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1 **Genome-wide differentiation in closely related populations: the roles of selection**  
2 **and geographic isolation**

3  
4 Safran RJ\*<sup>1</sup>, Scordato ESC<sup>1</sup>, Wilkins MR<sup>1,2</sup>, Hubbard, JK<sup>1,2</sup>, Jenkins, BR<sup>1</sup>, Albrecht T<sup>3</sup>,  
5 Flaxman SM<sup>1</sup>, Karaardıç H<sup>4</sup>, Vortman Y<sup>5</sup>, Lotem A<sup>5</sup>, Nosil P<sup>6</sup>, Pap P<sup>7</sup>, Shen S<sup>8</sup>, Chan,  
6 S-F<sup>8</sup>, Parchman T<sup>9</sup>, Kane NC<sup>1</sup>

7  
8 \* Corresponding author

9  
10 <sup>1</sup> Department of Ecology and Evolutionary Biology, University of Colorado, Boulder, CO,  
11 USA 80309

12  
13 <sup>2</sup> School of Biological Sciences, University of Nebraska-Lincoln, USA

14  
15 <sup>3</sup> Department of Zoology, Charles University in Prague and Institute of Vertebrate  
16 Biology, Academy of Sciences of the Czech Republic, Czech Republic

17  
18 <sup>4</sup> Elementary Science Education Department, Education Faculty, Alanya Alaaddin  
19 Keykubat University, Alanya, Turkey

20  
21 <sup>5</sup> Department of Zoology, Tel-Aviv University, Israel

22  
23 <sup>6</sup> Department of Animal and Plant Sciences, University of Sheffield, Sheffield, UK

24  
25 <sup>7</sup> Department of Taxonomy and Ecology, Babeş-Bolyai University, Cluj-Napoca,  
Romania

26  
27 <sup>8</sup> Biodiversity Research Center, Academia Sinica, Taiwan.

28  
29 <sup>9</sup> Department of Biology, University of Nevada, Reno, NV, USA  
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36 **Abstract**

37

38 Population divergence in geographic isolation is due to a combination of factors. Natural  
39 and sexual selection may be important in shaping patterns of population differentiation,  
40 a pattern referred to as 'Isolation by Adaptation' (IBA). IBA can be complementary to the  
41 well-known pattern of 'Isolation by Distance' (IBD), in which the divergence of closely  
42 related populations (via any evolutionary process) is associated with geographic  
43 isolation. The barn swallow *Hirundo rustica* complex comprises six closely related  
44 subspecies, where divergent sexual selection is associated with phenotypic  
45 differentiation among allopatric populations. To investigate the relative contributions of  
46 selection and geographic distance to genome-wide differentiation, we compared  
47 genotypic and phenotypic variation from 350 barn swallows sampled across eight  
48 populations (28 pairwise comparisons) from four different subspecies. We report a draft  
49 whole genome sequence for *H. rustica*, to which we aligned a set of 9,493 single  
50 nucleotide polymorphisms (SNPs). Using statistical approaches to control for spatial  
51 autocorrelation of phenotypic variables and geographic distance, we find that  
52 divergence in traits related to migratory behavior and sexual signaling, as well as  
53 geographic distance together, explain over 70% of genome-wide divergence among  
54 populations. Controlling for IBD, we find 42% of genome-wide divergence is attributable  
55 to IBA through pairwise differences in traits related to migratory behavior and sexual  
56 signaling alone. By (i) combining these results with prior studies of how selection  
57 shapes morphological differentiation and (ii) accounting for spatial autocorrelation, we  
58 infer that morphological adaptation plays a large role in shaping population-level  
59 differentiation in this group of closely related populations.

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63 Keywords: climate variability, genomic divergence, Genotyping By Sequencing, reproductive  
64 isolation, population genetics, speciation

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66

## 67 **Introduction**

68           Species are thought to arise primarily as populations diverge in spatial isolation  
69 and accumulate differences that preclude their reproduction upon secondary contact  
70 (Mayr 1963, Coyne and Orr 2004). This has been argued to be particularly true for birds  
71 (Price 2008). There is also long-standing interest in the extent to which population  
72 divergence can arise or increase in the context of incomplete spatial isolation, coupled  
73 with selection against immigrants or hybrids to offset homogenizing genetic exchange  
74 ('divergence with gene flow': Nosil et al. 2008, 2009). More generally, spatial isolation is  
75 a continuum, and divergence can arise as a consequence of: spatial isolation and  
76 genetic drift (isolation by distance, IBD; Wright 1943, Slatkin 1993); spatially varying  
77 natural and sexual selection (isolation by adaptation, IBA, or a subset of IBA: isolation  
78 by environment, IBE; Rundle and Nosil 2005, Crispo et al 2006, Nosil et al. 2009,  
79 Schluter 2001, 2009, Shafer and Wolf 2013); or both (Lande 1980, Shafer and Wolf  
80 2013, Wang and Bradburd 2014). While the contributions of spatial isolation and  
81 divergent selection to population differentiation have been research areas in  
82 evolutionary biology since the beginning of empirical population genetics (Fisher 1930),  
83 only more recently have we begun to characterize genomic divergence by simultaneous  
84 tests of IBA and IBD (e.g., Wang and Summers 2010, Wang et al. 2013, Bradburd et al.  
85 2013, Wang and Bradburd 2014).

