Report

Island songbirds as windows into evolution in small populations

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SUMMARY

Due to their limited ranges and inherent isolation, island species have long been recognized as crucial systems for tackling a range of evolutionary questions, including in the early study of speciation.^{1,2} Such species have been less studied in the understanding of the evolutionary forces driving DNA sequence evolution. Island species usually have lower census population sizes (N) than continental species and, supposedly, lower effective population sizes (N_e). Given that both the rates of change caused by genetic drift and by selection are dependent upon N_e , island species are theoretically expected to exhibit (1) lower genetic diversity, (2) less effective natural selection against slightly deleterious mutations,^{3,4} and (3) a lower rate of adaptive evolution.^{5–8} Here, we have used a large set of newly sequenced and published whole-genome sequences of Passerida species (14 insular and 11 continental) to test these predictions. We confirm that island species exhibit lower census size and N_e , supporting the hypothesis that the smaller area available on islands constrains the upper bound of N_e . In the insular species, we find lower nucleotide diversity in coding regions, higher ratios of non-synonymous to synonymous polymorphisms, and lower adaptive substitution rates. Our results provide robust evidence that the lower N_e experienced by island species has affected both the ability of natural selection to efficiently remove weakly deleterious mutations and also the adaptive potential of island species, therefore providing considerable empirical support for the nearly neutral theory. We discuss the implications for both evolutionary and conservation biology.

RESULTS

To assemble our dataset, we used population-level sequencing data (Table 1) from 25 passerine bird species or subspecies, consisting of 14 insular and 11 continental, with a total of 295 individual whole-genome sequences (89 newly sequenced). All species belong to the Passerida lineage, a species-rich clade of songbirds with fairly similar life-history traits. Our dataset includes at least 4 independent continental-island transitions that occurred across the songbird phylogeny (Figure S1) enabling us to efficiently account for phylogenetic structure in all statistical tests reported below (phylogenetic generalized least square [PGLS]; see also Table S1 for additional tests).

Do island species exhibit genomic signatures consistent with low N_e ?

Past effective population sizes were inferred using the pairwise sequentially Markovian coalescent (PSMC) approach for one randomly selected individual from each species (Figure S2) and were then averaged over the last one million years. The analyses confirmed that island species exhibit a significantly lower mean N_e than continental species over the last one million years (mean N_e = 362,456 and 94,944 for continental and island species, respectively, Figure 1A; log-transformed N_e , PGLS p = 1.0×10^{-4}). Specifically, inferred mean N_e values over the last million years range from 6.1×10^4 for the Tenerife blue chaffinch (*Fringilla teydea*) to 1.2×10^6 for a continental population of

Table 1. Sequencing data used in this study

	Species	Clade	Range	Individuals	Data
1	<i>Certhidea olivacea</i> (S)	Darwin's finches	island	5	9
2	<i>Certhidea fusca</i> (E)	Darwin's finches	island	10	9
3	<i>Certhidea fusca</i> (L)	Darwin's finches	island	10	9
4	<i>Platyspiza crassirostris</i> (Z)	Darwin's finches	island	5	9
5	<i>Camarhynchus pallidus</i> (Z)	Darwin's finches	island	5	9
6	<i>Pinaroloxias inornata</i> (C)	Darwin's finches	island	8	9
7	<i>Geospiza difficilis</i> (P)	Darwin's finches	island	10	9
8	<i>Geospiza septentrionalis</i> (W)	Darwin's finches	island	8	9
9	<i>Geospiza conirostris</i> (E)	Darwin's finches	island	10	9
10	<i>Ficedula albicollis</i>	Ficedula flycatchers	continental	20	10
11	<i>Ficedula hypoleuca</i>	Ficedula flycatchers	continental	20	10
12	<i>Ficedula speculigera</i>	Ficedula flycatchers	continental	20	10
13	<i>Ficedula semitorquata</i>	Ficedula flycatchers	continental	20	10
14	<i>Zosterops borbonicus</i>	white-eyes	island	6	11
				1	12
				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	continental	2	this study
18	<i>Zosterops virens</i>	white-eyes	continental	11	this study
19	<i>Fringilla coelebs</i>	chaffinches	continental	9	this study
20	<i>Fringilla coelebs palmarum</i>	chaffinches	island	15	this study
21	<i>Fringilla teydea</i>	chaffinches	island	10	this study
22	<i>Taeniopygia guttata castanotis</i>	Estrildidae	continental	19	13
23	<i>Poephila acuticauda acuticauda</i>	Estrildidae	continental	10	13
24	<i>Parus major</i>	Paridae	continental	10	14
25	<i>Phylloscopus trochilus</i>	Phylloscopidae	continental	9	15

