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Concurrent invasions by European starlings (*Sturnus vulgaris*) suggest selection on shared genomic regions even after genetic bottlenecks

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Abstract

A species' success during the invasion of new areas hinges on an interplay between demographic processes and the outcome of localized selection. Invasive European Starlings (*Sturnus vulgaris*) established populations in Australia and North America in the 19th century. Here, we compare whole-genome sequences among native and independently introduced European Starling populations from three continents to determine how demographic processes interact with rapid adaptive evolution to generate similar genetic patterns in these recent and replicated invasions. Our results confirm that a post-bottleneck expansion may in fact support local adaptation. We find that specific genomic regions have differentiated even on this short evolutionary timescale, and suggest that selection best explains differentiation in at least two of these regions. This infamous and highly mobile invader adapted to novel selection (e.g., extrinsic factors), perhaps in part due to the demographic boom intrinsic to many invasions.

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1 Introduction

Some species can establish and spread in a novel environment more 2 3 successfully than others, and defining what makes a species 'invasive' is hotly 4 contested¹⁻⁴. Invasion biologists continue to debate whether an invasion's success can 5 be better attributed to an intrinsic property of the founding population or to extrinsic 6 conditions experienced by the population. When invasive populations colonize a new environment, they often undergo genetic bottlenecks⁵. However, even populations with 7 limited genetic diversity, including those subject to founder effects, can adapt quickly to 8 novel environments⁶⁻⁸. For example, gene surfing during range expansion has led to 9 adaptation in experiments in microbes ⁹ and wild, invasive bank voles ¹⁰. Simultaneous 10 with demographic shifts, invasive populations may encounter environmental conditions 11 that exert novel selective regimes. Although genetic bottlenecks often limit the genetic 12 variation available to selection, local adaptation can occur even after a genetic 13 bottleneck^{6,11-13}. 14

Recent and replicated invasions are particularly useful for exploring the eco-15 evolutionary dynamics of population expansion¹⁴⁻¹⁶, since any divergence after 16 17 introduction likely reflects a combination of demographic and/or selective forces. One such recent invader is the Common or European Starling (Sturnus vulgaris), which was 18 introduced across south-eastern Australia in 1856 and to New York, United States of 19 20 America in 1890¹⁷. Both the American and Australian invasions were most likely founded by individuals from the United Kingdom^{18,19}, although multiple introductions in 21 Australia might contribute to ongoing gene flow among continents.²⁰ 22

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23	The starling's ecology is well-studied in both invasions, enabling us to explore
24	how environmental factors might impact genetic variation ²¹ . Native-range starlings thrive
25	in open pastures and urban environments, but starlings' ecology and life history vary
26	among populations. In range-wide studies of Australian ²²⁻²⁴ and North American ^{25,26}
27	invasions, temperature and precipitation influence genetic differentiation, even after
28	controlling for population structure. Migration and dispersal also vary among invasions.
29	In North America, starlings can migrate long distances each season ^{19,27} , but populations
30	in the Western USA likely disperse and/or migrate shorter distances ²⁸ . In contrast,
31	starlings in Australia exhibit strong population structure and likely migrate short
32	distances in search of food, even though environmental conditions are much more arid
33	than in the native range ²² . Finally, variation in the breeding cycle may also facilitate
34	invasion success as invasive populations tend to lay more than one clutch, whereas the
35	UK population generally lays only one clutch ^{29,30} .
26	We use concurrent starling invasions in Australia (ALI) and North America (NA) to

36 We use concurrent starling invasions in Australia (AU) and North America (NA) to 37 examine the evolutionary and genetic consequences of invasion. Both starling invasions rapidly expand from small founding populations in the late 19th century^{18,19}. Given this 38 39 natural control, we take advantage of a rare opportunity to compare intrinsic and extrinsic drivers of invasion success. Founder effects and other intrinsic demographic 40 properties of invasions certainly influence establishment ^{31,32}, and we predict some 41 42 divergence from the native, ancestral population (represented by UK samples here) due to genetic drift. We combine demographic models with fine-scale measures of genomic 43 diversity and differentiation to determine whether drift alone can explain observed 44 45 genetic variation, and we find strong evidence that extrinsic factors such as novel

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46 selective pressures generate differentiation in both invasions specific to only a few47 genomic regions.