86           In comparison to accumulating evidence for the role of ecological adaptation (one  
87 form of IBA) and geographic isolation (IBD) in population divergence and speciation  
88 (Shafer and Wolf 2013, Wang and Bradburd 2014), there is little empirical data on the  
89 contribution of sexual selection to population genetic divergence (e.g., Parchman et al

90 2013, Baldassarre et al. 2014, Morgans et al. 2014). Adaptation to the local  
91 environment may come in the form of sexual selection if competition over mates differs  
92 among closely related populations or signal traits reflect variation in the environmental  
93 context in which these traits are developed or advertised (Ingleby et al. 2010, Maan and  
94 Seehausen 2011, Baldassarre et al. 2013, Safran et al. 2013, Matute, 2014, Seehausen  
95 et al. 2014). The lack of such data represents a major gap in our understanding of  
96 speciation because divergence in sexual traits is a common form of phenotypic  
97 differentiation among populations that may strongly relate to reproductive isolation (e.g.,  
98 West-Eberhard 1983, Panhuis et al. 2001, Irwin et al. 2001, Ritchie 2007, Curie Network  
99 2012, Safran et al. 2012, Langerhans and Riesch 2013, Seddon et al. 2013). Moreover,  
100 previous work has shown that sexual signals can evolve more rapidly than traits related  
101 to ecological adaptation (e.g., Kingsolver et al., 2001, Svensson et al. 2006, Kingsolver  
102 and Pfennig 2007, Siepielski et al. 2011, Seddon et al. 2013). Collectively, these  
103 observations suggest that sexual signals could be an important feature of biological  
104 diversification, especially in the early, formative stages of speciation (Kraaijeveld et al.  
105 2011). Here, a testable prediction for a role of sexual selection in divergence is that  
106 divergence in aspects of phenotype related to mate selection is positively associated  
107 with genetic divergence, either due to increased local adaptation, reduced immigration,  
108 or both. This is similar to evidence for ecological speciation, where divergence in traits  
109 related to ecological adaptation predicts population genetic differentiation (Shafer and  
110 Wolf 2013).

111         Thus, in its most general form, IBA predicts associations between genomic  
112 divergence and differentiation in traits related to both ecology and mating success. Note

113 that ecological adaptations and mating adaptations are not mutually exclusive, and  
114 indeed, may even be related, as sexual signals necessarily evolve in an ecological  
115 context (Ingleby et al. 2010, Maan and Seehausen 2011, Baldassarre et al. 2013,  
116 Safran et al. 2013, Seehausen et al. 2014). Thus, important questions remain about the  
117 relative contributions and likely interactions of sexual selection, natural selection, and  
118 geographic distance to genomic differentiation among closely related populations (e.g.,  
119 van Doorn et al. 2009, Maan and Seehausen 2011, Wagner et al. 2012, Safran et al.  
120 2013). For example, sexual selection has been shown to interact with ecological context  
121 in the early stages of population divergence and the formation of premating barriers to  
122 reproduction (Scordato et al 2014), yet we know very little about the relative importance  
123 of each selective process in the accumulation of biologically relevant genomic  
124 differences among populations.

125       Closely related populations with variation in geographic distance are  
126 advantageous for identifying spatial and trait-based predictors of genomic divergence.  
127 To address the extent to which sexual and natural selection either interact or singly  
128 influence population divergence, we analyzed patterns of genomic divergence as a  
129 function of variation in aspects of phenotype known to be involved in mate selection and  
130 migration, as well as relevant measures of environmental variation (using elevation data  
131 and long-term climate databases), in a widespread, phenotypically divergent, yet young  
132 group of subspecies: barn swallows (*Hirundo rustica*). Barn swallows include six sub-  
133 species worldwide (Figure 1), with populations varying in: (1) geographic distance from  
134 one another, (2) trait combinations with known importance in sexual signaling, (3)  
135 breeding and non-breeding environments, and (4) migratory behavior (reviewed in

136 Scordato and Safran 2014). A recent phylogenetic reconstruction for this group of  
137 closely related sub-species indicates that they are monophyletic with respect to other  
138 members of the genus *Hirundo* (Dor et al. 2010). Additionally, recent mtDNA,  
139 phylogeographic, and microsatellite analyses suggest this group formed rapidly  
140 (between 100,000 and 27,000 years ago; Zink et al. 2006) and is not strongly  
141 genetically differentiated, despite marked sexual signal and behavioral differentiation  
142 among populations (Dor et al. 2010, Dor et al. 2012). Phenotypic and genomic  
143 divergence among barn swallow populations is likely due to a combination of selection  
144 and drift in the context of variable degrees of geographic isolation. For example,  
145 populations in North America and Eurasia have likely diverged without gene flow,  
146 whereas evidence suggests that populations in the Middle East have experienced  
147 recent historical or ongoing gene flow with populations in Europe (Dor et al. 2010,  
148 2012).

149         Previous research in barn swallows has demonstrated phenotypic divergence in  
150 traits important to both sexual selection (morphological and behavioral sexual signals;  
151 e.g., Safran and McGraw 2004, Safran et al. 2005, Vortman et al. 2011, Vortman et al.  
152 2013, Scordato and Safran 2014) and natural selection (traits related to flight, foraging,  
153 and migration; reviewed in Turner 2006). One of six well-characterized sub-species  
154 within the larger *H. rustica* complex, the European barn swallow (*H.r. rustica*), has been  
155 the subject of intense research activity over the last twenty years, with an emphasis on  
156 sexual selection (Turner 2006). This work clearly demonstrates that females mate  
157 preferentially with long-tailed males, and their offspring experience advantages over  
158 those of shorter-tailed males (e.g., Møller 1994). By contrast, in two North American

159 populations of *H.r. erythrogaster*, males with darker plumage color have greater social  
160 and extra-pair mating success (Safran et al. 2005, Safran et al. *in revision*). In the east-  
161 Mediterranean distributed *H.r. transitiva*, males with a combination of darker ventral  
162 plumage and longer streamer lengths are favored by females through social and extra-  
163 pair mating decisions (Vortman et al. 2011, Vortman et al. 2013).