The abbreviation in parenthesis following Darwin's finches names indicates the island of origin (C, Coco; E, Española; L, San Cristobal; P, Pinta; S, Santiago; W, Wolf; Z, Santa Cruz). See also [Methods S1](#).

the common chaffinch (*F. coelebs*), representing an ~20 fold difference ([Figure 1A](#); [Table S2](#)).

Such long-term differences in N_e between insular and continental species are expected to generate differences in nucleotide diversity levels, because genetic variation is determined by both mutation rate and effective population size. By estimating nucleotide diversity at synonymous (π_S) and at non-synonymous sites (π_N), we find marked differences between island and continental species. Using 6,499 orthologous genes on average (range: 5,018–7,514, among 8,253 orthogroups¹⁶), we find that π_S varies from 0.07% in the Tenerife blue chaffinch to 1.25% in the willow warbler (*Phylloscopus trochilus*), representing a 17-fold difference between these island and continental species ([Table S2](#); [Figure S3](#)). Island species exhibit significantly lower mean π_S than continental species (mean $\pi_S = 0.59\%$ and 0.18% for continental and island species, respectively, [Figure 1B](#); PGLS, $p = 5.29 \times 10^{-3}$).

In addition to strong evidence for lower N_e in island species, we also find lower census population sizes in the island species (island: 7 species, median: 1.1×10^4 [range: $6.2 \times 10^2 - 3.0 \times$

10^5]; continental: 6 species, median: 2.5×10^8 [range: $2.0 \times 10^5 - 5.7 \times 10^8$]; log-transformed census sizes, PGLS, $p = 8.29 \times 10^{-5}$). Furthermore, both log₁₀-transformed current census population sizes and geographical range in square kilometers are positively correlated with π_S ([Figures 2A](#) and [2C](#); PGLS, $p < 0.01$; [Table S1](#)). Taken all together, these results provide strong support for the view that long-term restrictions on census population sizes due to the limited surface area available to island species constrains the upper bound of effective population size.

Are deleterious mutations segregating more in island species?

Based on the nearly neutral theory of molecular evolution, the higher level of genetic drift associated with lower N_e is expected to contribute to an accumulation of slightly deleterious mutations in island species relative to their continental counterparts. Using the ratio of non-synonymous to synonymous mutations (π_N/π_S) as a proxy for the proportion of these slightly deleterious mutations, we recover, on average, a 40% higher π_N/π_S in island

Figure 1. Island species as models for evolution in small effective population sizes

(A) Local polynomial regression (LOESS with span = 1.25) between the ratio of nonsynonymous to synonymous nucleotide diversity (π_N/π_S) and the mean effective population sizes over the last million years (N_e), as inferred using PSMC (see Figure S3 for a log-log regression between π_N/π_S and π_S estimates, respectively).

(B and C) Variation in nucleotide diversity (π_S , B) and π_N/π_S between island endemic and continental species (C). Red asterisks indicate significance of the PGLS test (** $p < 0.01$).

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species than in continental species (Figure 1C; mean $\pi_N/\pi_S = 0.145$ and 0.201 for continental and island species, respectively, PGLS $p = 4.57 \times 10^{-3}$).