48

49 Results

50 Differentiation and population structure

51 We sequenced whole genomes of eight native-range individuals from Northumberland,

52 UK and eight from each of the two invasions (NA: all 8 from New York City, USA; AU: 2

each from four locations in New South Wales, Australia, Table S1). After filtering, we

obtained more than 11 million SNPs with minimum 5X coverage (average genome-wide

55 coverage = 18.44, for details on how choice of reference genome impacts variant-

56 calling, see SI Section 1). All patterns identified using a variant-called dataset concur

57 with those based on genotype likelihood (ANGSD); for details on how variant-calling

impacts patterns, see SI Section 2. We use a variant set filtered for minor allele

59 frequency, Hardy-Weinberg Equilibrium, and linkage disequilibrium for analyses of

60 population structure, but for more accurate estimates of genetic diversity and

differentiation, we report results from a genome scan of variants filtered only for quality

and depth. Differentiation of the two invasive populations from the native population is

low, which is expected given that these populations split less than two centuries ago.

However, differentiation between AU and UK (F_{ST} AU vs. UK = 0.016) is almost twice

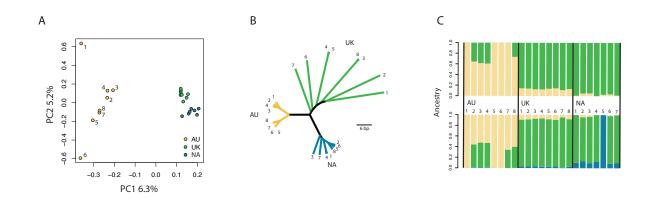
that between NA and UK (F_{ST} NA vs. UK = 0.008). We examined this contrast in genetic

66 differentiation using analyses of population structure.

All three populations are readily distinguished from one another by a principal
 component analysis of 40,488 unlinked SNPs (minor AF > 0.25): PC1 (6.3% of genetic

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69	variation) separates the UK and NA populations from the Australian population (Figure
70	1A). This evidence complements previous work that showed extensive population
71	structuring in Australia but nearly continuous gene flow across North America, based on
72	reduced-representation genomic data ^{22,33} . Principal component analysis in the genotype
73	likelihood framework of ANGSD ³⁴ shows nearly identical results (Figure S6).
74	Furthermore, individuals are reliably assigned to clades based on pairwise genetic
75	distances calculated in ANGSD (Figure 1B).
76	We note that the tight clustering of UK individuals in Figure 1A contrasts with the
77	large distances between these same individuals in Figure 1B. In the genotype likelihood
78	dataset, genetic distances between native UK individuals are much greater than
79	distances among individuals within each invasive population (Figure 1B). Because
80	these two datasets differ in variant-calling and filtering strategies, the genetic distance
81	among UK individuals in Figure 1B may reflect rare alleles that were filtered out of the
82	variant set in Figure 1A. However, PCs 3 and 4 in the PCA of the variant-called dataset
83	do indicate additional structure within the UK population (Figure S7).
84	