164 Both experimental and observational data suggest that tail streamer length and  
165 ventral color are under varying degrees of sexual selection in different populations of  
166 barn swallows, and previous work has defined characteristics of wing shape, such as  
167 wing length, as being associated with migratory behavior (Von Ronn et al 2016). We  
168 therefore use these traits to examine the roles of sexual and natural selection in  
169 contributing to genome-wide divergence among populations of barn swallows. Further,  
170 previous work has shown that variation in ventral color (Saino et al 2013, Hubbard et al.  
171 2015) and tail streamer length have heritable components (Møller 1994), indicating that  
172 they are subject to evolutionary change and thus relevant aspects of phenotype to  
173 investigate in terms of their influence on genome-wide divergence.

174 Here, we genotyped thousands of SNPs (using genotyping-by-sequencing, GBS)  
175 in 354 barn swallows from eight populations, representing four of six subspecies,  
176 distributed across the Northern Hemisphere. Additionally, we report a draft whole  
177 genome sequence for a male *Hirundo rustica erythrogaster*, which was constructed to  
178 ensure that GBS sequences cleanly assembled to a reference and to reduce problems  
179 with duplicates/paralogs. We use these population genomic data to address the extent  
180 to which population genomic divergence (based on average genome-wide  $F_{ST}$ ) is  
181 associated with divergence in (i) sexual signals, (ii) wing length, (iii) features of climate,

182 and/or (iv) geographic distance. The first three factors all relate to IBA, while the fourth  
183 relates to IBD. A specific objective of this study is to partition variance in pairwise  
184 genomic divergence as a function of geographic distance (IBD); population differences  
185 in ecology, i.e. climate, (IBA, influenced by natural selection); wing morphology related  
186 to flight and migratory behavior (IBA, influenced by natural selection); and differences in  
187 morphology related to sexual communication (IBA, influenced by sexual selection).

188

## 189 **METHODS**

### 190 **Field Data Collection**

191 We sampled individuals from eight locations representing four sub-species across the  
192 barn swallow breeding range (Table 1). The following samples or measures were taken  
193 from male barn swallows known to be breeding at each site: 1) a sample of ventral  
194 feathers for objective color quantification, 2) length of wing and tail streamers (outer  
195 rectrices; see Table 2 for a complete description), and 3) blood samples as a DNA  
196 source from each bird (approx 50  $\mu$ l, stored in 2% SDS lysis buffer). See Figure 1 and  
197 Table 1 for sampling location and final sample size information.

198

### 199 **Genomics Methods**

200 **Genome Assembly.** A draft genome was assembled, to provide a resource for future  
201 work and to ensure that regions used for GBS cleanly assembled to a reference. A draft  
202 genome assembled with moderate to high coverage enables the identification of single-  
203 copy portions of the genome, reduces the challenges associated with distinguishing  
204 close paralogs from alleles at the same locus, and (depending on quality and scaffold

205 sizes) can provide context of the genomic location of GBS loci. The DNA for our  
206 genome assembly came from a male barn swallow with a well-known reproductive  
207 history in our Boulder, Colorado study site (ID 2540-44680). This male was first  
208 captured in 2008 and had five successive breeding seasons at the same location (from  
209 2008 through 2012). To generate a draft reference genome we obtained sequences  
210 from four lanes of Illumina HiSeq platform, two lanes of 101 base-pair reads from two  
211 paired-end libraries and two lanes of 101 base reads from a mate-pair library from  
212 MacroGen ([www.macrogenusa.com](http://www.macrogenusa.com)). One paired-end library had an insert size of 176  
213 bp, while a second paired-end library had an insert size of 454 bp. The mate-pair  
214 average insert size was 1,458 bp. We obtained a total of 27 Gbp from paired-end library  
215 one, and 17.4 Gbp from library two. From the two lanes of mate pair sequence we  
216 obtained 43.4 Gbp. After cleaning to remove low quality reads (lower than a base  
217 quality of 20 on either end of the read) and common contaminants, 129.4 million pairs of  
218 reads remained in library one, 80.0 million pairs in library two, and 95.8 million pairs in  
219 the mate pair library remained, for a total of 61.7 Gb of sequence. Based on an  
220 estimated genome size of 1.3 Gb (Andrews et al. 2009) this is an average of 47x  
221 coverage.

222 Reads were assembled using SOAPdenovo 2.04, using a K-mer length of 47, an  
223 edge coverage cutoff of 3, a k-mer frequency cutoff of 3, and an arcweight filter of 3.  
224 Repeats were resolved with reads. Otherwise, parameters were as defaults. After  
225 removing short scaffolds (below 1000 bp) we had a total of 100,153 scaffolds, and a  
226 total assembly length of 1.1Gb, 85% of the estimated 1.3 Gb genome size (Andrews et  
227 al. 2009). The average scaffold length was 11,010 bp, the longest scaffold was 732,517

228 bp, the N50 38,844bp, and N90 was 3,718. Of this 1.1 Gb of assembled sequence, a  
229 total of 1.06 Gb could be conservatively mapped to the *Ficedula albicollis* genome  
230 (Ellegren et al. 2012), using blastn with a minimum e-value cutoff of  $10^{-80}$  and a  
231 minimum sequence similarity of 80%.

232         This alignment of our *Hirundo rustica* genome assembly to the most closely-  
233 related, well-annotated genome assembly for Collared Flycatcher *Ficedula albicollis*  
234 (Ellegren 2012), shows a high degree of sequence conservation. Over 91% of our  
235 single-copy, assembled sequence could be uniquely placed onto the genome. Thus,  
236 although we do not have a genetic or physical map placing our sequences onto  
237 chromosomes, we can provisionally place virtually all of our assembled genome using  
238 sequence similarity to existing avian genome sequences. Given the high degree of  
239 synteny found across passerines (Backström et al 2008, Ellegren 2013, Kawakami  
240 2014), and among much more distant avian lineages (Derjushcheva et al 2004, Zhang  
241 2014), taking advantage of existing related genomes enabled us to infer which markers  
242 were likely to be linked on chromosomes. We also identified markers likely to be on  
243 autosomes and on the sex-determining Z chromosome. As the barn swallow  
244 populations we focus on in this paper exhibit only slight genome wide differentiation  
245 (see Results), this draft genome is ideal for use as a reference for aligning the GBS  
246 SNPs described below.