In addition, we find substantial within-genome variation in the accumulation of slightly deleterious mutations, as well as in the levels of nucleotide diversity, in such a way that π_S and π_N/π_S are, respectively, positively and negatively correlated to the GC content at the third codon position (GC3). GC3 provides a robust proxy of recombination rate in birds^{17,18} (Methods S1). By comparing sets of genes exhibiting the lowest and highest GC3, we found a more marked π_N/π_S differences in genes exhibiting low GC3 ($\Delta\text{mean}_{\text{continental versus Island}} = 0.107$, 95% CI: $0.077\text{--}0.194$) than in those exhibiting high GC3 ($\Delta\text{mean} = 0.029$, 95% CI: $0.011\text{--}0.048$). 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 value model $< 2.2 \times 10^{-16}$, including $p < 2.2 \times 10^{-16}$, 2.8×10^{-10} , and 1.33×10^{-05} for GC3, insularity, and the interaction, respectively). These correlations are found to be stronger in island species relative to their continental counterparts, with a particularly pronounced difference in π_N/π_S in genes exhibiting a low GC3. These results are robust to a control for GC-biased gene conversion

(Methods S1). Recombination limits genetic interactions between selected mutations and can therefore improve the efficiency of selection.^{19,20} These results suggest that the intensity of the differences between island and continental species in the effectiveness of purifying selection relies heavily on the local genomic context.

We found strong negative correlations between (1) π_N/π_S and the $\log_{10}$ -transformed N_e averaged over the last one million years (Figure 1A; PGLS, $p = 1.0 \times 10^{-4}$) and (2) the non-transformed π_N/π_S and π_S values (PGLS, $p = 4.85 \times 10^{-5}$). $\log_{10}$ -transformed current census population sizes, as well as geographical range sizes, significantly correlate with π_N/π_S (Figure 2). In contrast, the IUCN red list assessments have no effect on π_N/π_S or π_S (Table S1; PGLS $p \gg 0.05$). Taken together, our results provide strong empirical evidence that differences in census population sizes between island and continental species translate into differences in N_e , and that these differences have a marked influence on genetic diversity and the efficiency of natural selection. These findings fit remarkably well with the expectation from the nearly neutral theory.

Do insular species show lower adaptive potential?

Theory predicts that lower N_e in island species should lead to a lower rate of adaptive substitutions than in continental species, if

Figure 2. Ecological-evolutionary correlations based on the variables investigated in this study

π_S and π_N/π_S are used as proxies of N_e and the efficiency of natural selection to remove deleterious variants and are correlated with both the median estimates of the current census population sizes (A and B) and the geographical range sizes (C and D). Both ecological and evolutionary parameters are log-transformed. Filled and open dots represent the island and the continental species, respectively (Figure 1 for details). Only the 13 species with estimates of the current census population sizes are included for the (A) and (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). See also Figure S3 and Tables S1 and S2.

adaptation is limited by the supply of new mutations⁸ and/or if slightly advantageous mutations become effectively neutral in low N_e species.⁶ For taxa with at least two species (i.e., all except *Parus* and *Phylloscopus*), we used the maximum likelihood method implemented in Grapes²¹ to estimate non-adaptive rate of substitution (ω_{NA}) and adaptive rate of substitution (ω_A) with ω (i.e., d_N/d_S) being the sum of ω_{NA} + ω_A . No significant difference in ω was observed between island and continental species ($\omega_{\text{island}} = 0.194$ and $\omega_{\text{continental}} = 0.187$). By contrast, island species showed a higher ω_{NA} ($\Delta\text{mean}_{\text{continental versus Island}} = 0.063$) and a lower ω_A ($\Delta\text{mean}_{\text{continental versus Island}} = 0.056$) (Figure 3; see Figure S4 for α estimates) than continental counterparts. However, these differences are only significant for tests that did not explicitly take phylogenetic structure into account (PGLS, $p = 0.257$ and $p = 0.237$; non-PGLS, $p = 0.014$ and $p = 0.002$ for ω_A and ω_{NA} , respectively; Table S1), and therefore they should be interpreted with caution.

We found that ω_A was positively correlated with log₁₀-transformed π_S (PGLS, $p = 0.029$; Figure 3A) and negatively correlated with the log₁₀-transformed π_N/π_S (PGLS, $p = 0.034$; Table S1). Reciprocally, ω_{NA} is significantly negatively correlated with

log₁₀-transformed π_S (PGLS, $p = 0.020$; Figure 3B) and positively with log₁₀-transformed π_N/π_S (PGLS, $p = 0.025$; Figure S4).