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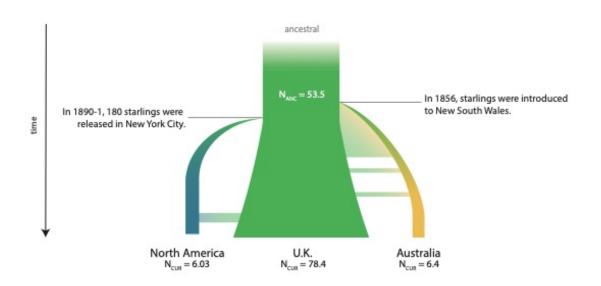
86	Figure 1. A) Principal components of 40,488 unlinked SNPs explain 6.3% (PC1) and
87	5.2% (PC2) of genetic variation. B) cladogram of genetic distances among samples
88	based on genotype likelihoods of 16,151,007 sites. C) ADMIXTURE analyses showing
89	K=2 (top) and K=3 (bottom row).
90	
91	Admixture analyses revealed statistical support for a two-population model,
92	which distinguished AU from NA+UK and was only slightly weaker (cross-validation
93	error = 0.88) than the best-supported model of K=1 (cross-validation error = 0.73, Figure
94	1C, SI Section 3). Regardless of the number of populations hypothesized in models of
95	admixture, the NA invasion consistently shares a higher proportion of its ancestry with
96	the UK population. Individuals from the Australian population are distinguished from the
97	other two populations in all tested values of K. Considered in concert, these tests of
98	population structure show that Australian and North American populations differ in the
99	amount of divergence from the native UK population. Founder effects likely contribute to
100	the observed population structure, and below we describe explicit models of
101	demographic processes.
102	
103	Invasive and native populations experienced bottlenecks and subsequent
104	expansion
105	To address the impact of demographic processes in generating the observed
106	patterns we used the site-frequency spectrum built from genotype likelihoods to
107	construct models of changes in the effective population size over time. We examined
108	the demographic history of all three populations using fastsimcoal2 ³⁵ . The demographic

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109	model shows that both invasions experienced a bottleneck upon introduction (Figure 2),
110	which is exactly what we expect given the small number of founding individuals in both
111	AU and NA. Bottlenecks often lead to inbreeding within a population, and we find that
112	inbreeding is negligible but slightly higher in the Australian population than in the NA
113	population (Table S2). In addition, relatedness among individuals is quite low, where
114	zero indicates no shared alleles among individuals (maximum AJK statistic = 0.06).
115	Each invasion appears to have recovered quickly by expanding in effective population
116	size, and at present, our data suggest similar effective population sizes in both the AU
117	and NA invasions. Much of the ancestral variation appears to be shared among
118	invasions, given that the genome-wide average F_{ST} between AU and NA is 0.04,
119	although we note that genetic differentiation among invasions confirms the expectation
120	that different variants will make it through each genetic bottleneck. The results
121	presented here concur with range-wide sampling that indicates genetic bottlenecks
122	followed by rapid expansion with little evidence of inbreeding in both Australia ²² and
123	North America ³³ .

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Figure 2. Demographic model of effective population size based on the site-frequency
spectrum. Schematic approximates population growth based on model output from *fastsimcoal2*.

129

Population bottlenecks often lead to a loss of genetic diversity when genetic drift 130 drives to fixation alleles once maintained at a moderate frequency. Since both the AU 131 132 and NA invasions experienced a genetic bottleneck, we expect that nucleotide diversity 133 (π) within each population should be higher in the older and larger native range population than in each invasive range. However, we find that 70% of 50-kb windows 134 show greater nucleotide diversity in AU (compared to UK diversity), and 74% show 135 136 greater diversity in NA. The filters applied should not lead to disproportional allelic dropout of rare alleles, because we did not filter for minor allele frequency prior to this 137 138 scan. Higher invasive diversity may be a sampling artifact: rare variants in the native 139 range may have 'surfed' to a higher frequency in the invasions, and our sampling of the 140 native range (in Northumberland, UK) likely represents only a small portion of genetic

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variants that could have been introduced in AU and NA. We also note that the UK
individuals sampled here do not capture range-wide diversity in the native population,
and therefore we expect that actual nucleotide diversity in the UK population is higher
than what we have sampled here. Additional investigation of native range starlings will
be needed to determine whether lower diversity in the UK is recovered with broader
geographic sampling.