247 **Population Genomics.** We generated DNA sequence data from 354 individual barn  
248 swallows sampled across eight populations (see Table 1). We constructed reduced  
249 genomic complexity libraries for each individual using a restriction fragment-based  
250 procedure (Parchman et al., 2012; Gompert et al., 2012). We first digested genomic

251 DNA with two restriction endonucleases (EcoRI and MseI) and ligated double-stranded  
252 adaptor oligonucleotides to the digested fragments. These oligonucleotides consisted of  
253 the priming sites for Illumina sequencing, followed by eight, nine, or ten base pair  
254 barcode sequences that allow for the unique identification of sequences from each  
255 individual. This method has been successfully used to generate population genetic data  
256 for a large number of projects (Gompert et al. 2012, 2014b, Parchman et al. 2012, 2013,  
257 Nosil et al. 2012b, Mandeville et al. 2015); a full version of the protocol is available at  
258 dryad (doi:XXX). We used 354 unique barcoded adaptors, which allowed us to  
259 sequence all individuals in one Illumina HiSeq sequencing lane. After the restriction and  
260 ligation reactions, we pooled all samples and used standard Illumina PCR primers to  
261 amplify the barcode-adapted fragments. We separated the amplification products on a  
262 2% agarose gel and excised fragments between approximately 350 and 450 bp in  
263 length. We purified these fragments using Qiagen's Qiaquick Gel Extraction Kit (Qiagen  
264 Inc.). Concentration and quality of the pooled library was evaluated on an Agilent  
265 BioAnalyzer qPCR. Sequencing was performed by the National Center for Genome  
266 Research (Santa Fe, NM, USA) with 100 base pair single end reads, and yielding  
267 98,401,301 reads following standard contaminant filtering.

268 We used a custom Perl script (dryad doi:XXX) to remove bases associated with  
269 the barcode and EcoRI cut site for all 98,401,301 sequences and to replace the  
270 sequence IDs with individual IDs for each DNA sample. This script also corrected  
271 barcodes with potential single or double base mismatches due to sequencing or  
272 oligonucleotide synthesis errors. Sequences were then split by individual for further  
273 analysis. Four of the individuals had fewer than 2,000 reads and were disregarded,

274 which left 350 individuals for further analysis. We used *bwa* v 0.7.12 (Burrows-Wheeler  
275 Aligner; Li & Durbin 2009) to assemble data for each individual against the barn swallow  
276 genome assembly, with an edit distance of 4 and the remaining parameters set as  
277 default. This edit distance ensures that, even with 1% sequencing errors typical of  
278 Illumina sequencing, distant alleles from divergent subspecies will map to the genome,  
279 while nevertheless preventing non-specific alignments. Across all individuals an  
280 average of 61.66% of reads assembled, with 97.5% of individuals having 57.9% or  
281 greater percentage of reads assembling. We used *samtools* v 1.2 and *bcftools* v 1.2 (Li  
282 et al., 2009) to identify variant sites in the assembled sequences and obtain genotype  
283 likelihoods for variable sites. To identify variants, we required data for 60% or more  
284 individuals. This resulted in 67,773 single nucleotide variants. A complete list of  
285 parameters used for assembly and variant calling are available from the authors by  
286 request.

287 We used point estimates of allele frequencies in all 350 individuals to separate  
288 this set into 22,328 common variants (minimum minor allele frequency of 5% or greater)  
289 that were used for further analysis and 45,445 rare variants that were disregarded. We  
290 focused on comparatively common variants because they are more likely to occur in  
291 more than one location and contain information about population histories (Gompert et  
292 al. 2012). To avoid analyzing SNPs with highly correlated allele frequencies, we  
293 randomly sampled a single SNP from each of the 100 bp regions in the assemblies  
294 (GBS loci are 100 bases in length). For the resulting final set of 9493 single nucleotide  
295 polymorphisms (SNPs), the average coverage depth per individual per site was 1.5×  
296 (0.09-3.57, 2.5% and 97.5% quantiles). This kind of low coverage genomic data is

297 appropriate for population-level inferences when analyzed with models that incorporate  
298 uncertainty arising from variability in sequencing coverage across individuals and loci  
299 (e.g., Neilsen et al. 2011, Gompert et al. 2012, Gompert and Buerkle 2014).

300 **Population genetic analyses.** We used a Bayesian model to estimate population allele  
301 frequencies for each of 9,493 variable nucleotides based on the point estimates of  
302 genotypes from *bcftools* (described in Gompert et al., 2012). Genotypes and the  
303 population allele frequencies were treated as unknown model parameters, and  
304 genotype probabilities and allele frequencies were simultaneously estimated for each  
305 sampling locality (separately from estimates for other localities). Importantly, this model  
306 incorporates uncertainty from stochastic variation in sequencing coverage depth across  
307 individuals and loci into the estimation process, and provides a sound approach for  
308 estimating population level parameters for low coverage sequencing approaches.  
309 Population allele frequencies serve as prior information, and genotype probabilities are  
310 inferred for each locus in each individual. We obtained posterior probabilities for  
311 parameters using Markov chain Monte Carlo (MCMC). Each analysis consisted of a  
312 single chain iterated for 2,000 steps following a 1,000 step burn-in, with samples  
313 retained every other step, yielding 1,000 samples from the posterior distributions. For  
314 these data and this simple model, mixing and convergence were clearly evident in plots  
315 of MCMC histories.