Overall, our analysis suggests that a lower N_e doubly affects island species relative to continental species, because (1) relatively fewer adaptive mutations can reach fixation, and (2) the lower efficiency of natural selection allows a greater proportion of weakly deleterious variants to reach fixation in insular species.

DISCUSSION

Our analysis of whole-genome resequencing data has allowed us to find lower nucleotide diversity, a higher frequency of slightly deleterious mutations, and lower adaptive substitution rates in the island species than in the continental ones. These results provide important insights for evolutionary biology, and they also have major implications for the conservation of species with small populations.

Island species as models for studying the evolutionary consequences of small N_e

The smaller land area available on oceanic islands should constrain the upper bound of both census and effective

Figure 3. Relationships between adaptive and non-adaptive substitution rates and synonymous genetic diversity

Proportion of adaptive (A) and non-adaptive (B) substitutions along the neutral genetic diversity gradient (π_S) as estimated by comparing the observed and the expected d_N/d_S under near neutrality assuming the polymorphism data using the DFE- α method (α shown Figure S4; 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). See also Table S1.

population sizes of insular species, to such an extent that demography affects the ability of purifying selection to remove weakly deleterious mutations. Our results are largely consistent with this general hypothesis and suggest that contemporary census sizes provide information on long-term N_e (but see also Díez-Del-Molino et al.²² and Peart et al.²³). For most population genomic estimates we investigated, including π_S , π_N/π_S and PSMC-inferred N_e , we observed significant differences between continental and island species that are consistent with theoretical expectations.

Previous taxon-specific studies have reported low N_e in a diverse range of island organisms (e.g., giant Galápagos tortoises,²⁴ woolly mammoths,²⁵ island foxes,^{26,27} and *Corvus*²⁸). Therefore, it is very likely that island species predominantly exhibit lower N_e than their more abundant, broadly distributed, mainland relatives, and this pattern may not be restricted to some specific animal clades such as birds or mammals but may also be true for a large range of taxa (e.g., Hamabata et al.²⁹ for plants). More broadly, this result opens up new opportunities for using island species as models to understand the impact of N_e on genome evolution in natural populations, including genome size, or of natural selection on non-coding genomic regions.

Broad support for the nearly neutral theory of molecular evolution

Fifty years after the introduction of the neutral theory of molecular evolution by Kimura³⁰ and King and Jukes,³¹ and, after being extended into the nearly neutral theory,⁴ the neutralist-selectionist controversy remains one of the sharpest and most polarized debates in biology. Based on our large genome-scale empirical data, our results match theoretical expectations of the nearly neutral theory remarkably well. This is consistent with the strength of this theory in explaining patterns of DNA sequence evolution, allowing us to affirm that the nearly neutral theory is overwhelmingly supported by our dataset. Slightly

deleterious mutations are frequent and become effectively neutral when the effect of genetic drift increases, as is typically observed in insular species.

Selective processes, including positive selection on beneficial alleles and background selection, play an important role in the sequence evolution of the investigated species but cannot be used to reject the theory as a whole. Empirical investigations found that the proportion of adaptive substitutions does not overall scale with N_e when distant taxa are considered all together (e.g., Galtier²¹), but taxa-specific investigations were able to find such a relationship, with a lower proportion of adaptive substitutions in species with a lower N_e , as recently reported for several groups of animals.⁸ First, our analyses provide additional evidence for such a relationship in passerine birds. Second, we indeed observe that local recombination rates influence both local levels of nucleotide diversity and the number of deleterious mutations, which is consistent with heterogeneous landscapes of N_e throughout genomes.¹⁸ However, significant differences between island and continental species were similarly recovered in both lowly and highly recombining regions of the genome, supporting the claim that background selection does not fundamentally change the predictions that can be drawn from the theory.

