see above) with BWA mem (v. 0.7.12; Li 2013) using default settings. Unmapped reads and mapped reads with a quality (MQ) below 20 were then discarded. Potential PCR duplicates were then removed using MarkDuplicates v. 1.140 (Picard tools). Variant calling was then performed using GATK (v. 3.7; McKenna et al. 2010). First, we used HaplotypeCaller on single samples (gVCF) to call SNPs using default parameters. For each species, we then performed a joint genotyping ("GenotypeGVCFs"). To ensure high quality in our dataset, we filtered out low-quality SNPs using several settings: a quality by depth (QD) < 2.0, a Fisher Strand (FS) bias >60, a mapping quality (MQ) <40, a MQranksum < -2 or a ReadPosRankSum < -2 or a Raw Mapping Quality (Raw\_MQ) < 45,000. SNPs satisfying one of the conditions described above were discarded. For every group of species, we performed principal components analyses (PCA) based on a random sampling of SNPs over the genome (50-200k) to capture additional levels of population structure or an unfortunate misnaming of an individual that could have occurred at one point between the bird sampling campaign and the analysis of the raw sequencing data.

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## 592 **Empirical estimate of the SNP calling error rate**

One individual (namely *Zosterops virens* RCHB1917) was sequenced twice at coverage 10x and both replicates went through all the genotyping pipeline blindly. We estimate the error rate by counting the number of differences between the two replicates. All the differences detected correspond to heterozygous sites. We estimated a rate of  $2.9 \times 10^{-5}$ , which is approximately fifty times lower than the expected genome-wide heterozygosity for this focal species ( $1.5 \times 10^{-3}$ ).

## **Sequence reconstruction**

For each individual, we then reconstructed fasta sequences from VCF files. First, we discarded all indel variations to ensure an exact match between positions in sequences and coordinates in the reference gff file (*i.e.* no frameshift mutation allowed). For each scaffold, we performed a base-by-base reconstruction of two sequences per individual by adding either a reference or an alternate allele based on the genotype information available in the vcf file at the position under investigation. Albeit simplistic, this approach is valid here because our study is only based on the nucleotide levels and does not require haplotype phase information. To exclude sites with abnormally low or high coverage, we compute percentiles of the distribution of coverage over the whole genome for each individual and then calibrate a minimum and maximum coverage. All sites exhibiting very low or very high coverage were masked. A minimum threshold value of 3 was set for all species. We then used a home-made python script to extract CDS from these reconstructed sequences based on the gff files.

## **Mitochondrial genomes assembly and phylogeny**

For the species with mtDNA genomes available on genbank, we used the data (accessions: *Ficedula albicollis* : KF293721; *Fringilla coelebs* : NC\_025599; *Fringilla teydea* : KU705740; *Parus major* : NC\_026293; *Taeniopygia guttata* : NC\_007897; *Zosterops borbonicus* : MK529728; *Zosterops pallidus* : MK524996. We also add three outgroups : *Corvus brachyrhynchos* : NC\_026461; *Lanius cristatus*: NC\_028333; *Menura novaehollandiae* : NC\_007883).

For the remaining species, we used MitoFinder (vers. 1.1; <https://github.com/RemiAllio/MitoFinder>; Allio et al. 2020) to extract all the mitochondrial

protein coding genes. One individual per species was randomly selected to use in MitoFinder with default parameters. Alignments were performed gene by gene using macse (vers. 2; Ranwez et al. 2018). Alignments are available at the following URL: [https://osf.io/uw6mb/?view\\_only=a887417cbc91429dac0bbdc1705e2f2b](https://osf.io/uw6mb/?view_only=a887417cbc91429dac0bbdc1705e2f2b). Then all genes were concatenated to form a supermatrix (missing data = 20% on average, sd = 14%). Next, a phylogenetic analysis was performed using IQTREE (Nguyen et al. 2015) using the “GTR+G4” substitution model and an ultrafast bootstrap option to have a crude approximation of each node support.

## Summary statistics of the polymorphic data

$\pi_N/\pi_S$  ratios were computed using *seq\_stat\_coding*, a publicly available Bio++ script already used by Leroy et al. 2019 (<https://osf.io/uw6mb/>). In addition, we used the  $\pi_N/\pi_S$  estimates based on the site frequency spectra at both non-synonymous and synonymous sites and described in Rousselle and collaborators (2020) to check the accuracy of these estimates (method 2 in Figs. S4 & S5, see also Table S1). Guanine-Cytosine (GC) content at third-codon positions of protein-coding genes (hereafter GC3), an excellent proxy of the local recombination rate in birds (Bolívar et al. 2016) was also computed under *seq\_stat\_coding*. To estimate the within-genome variation in the efficacy of selection, we estimated  $\pi_N/\pi_S$  on sets of genes representing a total concatenated coding alignment of 2 Mb. To do so, genes were sorted in ascending values of GC3 in order to compute  $\pi_N/\pi_S$  ratios on genes exhibiting roughly the same GC3 values (median GC3 value for the bin was then considered, e.g. in Fig. 3). To ensure reliable estimates, we excluded the last  $\pi_N/\pi_S$  estimate corresponding to genes exhibiting the highest GC3 values when the number of genes was likely insufficient to get a reliable estimate (<1 Mb of coding sequence used for the computations).

## Summary statistics of the divergence data

We used the method implemented by Galtier (2016) (Grapes. v1.0) to estimate  $\alpha$ ,  $\omega_A$  and  $\omega_{NA}$  using the approach introduced by Eyre-Walker & Keightley (2009). Briefly, we fitted both a negative Gamma distribution and an exponential distribution to the synonymous and non-

synonymous SFS (the so-called GammaExpo model of Galtier 2016) to model the DFE. Fitted parameters of the DFE were then used to compute the expected  $D_N/D_S$  under near neutrality, which was compared to the observed  $d_N/d_S$  to estimate the adaptive substitution rate ( $\omega_A$ ) and the proportion of adaptive substitutions ( $\alpha$ ) [with  $\omega_A = \alpha(d_N/d_S)$  &  $\omega_{NA} = (1-\alpha)(d_N/d_S)$ ]. Potential recent changes in population size that affect the SFS were taken into account via the use of nuisance parameters capturing distortions of the SFS optimized alongside the DFE parameters (Eyre-Walker et al. 2006).

## **Model comparisons**

All statistical analyses were performed using R (R core team 2018). We only considered models with a similar number of observations and compared these models based on the Akaike information criterion (AIC). To test the influence of the explanatory variables on  $\pi_S$  and  $\pi_N/\pi_S$ , we either used simple linear models or Phylogenetic Generalized Least Square (PGLS) models. For the latter, we used the model implemented in the nlme package (Pinheiro et al. 2020). The mitochondrial phylogeny was considered as the species tree taken into account assuming a Brownian correlation structure (using “corBrownian” from the “ape” package; Paradis and Schliep 2019). P-value and AIC were computed using the anova.gls function. The rationale of the phylogenetic control is to account for the shared polymorphisms (part of species similarity that is explained by the inheritance from a common ancestor). The level of polymorphism of a given species is dynamically controlled by drift, mutation rate and natural selection. As soon as two species do not share a significant fraction of their polymorphism, there is no need to account for their phylogenetic proximity because their polymorphisms evolved independently.

R Plots were performed using a series of R packages: cowplot (Wilke 2016), ggplot2 (Wickham 2016), ggpubr (Kassambara 2018), ggrepel (Slowikowski 2019) and ggtree (Yu et al. 2017).

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