316 We obtained the mean genotype (scale of 0 to 2) from the posterior distributions  
317 for each SNP in each individual. Furthermore, we calculated the mean genotype across  
318 all individuals to center the genotypes at each locus before calculating the genetic  
319 covariance between the genotype vectors for all pairs of individuals. We summarized

320 the genetic covariance matrix using principal component analysis (PCA). Because the  
321 PCA will reflect all genetic covariances and will be affected by uneven sampling (e.g.,  
322 144 samples from Colorado and 16 from Romania), we performed a second PCA in  
323 which we randomly down-sampled all localities to the smallest sample size from any  
324 locality (N=16). We also obtained mean allele frequencies from the posterior  
325 distributions and transformed these to estimates of mean  $F_{ST}$  (Hudson 1992) and Nei's  
326 D (Nei et al. 1983; Takezaki & Nei 1996) between all pairs of populations.  $F_{ST}$   
327 and Nei's D were highly correlated ( $r=0.996$ ), so we only present results for  $F_{ST}$ . We  
328 performed the PCA and calculated  $F_{ST}$  and Nei's D in R (using *prcomp* and custom  
329 functions, R Core Team, 2015).

330

### 331 **Phenotypic and Environmental Variables**

332 Our goal was to compare divergence among traits known to be related to sexual  
333 signaling (plumage color and streamer lengths) and migratory behavior (i.e., wing  
334 length). We also compared aspects of environmental variability known to affect the  
335 aerial insect populations that barn swallows prey upon. These variables include  
336 elevation and several measures of temperature variation (e.g., Thomsen et al 2016).

337

338 **Quantifying Feather Color.** We collected 5-10 feathers from the throat and breast  
339 (upper ventral region) and vent (lower ventral region, below the attachment of tail  
340 streamers) and stored them in small envelopes in a dark, dry environment prior to  
341 measurement (*following* Safran and McGraw 2004). We assessed the color of these  
342 samples by measuring plumage brightness using an Ocean Optics USB4000

343 spectrometer (Dunedin, FL). Reflectance data were generated relative to a white  
344 standard (Ocean Optics WS-1) and a dark standard (all light excluded), and spectra  
345 were recorded with the SpectraSuite software package (version 2.0.125, Ocean Optics  
346 Inc.). We find no evidence of UV reflectance in the ventral plumage color of barn  
347 swallows (Safran and McGraw 2004) and thus used three traditional axes of color for  
348 objective measurement of color variation. We used average brightness, which was  
349 calculated from three separate measurements of the collected throat and breast  
350 feathers, as a representative metric of overall ventral plumage color. Average  
351 brightness is a good color metric, as all three traditional axes of color (hue, chroma, and  
352 brightness) were previously found to be highly correlated across the ventral region of  
353 individual barn swallows (McGraw et al. 2005, Safran and McGraw 2004, Hubbard  
354 unpublished data), and brightness is the most variable dimension of color in this region  
355 (Hubbard unpublished data). Lower brightness scores (% reflectance) indicate plumage  
356 color that appears darker, redder, and more saturated when compared to feathers with  
357 higher brightness scores.

358

359 ***Estimates of Phenotypic Divergence.*** To determine pair-wise distance in phenotypes  
360 among environmental and morphological traits, we used an unbiased effect size statistic  
361 ( $\Delta P$ ; Safran et al. 2012) to calculate trait distance for each trait in pair-wise comparisons  
362 among the 8 sampled populations.  $\Delta P$  is calculated based upon a joint cumulative  
363 distribution function (CDF) from all populations in the data set. Distances were  
364 calculated for each pairwise comparison using the population median percentile in the  
365 overall CDF.  $\Delta P$  was developed specifically to analyze phenotypic distance among

366 closely related populations, as it easily accommodates simultaneous comparisons of  
367 any number of traits across any number of populations, and is relatively insensitive to  
368 unequal variances and sample sizes among populations (Safran et al. 2012). For all  
369 analyses, we use the absolute value of pairwise distances.

370

371 ***Climate and Elevation Data.*** Temperature and elevation data were obtained using the  
372 CRUTEM database maintained by the Climatic Research Unit and available at  
373 <http://www.cru.uea.ac.uk/cru/data/crutem/ge/>. Using this database, we downloaded the  
374 last 50 years of temperature data from the 3 weather stations closest to each of our  
375 sampling sites (see supplementary materials). From these data, we derived the mean,  
376 minimum, and coefficient of variation in temperature during the breeding season.  
377 Breeding seasons vary among our populations, and, accordingly, we used the following  
378 months for climate data collection for each site: Colorado, USA: April-September; New  
379 York, USA: May-August; UK: April-August; Israel, January-April; and Czech Republic,  
380 Romania, Turkey and Taiwan, April- July. These measurements have been routinely  
381 employed in studies of other avian taxa with widespread geographic ranges  
382 (Rubenstein and Lovette 2007, Rubenstein 2007, Botero et al. 2009, Botero et al.  
383 2014).

384

385 ***Estimates of Geographic Distance.*** Geographic distances between study sites were  
386 calculated in the R package geosphere using the Haversine great circle distance  
387 between points. This is the shortest “as the crow flies” distance, assuming a spherical  
388 earth.