Ecological-evolutionary ties and perspectives

At the macroevolutionary scale, strong correlations between life-history traits and both levels of polymorphism and ratios of non-synonymous to synonymous mutations have been reported in the literature for both animals and plants,^{32–34} suggesting that determinants of genetic diversity are mostly ecologically driven. We found that nucleotide diversity scales positively with species range, which therefore suggests a gradual transition between species restricted to small islands and species widely distributed over continents. Recently, Peart et al.²³ proposed that conservation priorities should be defined based on the ratio of census size

to N_e . However, whether population genomic estimates of N_e are informative enough to assess conservation status is questionable. A general outcome is that animal species classified as threatened generally exhibit lower genetic diversity than those classified as non-threatened, including birds (at least at microsatellite loci^{35,36}). Based on our whole-genome analyses, we can report no obvious contrast between the four island species classified as threatened (vulnerable status) and the species classified as non-threatened, neither for the levels of nucleotide diversity nor for their efficiency of natural selection (but see Brüniche-Olsen et al.³⁷). Díez-del-Molino et al.²² were also unable to recover a significant effect of the IUCN assessment on the levels of nucleotide diversity in birds and mammals. Using 78 mammal species, Brüniche-Olsen et al.³⁸ only recovered this pattern when the animals' diets were explicitly taken into account. Consequently, it seems that we still have a long way to go toward precisely describing whether these genomic features are completely independent or are correlated to some extent with the current conservation status.

Another open question is whether population genomics can provide information so that short-term IUCN objectives can be extended over a longer time frame? Even if some island species accumulate slightly deleterious mutations,³⁹ supposedly leading to increased maladaptation, we can question whether this burden of slightly deleterious mutations can lead to species extinction. This hypothesis holds true only if these deleterious mutations are neither purged nor opposed by compensatory or beneficial mutations.⁴⁰ Remarkably, the four species classified as threatened are not those exhibiting the lowest proportion of adaptive substitutions (mean $\omega_A = 0.053$ compared to 0.036 for the 15 species with a least concern status). Recent macro-evolutionary investigations, however, provide support for this increased risk of (1) being endangered depending on the time since the species colonized the island⁴¹ or (2) becoming extinct depending on the island size.⁴² Age-dependent processes such as ecological specialization were proposed, but the accumulation of deleterious mutations might explain this phenomenon as well. Rogers and Slatkin²⁵ propose that, after a tipping point, this mutational meltdown might contribute to the ultimate steps in the road to extinction. Endemic island species therefore represent taxa of high interest in the evaluation of the long-term consequences of evolution under low effective population sizes.

STAR★METHODS

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SUPPLEMENTAL INFORMATION

Supplemental Information can be found online at <https://doi.org/10.1016/j.cub.2020.12.040>.

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AUTHOR CONTRIBUTIONS

T.L. designed the research, conducted the analyses, and wrote the paper. B.N. designed the research, contributed to the analyses, supervised the work, and wrote the paper. M.R. contributed to the analysis of the divergence data. M.-K.T. performed the wet lab work. A.E.C. performed preliminary investigations using the publicly available data and performed the analysis to identify the Darwin's finch populations and species to consider for the paper. C.S. conducted the analyses of orthologous assignment. M.R., G.B., and B.M. provided the sequence data of *Fringilla coelebs* species and the reference *Fringilla* genome. J.C.I. provided the *Fringilla* samples used in this study. D.H.D.S. and J.F. helped in field work to collect mainland and island *Zosterops* samples. C.T. and B.M. helped during field work in Reunion and revised the manuscript. All the authors read and approved the manuscript.

DECLARATION OF INTERESTS

The authors declare no competing interests.