Our demographic models show that both the AU and NA invasions rapidly 147 expanded in population size after the initial bottleneck, which predicts that many loci 148 149 would be lost upon establishment of the invasive populations even as variants in other 150 loci increase in frequency. If gene surfing facilitated by rapid population expansion 151 explained genome-wide differentiation, then we would expect to find that regions where 152 invasive diversity is higher than diversity in the native range are distributed across many chromosomes. However, shifts in diversity and differentiation that occur only in a few 153 154 narrow regions of the genome would suggest evolutionary dynamics specific to that 155 region. If regions that have differentiated (e.g., high F_{ST}) from the native range also 156 show higher diversity in the invasions, this may suggest either a relaxation of purifying 157 selection or an increase in diversifying selection at specific regions of the genome. We note here that the recombination landscape may also contribute to these patterns, and 158 159 discuss this factor in the following section.

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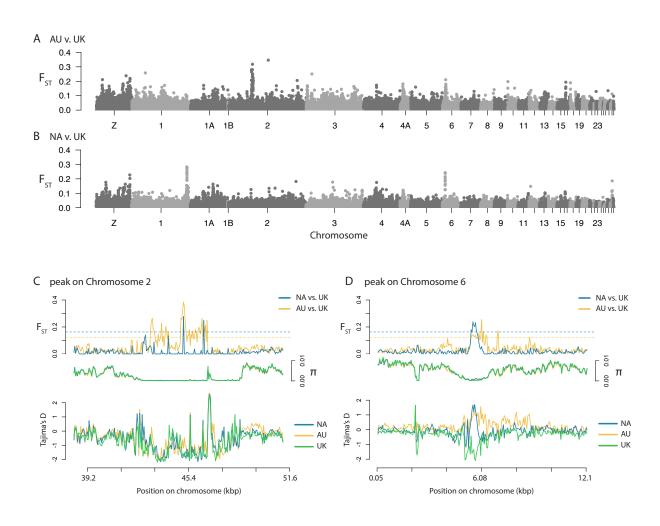
161 Differentiation of a few genomic regions reveal consequences of invasion

162 Starling populations colonized their invaded ranges less than two centuries ago, and the

age of these populations makes it somewhat surprising to find loci specific to the

164	derived populations with fixation indices as high as 0.57 in a single 10-kb window
165	(Figure 3). However, no putative outlier windows approach fixation in any comparison
166	across all individuals in a population. We consider only the top 0.1% of 50-kb windows
167	to be F_{ST} outliers (AU vs. UK: F_{ST} > 0.30; NA vs. UK: F_{ST} > 0.22); subsequent
168	references to outliers indicate these windows. Outliers include Chromosomes 1, 2 and
169	6, and to a lesser extent, 1A, 4, and 4A. Examining localized shifts in diversity and
170	differentiation at these regions can help to resolve the relative contributions of
171	demography and selection.
172	Genome scans like this one tend to search for genomic islands of divergence,
173	where recombination can explain divergence among populations ³⁶ . High levels of
174	differentiation paired with a reduction in diversity may stem from suppressed
175	recombination (e.g., proximity to the centromere or a structural rearrangement), or from
176	alleles approaching fixation or loss due to drift or selection (e.g., a selective sweep). We
177	find that most putative outlier regions are distant from the centromere location
178	(predicted via homology with the zebra finch (Taeniopygia guttata) genome,
179	Supplementary Information). Nevertheless, genomic architecture—including linkage
180	disequilibrium independent of the centromere—plays a role in how differentiation among
181	populations is generated. Below, we discuss how demography and selection drive
182	divergence in invasions, considering the context of the starling's genomic architecture.
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Figure 3. A-B) Manhattan plots show 50-kb windowed F_{ST} between AU & UK (A) and NA & UK (B) populations. C-D) 50-kb windowed F_{ST} (top row), π within each population (middle), and Tajima's *D* within each population (bottom row), centered on the elevated F_{ST} regions of Chromosome 2 (C) and Chromosome 6 (D). Color represents each population, except in F_{ST} plot where yellow indicates F_{ST} AU vs. UK and blue indicates F_{ST} NA vs UK.