389 **Associations of Geographic Distance, Environmental Context and Phenotype with**  
390 **Population Genomic Divergence.**

391 To assess the degree to which genome-wide divergence is associated with geographic  
392 and phenotypic distance, we analyzed correlations between pairwise trait distance and  
393 both pairwise geographic and genetic distances ( $F_{ST}$ ), and assessed their significance  
394 (999 permutations) using Mantel tests using the R package 'vegan'. Next, to quantify  
395 the relative strength of association among geographic, phenotypic, and genomic  
396 variables, accounting for correlations between phenotypic and geographic distance, we  
397 used two complementary statistical approaches. First, we applied multiple matrix  
398 regression (Wang 2013). This performs multiple regression on distance matrices and  
399 uses permutation tests ( $n = 10,000$ ) to obtain  $p$  and  $R^2$  values using the MRM function  
400 in the R package 'ecodist' (Wang 2013). In the second approach, we employed  
401 constrained redundancy analysis and variance partitioning to analyze the relative  
402 contributions of traits related to natural and sexual selection and environmental context  
403 in explaining pairwise genetic divergence (Legendre and Fortin 2010) using the R  
404 package 'vegan' (Oksanen et al 2015 v. 2.2). Redundancy analysis is a type of  
405 constrained ordination that quantifies how much variation in a set of variables is  
406 explained by a second set of variables, with the option of conditioning on a third set.  
407 This analysis is ideal for correlated matrices, as is often the case with the matrices  
408 related to phenotype, genotype, environmental variables and geographic distance  
409 among closely related populations (Shafer and Wolf 2013, Wang and Bradford 2014).  
410 Using these methods, variance can then be partitioned between the constrained,  
411 conditioned, and joint variable sets. Our data were structured in two different formats,

412 depending on the analysis. For Mantel tests, we employed matrices of pairwise  
413 differences for each population. For MRM and the variance partitioning analyses, these  
414 matrices were transformed into vectors of pairwise differences for each variable and  
415 each pair of populations. When the response variable is a single vector (as here, using  
416 pairwise differences in mean  $F_{ST}$  between populations), variance partitioning is done by  
417 partial regression.

418

## 419 **Results**

### 420 **Population Genomics**

421 Principal component analysis of genetic covariances between an even sample of  
422 individuals from each population ( $n = 16$ ) revealed clear genomic differences between  
423 some localities, but greater genomic similarity among the nearby sampling locations in  
424 Czech Republic, Romania, Turkey, and Israel (Figure 2). The first two principal  
425 components explained 52% of the genomic variation among individuals in each locality  
426 (Figure 2). The separation of populations on PC1 is consistent with the phylogenetic  
427 hypothesis (Dor et al. 2010) that separates eastern and western barn swallows, with  
428 samples from Israel and Europe recoverable as a distinct lineage from samples in North  
429 America (Colorado and New York) and southern Asia (Taiwan). Interestingly, while  
430 Israel is considered a distinct subspecies based on phenotype, it cannot be  
431 differentiated from European populations based on genomic covariance. PC2 explains a  
432 relatively small amount (10.3%) of genetic variation, compared to PC1 (41.9%), and  
433 further separates subspecies within the eastern and western clades. The PCA based on  
434 all 350 individuals gave considerable weight in the first axis to distinction between the

435 large sample from North America (N=144 from Colorado and 27 from New York) and  
436 samples from elsewhere, with PC1 accounting for 67.9% of the genetic variation.  
437 Pairwise  $F_{ST}$  ranged from 0.024 to 0.073 (Table 3).

438

### 439 **Phenotypic and Environmental Variables**

440 ***Phenotypic and environmental divergence.*** Populations differed both in the extent of  
441 phenotypic variation and environmental context (Tables 4 and 5). Mean percentiles  
442 within a cumulative frequency distribution varied considerably among populations; the  
443 distribution of variables measured from each population differed in terms of its  
444 placement on the overall cumulative frequency distribution (Table 4). For example,  
445 populations in the UK have the longest wing length, whereas populations in Taiwan  
446 have the shortest. Populations in North America (Colorado and Ithaca) have the lightest  
447 throat color yet the darkest breast color; thus, there is variation in the extent and  
448 direction in which color patches differ among sub-species. The spread of percentile  
449 values is indicative of the degree of pairwise divergence for each trait or climate  
450 variable. For example, of all phenotypic traits, populations exhibited the most extreme  
451 differences in wing length, with a range of percentiles from 9.02–91.16. Note that the  
452 percentile values for the phenotypic measures are based on multiple individuals within  
453 populations (Table 1), whereas the percentile values for the environmental measures  
454 are based on one measure within each population. Analyses of variance reveal that all  
455 of the phenotypic traits we measured in this study showed statistically significant  
456 differences among populations (adjusted R-squared values range from 0.25 to 0.52;  
457 Table 5).

458 **Associations of Geographic Distance, Environmental Context and Phenotype with**  
459 **Population Genomic Divergence.**

460 **Geographic Distance.** Geographic distance between pairs of populations predicted  
461 genome-wide divergence, consistent with a model of Isolation by Distance ( $r = 0.628$ ,  
462 Mantel  $p$  value  $< 0.008$ , Figure 3, Table 3).

463 **Phenotypic Divergence and Environmental Context.** Distance matrices (based on  
464  $\Delta P$ ) for various features of phenotype, including measures of ventral color, tail and wing  
465 length, were positively correlated with genome-wide divergence among pairs of eight  
466 geographically isolated populations (Mantel tests; Figure 4, Table 6). None of the  
467 pairwise differences in environmental variables (Mantel tests; elevation and various  
468 measures of breeding season temperature) were significantly associated with genome-  
469 wide divergence among our study populations (Figure 5, Table 6).

470 **Geographic Distance, Environment, Phenotype.** Several features of phenotypic  
471 divergence also co-varied with geographic distance (Mantel tests; Figure 6, Table 6),  
472 whereas environmental and climate features did not (Mantel tests; Figure 7, Table 6).  
473 Thus, genome-wide divergence among closely related populations was associated with  
474 both geographic and phenotypic trait distance, and the two are sometimes correlated  
475 with each other. It is therefore necessary to adequately control for correlations between  
476 phenotypic and geographic distance in order to infer the relative significance of IBA and  
477 IBD. We did this using two complementary approaches.