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STAR★METHODS

KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Biological samples		
18 <i>Zosterops borbonicus</i> samples	Field sampling in Reunion , France	See Table S3
15 <i>Zosterops olivaceus</i> samples	Field sampling in Reunion , France	See Table S3
9 <i>Zosterops mauritinaus</i> samples	Field sampling in Mauritius	See Table S3
1 <i>Zosterops pallidus</i> sample	Field sampling in South Africa	See Table S3
11 <i>Zosterops virens</i> samples	Field sampling in South Africa	See Table S3
9 continental <i>Fringilla coelebs</i> samples	Field sampling in continental Spain	See Table S3
15 <i>Fringilla coelebs palmae</i> samples	Field sampling in La Palma, Spain	See Table S3
10 <i>Fringilla teydae</i> samples	Field sampling in Tenerife, Spain	See Table S3
Deposited data		
Raw reads <i>Taeniopygia</i> & <i>Poephila</i>	13	BioProject PRJEB10586
Raw reads <i>Parus</i>	14	BioProject PRJNA381923
Raw reads <i>Phylloscopus</i>	15	BioProject PRJNA319295
Raw reads Darwin's finches	9	BioProject PRJNA263122
Raw reads <i>Ficedula</i>	10	BioProject PRJEB7359
Raw reads <i>Zosterops</i>	11,12 and this study	BioProjects PRJEB18566, PRJNA530916, PRJNA661201
Raw reads <i>Fringilla</i>	This study	BioProject PRJNA661201
Software and algorithms		
Meraculous v. 2.2.2.5	43	https://jgi.doe.gov/data-and-tools/meraculous/
HiRise	44	https://github.com/DovetailGenomics/HiRise_July2015_GR
Trimmomatic	45	http://www.usadellab.org/cms/?page=trimmomatic
BWA mem	46	http://bio-bwa.sourceforge.net/
Picard v. 1.140	47	http://broadinstitute.github.io/picard/
GATK v. 3.7	48	https://gatk.broadinstitute.org/hc/en-us
MitoFinder v.1.1	49	https://github.com/RemiAllio/MitoFinder
Macse v.2	50	https://bioweb.supagro.inra.fr/macse/index.php?menu=releases
IQTREE	51	http://www.iqtree.org/
genBlastG	52	http://genome.sfu.ca/genblast/download.html
HMMER toolkit	53	http://hmmer.org/
PSMC	54	https://github.com/lh3/psmc
Grapes. v1.0	21	https://github.com/BioPP/grapes
R v3.6.3	55	https://cran.r-project.org/
Scripts used	This study	https://osf.io/uw6mb/

RESOURCE AVAILABILITY

Lead contact

Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, Benoit Nabholz (benoit.nabholz@umontpellier.fr).

Materials availability

This study did not generate new unique reagents.

Data and code availability

All raw sequencing data have been deposited in the Sequence Read Archive (SRA) under BioProject:PRJNA661201. All scripts and programs used are available at the following Open Science Framework repository: <https://osf.io/uw6mb/>

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Species included in the study

In this study, we both reanalyzed publicly available data and generated our own sequencing data from 25 passerine species (Table S3). By generating new sequencing data, our objective was to target taxa containing both island and continental relatives (chaffinches and white-eyes) in order to increase our statistical power. More broadly, our comparison is only based on species with relatively similar body-mass, longevity and clutch-size. This control was introduced to reduce the risk of some confounding factors that could correlate with N_e ³⁴ in order to be able to truly assess the effect of insularity.

Species distribution and IUCN red list status

Species range sizes were obtained from BirdLife (<http://datazone.birdlife.org/home>) or estimated based on the information shown on the IUCN-red list webpage using CalcMaps (<https://www.calcmaps.com/map-area/>). For endemic island species, we considered the total island area as a maximum bound for the population range. The IUCN red list conservation status is given at the species level and not at below-species level. As a consequence, we either considered this information to be missing for both populations (e.g., *Certhidea fusca* E and *C. fusca* L) or we only used the status for the most widely distributed species (e.g., the Least Concern (LC) status for the population with a large continental distribution rather than for the island one as in *F. coelebs palmae*). For *Ficedula speculigera*, a species with a $D_A > 0.002$ (Methods S1) and recognized as a distinct species from *F. hypoleuca*, no information is yet available in the IUCN red list database.

METHOD DETAILS

DNA extraction and sequencing (*Zosterops* and *Fringilla* species)