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194 Some F_{ST} outlier regions show clear signals of selection in one or both invasions

If directional selection were driving differentiation between an invading population 195 and its native ancestral population, we would expect to see a decline in nucleotide 196 197 diversity specific to the invading population. But, as described above, local reductions in π could also result from population bottlenecks experienced during founder events; to 198 clarify the impact of selection, we look Tajima's D. We infer directional selection where 199 200 most variants are either very rare or very common (e.g., negative Tajima's D). We 201 examine these predictions by looking at two F_{ST} peaks on Chromosome 1A; in this 202 region, we find very low nucleotide diversity and negative Tajima's D in both AU and 203 NA, which is exactly the signature we would expect under purifying selection (Figure 204 S12). Around that region, Tajima's D in the UK population varies between 0 and 1. We note that genetic distance (dxy) could also clarify mechanisms, but as dxy tracks FST 205 206 perfectly, we argue that linked selection offers the clearest explanation. This pattern is 207 echoed on Chromosomes 4 and 4A, but the remaining outlier regions show more 208 complicated signals (Figures S13-14).

Where F_{ST} is highest on Chromosome 2, we find strong evidence of both 209 210 purifying and balancing selection in all three populations (Figure 3C). We find that 211 nucleotide diversity is very low within every population, and immediately after the block 212 of elevated F_{ST}, we see a sharp increase in nucleotide diversity in all three populations (Figure 3C). In addition, Tajima's D tracks these changes in diversity: where F_{ST} and π 213 are low, Tajima's *D* is also negative in all three populations—indicating an excess of 214 low-frequency variants and perhaps purifying selection—but this statistic climbs to high 215 216 positive values immediately before and after the block of elevated F_{ST}, indicating

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217	balancing selection. The concordance of Tajima's <i>D</i> in all three populations suggests a
218	release of some kind, whether it is a relaxation of purifying selection or a recombination
219	breakpoint. Even though the centromere is predicted to be 30-Mb downstream of this
220	region, these signatures are consistent with linkage disequilibrium in this 4-Mb region:
221	eukaryotes generally show suppression of recombination near the centromere, leading
222	to a build-up of linkage disequilibrium if this suppression extends for 30-Mb. It is
223	possible that a structural variant in the founding population could generate this pattern.
224	However, we note that F_{ST} among AU and the other populations in fact declines
225	dramatically (to around 0.1) in the middle of the 4Mb region, and in the same location,
226	$F_{\mbox{\scriptsize ST}}$ between NA and UK increases slightly. If this genomic region differentiated as a
227	large linkage block, we would not expect such a decline in F_{ST} and a weakening of
228	selective pressure (as evidenced by the increase in Tajima's D). For these reasons, we
229	suggest that the peak on Chromosome 2 indicates both purifying and balancing
230	selection in the AU invasion.
231	
232	One region on Chromosome 6 reveals how population expansion could interact
233	with selection
234	Both invasions have differentiated from the native range in a 4-Mb region of
235	Chromosome 6 (Figure 3). As a preliminary check, we note that the large distance
236	between this F_{ST} peak and the centromere suggests that low recombination is unlikely to
237	explain differentiation. We suggest that the clearest explanation of this peak invokes

238 selection on previously rare variants, based on three lines of evidence.

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239	First, we suggest that rapid population expansion allowed previously rare
240	variants to surf to a higher allele frequency in the invasions. In this 4-Mb region,
241	invasive diversity (π_{AU} and π_{NA}) are each more than three times the native diversity. This
242	shift in within-population diversity is not random; in fact, when we examine invasive
243	nucleotide diversity directly under the F_{ST} peak, we find only three other windows ~4.2-
244	Mb upstream of this peak show invasive diversity that is notably higher than native
245	diversity. This evidence supports the hypothesis that upon establishment, starlings
246	experienced either (1) balancing selection (strong positive Tajima's D) in both invasions
247	due to novel selective pressures or (2) a release of purifying selection that led to an
248	accumulation of variants and thus higher invasive diversity—but only in this specific
249	region. These patterns could be driven by a small number of individuals, or they could
250	indicate a population-wide shift, which leads us to our next point.
251	Second, in this region, we find that these higher-diversity alleles in both the NA
252	and AU populations have increased in frequency relative to the native range. In the