478 First, to investigate the associations of specific traits with pairwise, genome-wide  
479  $F_{ST}$ , while accounting for correlations among variables in our model (e.g., phenotypes  
480 correlated with one another and with geographic distance), we used multiple matrix

481 regression (Wang 2013). We started with a maximal model that included the pairwise  
482 distance matrix of  $F_{ST}$  values as the response variable and distance matrices (based on  
483  $\Delta P$  values) for all nine phenotypic and ecological variables as predictors. We then used  
484 backwards-stepwise model selection, sequentially deleting the least significant term and  
485 rerunning the model, until coefficients for all remaining predictor variables were  
486 significantly different from zero. The final multiple matrix regression model included two  
487 aspects of phenotype that explained significant variation in pairwise  $F_{ST}$  while controlling  
488 for spatial autocorrelation: wing length and throat color (full model  $F_{2,17} = 24.95$ ,  $P =$   
489  $0.002$ ,  $r\text{-squared} = 0.757$ ; parameter estimate and t-test: wing length coefficient = 0.82,  
490  $T=6.12$ ,  $P = 0.002$ , throat color coefficient = 0.22,  $T = 1.79$ ,  $P = 0.03$ , geographic  
491 distance coefficient = 0.01,  $T = 0.09$ ,  $P = 0.94$ ).

492 We used the significant predictor variables from the multiple matrix regression in  
493 variance partitioning and redundancy analyses to further examine the relative  
494 contributions of wing length, throat color, and geographic distance in explaining  
495 variation in genome-wide divergence among the populations in our data set. The  
496 variance partitioning approach enabled us to test hypotheses about the relative  
497 significance of IBA and IBD, taking into account correlations among phenotypic,  
498 environmental, and geographic distance variables.

499 The best model fit from the variance partitioning analyses (Table 7) included the  
500 effects of all three matrices (geographic distance, throat color distance, and wing length  
501 distance). These models, in which various combinations of matrices are conditioned  
502 upon one another, demonstrated that collectively 73% of genome-wide divergence is  
503 attributable to three variables: wing length, throat coloration, and geographic distance

504 between populations (Figure 8). Further analyses enabled us to analyze the association  
505 between each matrix (geographic distance, throat color, wing length) separately by  
506 conditioning each variable on the others in all possible combinations (Table 7). For  
507 example, when the matrix containing pairwise distance in wing length is conditioned on  
508 the matrices containing pairwise differences in geographic distance and throat color, the  
509 influence of wing length on its own accounted for 42% of pairwise genomic distance  
510 among the populations in our sample (Figure 8, Table 7). When the matrix containing  
511 pairwise distance in throat color is conditioned upon the matrices containing pairwise  
512 differences in geographic distance and wing length, the influence of throat color on its  
513 own explained 5% of pairwise genomic distance among the populations in our sample.  
514 Finally, when conditioned upon the wing and color matrices, the effect of geographic  
515 distance on its own did not explain additional variation in genomic distance among the  
516 populations in our sample (Figure 8, Table 7). Interestingly, the only place where the  
517 geographic distance matrix explained a significant amount of variation in genomic  
518 divergence is when both geographic distance and wing length were considered side by  
519 side and conditioned upon their correlation with the color matrix (Figure 8, Table 7).

520

## 521 **Discussion**

522 Geographic distance and phenotypic distance are strongly correlated among our study  
523 populations. Thus, we applied two complementary methods of variance partitioning—  
524 multiple matrix regression, and constrained redundancy analysis—which enabled us to  
525 analyze the relative contributions of correlated matrices (geographic distance and  
526 phenotypic distance) to genome-wide divergence. Overall, our results demonstrate clear

527 evidence that both IBA and IBD contribute to genome-wide divergence among these  
528 closely related populations of barn swallows. When spatial autocorrelation between  
529 phenotype and geographic distance are accounted for, our results suggest that  
530 divergence in an ecological trait (wing length) and a sexual signaling trait (throat color)  
531 play a larger role in population genetic divergence than does geographic distance.

532 Surprisingly, we found no evidence that elevation and temperature differences  
533 were influential to genome-wide divergence. Although these features of the  
534 environment, which are relevant for obligate aerial insectivores, did vary spatially, they  
535 were neither associated with geographic distance nor genome-wide distance among the  
536 populations in our sample. In a further investigation, we also found no evidence that  
537 maximal temperatures or precipitation patterns at each location differed as a function of  
538 geographic distance or were associated with genome-wide divergence (Safran,  
539 unpublished data). Given the span of our sampling locations, ranging from Israel to  
540 North America, these results either suggest that barn swallows occupy fairly similar  
541 environments with respect to elevation and temperature or are not particularly sensitive  
542 to these ecological variables. The latter explanation seems most likely as these  
543 populations are very cosmopolitan in distribution both during the breeding and non-  
544 breeding season and during long migratory trips where they likely inhabit a wide range  
545 of environments.

546 ***Population Genomic Structure and Phenotypic Divergence.*** Our data demonstrate  
547 genetic similarity of individuals within sampling localities, with greater differences  
548 between populations that are separated by large geographic distances and genome-  
549 wide clustering that generally corresponds to named subspecies. Principal component

550 analysis shows that the most genetically different populations along PC1 correspond to  
551 a highly supported east – west split in the current phylogenetic hypothesis for barn  
552 swallows (Zink et al. 2006, Dor et al. 2010). Populations in Asia (*H.r. gutturalis*) and  
553 North America (*H.r. erythrogaster*) are more closely related to one another than either  
554 are to populations in Europe (*H.r. rustica*) and the Middle East (*H.r. transitiva*). PC2  
555 further separates *H.r. gutturalis* (TW) from two samples from the North American *H.r.*  
556 *erythrogaster* populations (IA and CO), and the UK population from other mainland  
557 populations of *H.r. rustica* (TR, CR and RM). Samples from the Israeli subspecies (IL)  
558 are clustered closely with other mainland European populations of barn swallows,  
559 consistent with an unresolved relationship between this subspecies and *H.r. rustica* in  
560 the current phylogeny (Dor et al. 2010). Despite being relatively genetically similar,  
561 individuals from Israel and the continental European populations are fairly divergent in  
562 phenotype, particularly with respect to ventral color.