All *Zosterops* and *Fringilla* individuals were captured using mist nets. With the exception of African *Zosterops* species (*Z. pallidus* and *Z. virens*, see below), we collected blood samples for each bird by venipuncture of the brachial vein and stored blood in absolute ethanol at -20°C until DNA extraction. For African species, *Z. pallidus* and *Z. virens* individuals, DNA was extracted from liver, muscle or blood. For these samples, voucher specimens are stored at the Museum National d'Histoire Naturelle (MNHN), Paris, France and a tissue duplicate is deposited in the National Museum Bloemfontein (South Africa). For all *Zosterops* and *Fringilla* samples, total genomic DNA was extracted using the DNeasy Blood and Tissue kit (QIAGEN, Valencia, CA) following the manufacturer's instructions. Library preparation (1.0 μg DNA used per sample) and Illumina high-throughput sequencing using a paired-end 150 bp (PE150) strategy were performed at Novogene (Cambridge, UK) to a minimum sequencing yield of 18 Gb per sample (i.e., $\sim 15\times$ coverage). Details on samples are available in Table S3. For these species, we used exactly the same approach as for the publicly available data for variant identification and sequence reconstruction strategies, as described in Methods S1. All newly sequenced raw reads are available under the SRA BioProject : PRJNA661201.

Publicly available sequencing data

We collected publicly available raw sequencing data on SRA from a large range of studies (Table S3). The phylogenetic relationships among Darwin's finches are not fully resolved,^{9,56,57} so we first evaluate the net divergence between all pairs of species to delimit 9 groups of species with a net divergence ($D_A > 0.1\%$; Methods S1). Within each group, we selected a single population based on the number of sequenced individuals that were publicly available.⁹ Variant identification and sequence reconstruction steps are described in Methods S1.

Variant identification

We used Trimmomatic (v.0.33⁴⁵) 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. 0.7.12⁴⁶) using default settings. Unmapped reads and mapped reads with a quality (MQ) below 20 were then discarded. Potential PCR duplicates were then flagged using MarkDuplicates v. 1.140 (Picard tools⁴⁷). Variant calling was then performed using GATK (v. 3.7⁴⁸). 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 or more of these conditions were discarded. For every group of species, we performed principal component 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 some point between the bird sampling campaign and the analysis of the raw sequencing data.

QUANTIFICATION AND STATISTICAL ANALYSIS

Gene models & orthology prediction

We used one reference genome for all species belonging to the same clade (Table S3). We used the genome and gene models of a medium ground-finch individual (*Geospiza fortis*; assembly GeoFor_1.0; GCF_000277835⁵⁸) for all Darwin's finches, a collared

flycatcher (*Ficedula albicollis*, GCF_000247815; assembly FicAlb_1.4⁵⁹) for all *Ficedula*, a zebra finch (*Taeniopygia guttata*; GCF_000151805; assembly taeGut3.2.4⁶⁰) for the Estrildidae, a Reunion grey white-eye (*Zosterops borbonicus*; GCA_007252995; assembly ZoBo_15179_v2.0¹²) for all *Zosterops*. We also used the genome of the willow warbler (*Phylloscopus trochilus*; GCA_002305835; assembly ASM230583v1¹⁵) and the great tit (*Parus major*; GCF_001522545.2; assembly Parus_major1.1⁶¹). For all the investigated chaffinches, we used a newly generated assembly of *Fringilla coelebs* (Methods S1, version “HiRise” of Recuerda et al.⁶²). For this latter species, as well as for the willow warbler (*P. trochilus*), no gene models were available and we therefore first performed a protein homology detection and intron resolution using genBlastG⁵² (<http://genome.sfu.ca/genblast/download.html>) with the following options “-p genblastg -c 0.8 -r 3.0 -gff -e 1e-10.”

To analyze the same orthologous sequences in all species, we used the set of 8253 orthologs identified by Jarvis et al.¹⁶ (<http://gigadb.org/dataset/101041>). Then, we added the sequence of our species to this set of orthogroups using the method described in Scornavacca et al.⁶³ Briefly, each orthogroup was used to build an HMM profile using the HMMER toolkit.⁵³ Then, for each new sequence, hmmscan 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 hmmscan were considered to be 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.⁶³

Effective population size estimates

Historical demographic variations in N_e were estimated using the Pairwise Sequentially Markovian Coalescent (PSMC) model implemented in the software PSMC.⁴⁸ Fasta sequences were converted to the PSMC fasta format using a C++ program (Fasta2PSMCFasta: <https://osf.io/uw6mb/>) written using BIO++ library.⁶⁴ Only scaffolds longer than 500Kb were considered. We used block length of 100bp, with no more than 20% of missing data per block, as implemented in “fq2psmcfa” (<https://github.com/lh3/psmc>).