253 same region, we find strong positive values of Tajima's D in the invasions—indicating a 254 moderate-to-high frequency of the alternative allele—and negative Tajima's D in the UK 255 population at this peak, since these signatures suggest that previously rare variants 256 have increased in frequency in the invasions only. Alternatively-or simultaneously-257 purifying selection may have driven these same variants to a lower frequency in the UK 258 population. The most parsimonious explanation of these shifts in diversity is a single 259 event in the UK population, but we note that this shift is specific to only a small region of 260 the genome.

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261	Third, and most importantly, these patterns are found in this region of the
262	genome only, and it is notable that this shift in diversity co-occurs with one of the
263	highest F_{ST} peaks. We would expect to see similarly high invasive diversity under other
264	$F_{\mbox{\scriptsize ST}}$ peaks if population expansion alone could explain these patterns. However,
265	nowhere else in the genome do we find such high invasive diversity where native
266	diversity is low. We suggest a selective explanation given that a genetic bottleneck is
267	not likely to produce this pattern. Taken together, these results provide evidence that
268	rapid expansion of these starling invasions may have facilitated selection to drive
269	previously rare variants higher in frequency, independently in both NA and AU
270	populations.
271	

272 Genes under putative selection may aid in invasion success

273 The region under putative selection on Chromosome 6 overlaps with the coding regions 274 of four genes with dramatically different functions (JMJD1C, RTKN2, NRBF2, and 275 ARID5B), and we suggest that selection on one of these genes might explain the 276 differentiation in the region, with the other genes remaining in linkage disequilibrium with 277 the possible candidate. Among these candidates, we can speculate that ARID5B has 278 the most intuitive link to hypothesized selective pressures: this protein is required for adipogenesis and involved in smooth muscle differentiation. The first exon of this gene 279 280 lies directly under the F_{ST} peak between AU and UK starlings, and muscle growth and 281 fat storage may have been key to dispersal ability. The three other genes that overlap this window are involved in the DNA-damage response (JMJD1C), lymphopoiesis 282 283 (*RTKN2*), and regulating autophagy (*NRBF2*). For details on GO term enrichment in

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these outlier regions, see SI Section 6. Regardless of the mechanism driving these loci
toward an intermediate frequency, it remains possible that variation at one or more of
these loci influenced invasion success by maintaining heterozygosity.

287

288 Conclusion

289 An open question in invasion biology is whether an invasive population's success is better attributed to intrinsic properties of the invasive species or to extrinsic factors 290 specific to the novel environment. These results demonstrate how demographic shifts 291 292 during and immediately after any establishment may support a species' success and 293 even lead to local adaptation to environmental conditions. The Australian and North 294 American starling invasions colonized each continent around the same time, and 295 experienced similar contractions in population size that led to classical founder events upon establishment. The shared decline in genetic diversity represents a shared 296 intrinsic determinant of invasion success. However, differences in propagule pressure 297 298 and dispersal likely influence the evolutionary trajectory of each population ^{37,38}. 299 Propagule pressure (also termed introduction effort) is a composite measure of the 300 number of individuals released, and we note that founding population sizes may have varied slightly among AU and NA introductions. Even without dramatic variation in 301 302 founding population sizes, dispersal itself shapes genetic diversity and thus adaptive potential, as shown in a recent study of invasive plants ³⁹. Starlings' dispersal and 303 migration varies among populations: starlings in Australia tend to migrate less frequently 304 305 and across shorter distances, but starlings migrate and/or disperse hundreds of kilometers in North America^{19,40} and South Africa⁴¹. 306

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307	It is remarkable that despite this contrast in life history strategies and the
308	stochastic nature of evolution during range expansion, we find $F_{\mbox{\scriptsize ST}}$ peaks shared among
309	invasions at only a few regions of the starling genome. Although these F_{ST} peaks could
310	arise via drift, footprints of other population genetic metrics are consistent with selection.
311	We note that mutations in specific chromosomal regions could also be accelerated by
312	extrinsic environmental properties (climate, food availability, and more) through
313	epigenetic CpG DNA hypermethylation events, which are known to increase frequency
314	of genetic mutations by spontaneous deamination CG>TG transition ⁴²⁻⁴⁴ . For example,
315	such epigenetic shifts supported the invasion of another avian species (the house
316	sparrow) into Australia ⁴⁵ . Regardless of genetic mechanism, we suggest that
317	differentiation in these genetic regions is simultaneously shaped by intrinsic and
318	extrinsic drivers of invasion success. We find it notable that some differentiated regions
319	(in particular, Chromosome 6) are shared among invasions despite differences in the
320	selective environment as well as stochastic processes that shape the starling's
321	evolution on each continent. The European starling invasions compared here suggest
322	that rapid population growth may support local adaptation.

323

324 Methods

325 Whole-genome re-sequencing

Libraries for each individual starling were constructed using a TruSeq DNA PCRfree High Throughput Library Prep Kit (Illumina, San Diego, CA). All individuals passed the initial quality check with FASTQC (Babraham Bioinformatics, Cambridge, UK). Adapters were removed using AdapterRemoval⁴⁶ and reads mapped to the reference *S*.

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vulgaris vNA genome (GCF 001447265.1)⁴⁷ genome using BOWTIE2⁴⁸ and checked for 330 mapping quality using qualimap⁴⁹. Sequencing quality was relatively high: 96.4% of 331 332 reads mapped to the S. vulgaris genome with a coverage of 18.4X and a mapping 333 guality of 26.9 (Table S1). Reads were also mapped to a pseudo-chromosome-level S. vulgaris vNA genome, where scaffolds were assigned to chromosomes based on 334 orthology to the zebra finch reference genome (GCF 000151805.1)⁵⁰. Assuming 335 orthology, we were able to predict centromere positions based on the known genomic 336 architecture of the zebra finch⁵¹, but this study does not directly define centromere 337 338 position. We called variants using GATK's HaplotypeCaller in GVCF mode and flagged 339

low-quality variants using GATK Best Practices (QD<2, FS>60, MQ<40, and SOR>3,

accessed March 21, 2018 ⁵²). We filtered sites for missing data, depth, and quality using

vcftools (parameters: --max-missing 0.8 --min-meanDP 2 --max-meanDP 50 --remove-

filtered-all), which removed 4.1 million sites from the original SNP set and left a total of

23.4 million sites for downstream analyses. Starting from the mapped reads used in the

345 GATK pipeline, we also called SNPs based on a minimum p-value of the correct

346 genotype probability at each site using ANGSD^{34,53}. Filtering for SNP p-value (0.0001),

depth (between 60 and 400 sequences), and mapping quality (>20) left 16,151,007

348 sites. All scripts used in read processing and filtering are available on GitHub:

349 <u>https://github.com/nathofme/global-RESEQ</u>.

350

351 Population structure

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352	Population structure analyses used a dataset of biallelic SNPs in Hardy-
353	Weinberg Equilibrium, where minor allele count (MAC) is > 2 and SNPs were pruned for
354	LD by removing all sites with an $r^2 > 0.6$ within 1kb sliding windows. This filtering left
355	868,685 sites. Since some individuals showed much lower coverage (minimum 5.58X),
356	all tests of population structure were run with both variant-called (GATK) and genotype
357	probability (ANGSD) datasets. Scripts for variant-called analyses are stored at

- 358 https://github.com/nathofme/global-RESEQ/blob/master/filter-scan.sh, and scripts for
- 359 probability-based analyses at https://github.com/nathofme/global-
- 360 <u>RESEQ/blob/master/angsd.sh</u>.
- 361 We estimated variance among and between individuals using a principal
- 362 components analysis in SNPRELATE⁵⁴ (GATK) and a covariance matrix built in
- 363 ngsTools⁵⁵ (ANGSD). We used ADMIXTURE ⁵⁶ (GATK) and NGSADMIX⁵⁷ (ANGSD) to
- 364 examine shared ancestry among individuals, and we also measured pairwise genetic
- 365 distances using ngsDist for the ANGSD dataset only.
- 366
- 367 Demographic inference