For each species, PSMC analyses were run using two randomly selected individuals. To identify suitable parameters, several -t and -p parameters were tested including -p “4+30*2+4+6+10” (as in Nadachowska-Brzyska et al.⁶⁵) and -p “4+25*2+4+6” (as in Kim et al.⁶⁶) but also -p “4+10*3+4” and -p “5*1+25*2+6.” The best combination (t15 -r4 -p “5*1+25*2+6”) was manually chosen after excluding other parameter values leading to large differences between the two individuals from the same species. Then, we randomly selected one individual and excluded the first four atomic time intervals to exclude the noisy estimates generally generated by PSMC for very recent times and therefore strengthen the reliability of the average estimates of N_e over the last million years.

Time was scaled assuming a mutation rate of 4.6×10^{-9} mutation/site/generation as estimated⁶⁷ and a generation time of 2 years.^{65,68} Results were plotted in R (v3.6.3⁵⁵) using the function “psmc.results” [69; <http://datadryad.org/resource/doi:10.5061/dryad.0618v4>] and with the R packages ggplot2⁶⁹ and cowplot.⁷⁰

Summary statistics of the polymorphic data

π_S and π_N/π_S ratios were computed using seq_stat_coding from reconstructed fasta sequences (Methods S1) using a publicly available Bio++ script and a procedure previously described⁴² (<https://osf.io/uw6mb/>). We empirically validated that our π_S and π_N/π_S estimates were not impacted by the variable number of samples per species. In addition, we used the π_N/π_S estimates based on the site frequency spectra at both non-synonymous and synonymous sites as described in Rousselle et al.¹⁸ to check the accuracy of these estimates (see also Table S2). Guanine-Cytosine (GC) content at third-codon positions of protein-coding genes (hereafter GC3), an excellent proxy of the local recombination rate in birds⁷¹ was also computed under seq_stat_coding. To estimate the within-genome variation in the efficacy of selection, we estimated π_N/π_S on sets of genes representing a total concatenated coding alignment of 2 Mb, after sorting genes by ascending values of GC3. The last window corresponding to genes exhibiting the highest GC3 values was only considered if this window contained at least 1 Mb of coding sequence.

Summary statistics of the divergence data

We used the method implemented by Galtier²¹ (Grapes. v1.0) to estimate α , ω_A and ω_{NA} using the approach introduced by Eyre-Walker & Keightley.⁷² Briefly, we fitted both a negative Gamma distribution and an exponential distribution to the synonymous and non-synonymous Site Frequency Spectrum (SFS) (the so-called GammaExpo model²¹) to model the distribution of fitness effect (DFE). Fitted parameters of the DFE were then used to compute the expected d_N/d_S under near neutrality (i.e., without adaptive substitutions but including weakly deleterious substitutions), which was compared to the observed d_N/d_S to estimate the adaptive substitution rate (ω_A) and the proportion of adaptive substitutions (α) [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.⁷³

Statistical analyses

All statistical analyses were performed using R.⁵⁵ We only considered models with a similar number of observations and compared these models based on the Akaike information criterion with a correction for small sample sizes (AICc). To test for the influence of the explanatory variables on π_S and π_N/π_S , we used Phylogenetic Generalized Least Square (PGLS) models. Explanatory variables were always log₁₀-transformed as this violated less frequently the assumption of normality, heteroscedasticity and independence of the residuals using a simple linear model. For PGLS, we used the model implemented in the “nlme” package.⁷⁴ The mitochondrial phylogeny was considered as the species tree (Figure S1, newick file is available at <https://osf.io/uw6mb/>) taken into account assuming a

Brownian correlation structure (using “corBrownian” from the “ape” package⁷⁵). P value and AICc 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. Therefore, the results of all the tests including with and without phylogenetic controls, transformed and untransformed π_S and π_N/π_S are presented in Table S1.

R plots were generated using a series of R packages: cowplot,⁷⁰ ggplot2,⁶⁹ ggpubr,⁷⁶ ggrepel⁷⁷ and ggtree.⁷⁸