We used FASTSIMCOAL2³⁵ to explicitly test for genetic bottlenecks in each population.
FASTSIMCOAL2 takes a site-frequency spectrum (SFS) as input, and we used the SFS
estimated from ANGSD given that likelihood-based estimates are more robust to
sequencing error⁵⁸. Demographic models in *fastsimcoal2* used priors on the time (TBOT
= 10 to 300) and size of the bottleneck (NBOT = 10 to 1000). The command line
arguments were as follows: -M -n 1000000 -L 50 -q -k 100000. Each run began with a
randomly generated seed (-r), and the -k flag simply writes polymorphic sites to a

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temporary file to cope with the high memory usage of this analysis. Scripts for

376 demographic analyses can be found at <u>https://github.com/nathofme/global-</u>

377 <u>RESEQ/blob/master/demography.sh</u>.

378 To verify this demographic model, we also estimate inbreeding coefficients (F-

379 statistics) using the –het command in VCFTOOLS, and calculate relatedness among each

pair of individuals using the –relatedness method of Yang et al. (2010) in VCFTOOLS.

381

382 Sliding window scans

383 For scans of genetic divergence and diversity, we used a variant dataset filtered only for depth and quality: we kept variants that had less than 20% missing data across all sites, 384 385 but did not apply minor allele frequency (MAF) filters, filter for HWE, linkage, or any 386 other factors. Given that rare alleles likely provide the strongest evolutionary signals in this system, we did not want to filter out any alleles that might have been rare in one 387 population (e.g., the native UK population) and increased in frequency in another 388 389 population (e.g., the AU or USA invasions). Nevertheless, we test for sensitivity to this filtering choice in SI Section 2B. 390

We calculated F_{ST} and nucleotide diversity (π) using overlapping 50-kb sliding windows with a step size of 10kb using VCFTOOLS ⁵⁹. We calculate nucleotide diversity separately for each population. We then calculate d_{xy} using a Python script by Simon Martin (accessed at

395 https://github.com/simonhmartin/genomics_general/blob/master/popgenWindows.py_on

396 February 12, 2020). This analysis includes all confident variant calls (parameters: --

397 output-mode EMIT_ALL_CONFIDENT_SITES in GATK's HaplotypeCaller); only with

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398	this modification do we recover levels of d_{xy} similar to other systems. We also measured
399	F_{ST} and π in overlapping 10-kb windows to localize elevated F_{ST} to an even smaller
400	region of the genome. To visualize relationships between diversity metrics, we plotted
401	the mean values of each metric in a 50-kb window. All scripts for plotting are stored on
402	GitHub: https://github.com/nathofme/global-RESEQ/.
403	
404	Identifying candidate genes under selection
405	In contrast to identifying single genes, network analyses of gene ontology (GO) terms
406	can provide a more holistic and objective method of identifying shared functions.
407	Network analyses can also provide more statistical power, correcting for the usual
408	problem of multiple testing. To identify functions of candidate regions, we quantified the
409	uniqueness and dispensability of each GO term using a method that quantifies semantic
410	similarity ⁶⁰ . This analysis by default emphasizes GO terms that are rare in the list of
411	candidates provided; because we are interested in functions that are common across
412	outlier regions, we manually curate category labels to choose GO terms that are less
413	unique and more dispensable as representative titles.
414	
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- 426
- 427
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- 430 Field work: Melissa Bateson, Lee Ann Rollins, Adam Cardilini, Scott J Werner
- 431 Lab work: Wes Warren and the McDonnell Group at WUSTL.
- 432 Data analysis and manuscript writing: Natalie R Hofmeister, Katarina Stuart
- 433 Manuscript editing: all authors
- 434

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