563         Despite shallow genomic divergence, phenotypic differentiation is apparent in all  
564 aspects of morphology we analyzed in this study, ranging from traits related to body  
565 size and flight (wing and tail length) to ventral color. Phenotypic variation despite  
566 shallow genomic divergence appears common among many taxa, including cichlid  
567 fishes (e.g., Wagner et al. 2012) and particularly in birds (e.g. Parchman et al. 2006,  
568 Rodrigues et al. 2014, Poelstra et al. 2014, Mason and Taylor 2015), where it is often  
569 the case that a few genes are implicated in morphological variation against a fairly  
570 homogenous genomic background (e.g., Poelstra et al. 2014, Kardos et al. 2015). In  
571 other words, although a larger number of genes may be involved in generating plumage  
572 coloration, studies to date have suggested that a large proportion of segregating trait

573 variation is due to variation in a small number of genetic loci. Collectively, these studies  
574 suggest an important role of divergent selection on signaling traits in population  
575 differentiation (Wagner et al. 2012, Poelstra et al. 2014), which might play a particularly  
576 important role during the earliest stages of speciation (Kraaijeveld et al. 2010). Finer-  
577 scale genomic analyses in a broader geographic context will enable us to test whether  
578 phenotypic differentiation is a barrier to gene flow when different subspecies of barn  
579 swallows are in secondary contact.

580 ***Isolation by Adaptation, Controlling for Isolation by Distance.*** Because trait  
581 divergence is correlated with geographic distance among populations, we applied two  
582 complementary statistical methods to tease apart the relative significance of geographic  
583 and phenotypic distance in explaining genome-wide divergence. Both sets of analyses  
584 reveal that wing length and, to a lesser extent, throat color, are most strongly associated  
585 with genome-wide divergence, when geographic distance among populations is  
586 accounted for statistically (Figure 8). The variance partitioning model enables us to  
587 directly quantify the influence of correlations among geographic distance, wing length,  
588 and throat color and to partition the contribution of each of these variables towards  
589 differences in genome-wide divergence (Figure 8, Table 6). Thus, our study reveals an  
590 important role of phenotypic divergence, and supports a model of IBA for explaining  
591 genomic differentiation of geographically isolated populations, bearing in mind that  
592 phenotypic divergence is also strongly correlated with geographic distance, and a model  
593 of IBD, among populations (Figure 8).

594 IBD is one of the most common patterns in population genetic data (e.g., Jenkins  
595 et al. 2010). Evidence is accumulating to suggest that IBA (which includes models of

596 Isolation by Environment, or IBE) is also well supported in a variety of empirical systems  
597 among closely related populations (e.g., Edelaar et al. 2012, Lasky et al 2012, Lee and  
598 Mitchell-Olds 2012, Shafer and Wolf 2013, Morgans et al. 2014, Wang and Bradburd  
599 2014). A recent meta-analysis further revealed that the effect of geographic distance or  
600 spatial autocorrelation (between phenotypes, environmental variables, and geographic  
601 distance) is critical to control for in statistical tests of IBA but has rarely been done in  
602 previous studies (Shafer and Wolf 2013). Simulation studies reveal that a failure to  
603 account for spatial autocorrelation can lead to biased results (Shafer and Wolf 2013). In  
604 our own study, separate analyses of IBA and IBD each revealed significant effects on  
605 genome-wide divergence. Analyses that explicitly considered the correlations of  
606 phenotypic and geographic distance were critical to separating out the relative  
607 significance of each factor. Variance partitioning (Legendre and Fortin 2010) and  
608 multiple matrix regression (Wang 2013) are excellent methods for dealing with spatial  
609 autocorrelation, analyzing the relative significance of IBA and IBD, and are particularly  
610 important alternatives to partial Mantel tests (Bradburd et al 2013, Wang 2013, Wang et  
611 al. 2013), which are subject to false positive results (Diniz-Filho et al 2013).

612 ***Barn swallows and divergent selection.*** Our results support an important role of IBA  
613 in genome-wide divergence. In particular, differences in wing length, the most divergent  
614 phenotypic trait among populations, explained a significant amount of variation in  
615 among-population genomic divergence; throat color is also associated with genomic  
616 divergence, but to a lesser extent. From these results and our understanding of the  
617 function of these traits, we infer an influence of both natural selection and sexual  
618 selection in genomic divergence among these eight populations of barn swallows.

619 These results are intuitive for several reasons. First, wing length and shape are traits  
620 associated with migratory behavior (Marchetti et al 1995, Lockwood et al 1998). Barn  
621 swallows vary in migratory distance, and there is evidence of migratory divides  
622 throughout their range (e.g., Irwin and Irwin 2005, von Rön­n et al. 2016). All four  
623 representatives of the *H.r. rustica* subspecies (UK, Czech Republic, Romania, and  
624 Turkey) have the longest wing lengths among populations sampled, whereas individuals  
625 from distantly-related Taiwan (*H.r. gutturalis*) have the shortest. Interestingly, the sole  
626 non-migratory population in our study, Israel, has intermediate wing lengths that overlap  
627 in variation with individuals from Turkey and Romania and is genomically  
628 indistinguishable from *H. r. rustica*, based on pairwise  $F_{ST}$ . In other migratory bird  
629 populations, variation in migratory behavior can influence both the evolution of wing  
630 shape and genomic differentiation (e.g., Ruegg 2007, 2008, Delmore et al. 2012, 2014,  
631 2015, Rolshausen et al 2009, von Rön­n et al. 2016). Thus, variation in migratory routes  
632 and behaviors (e.g., timing of arrival to breeding grounds, which may affect mate  
633 selection) is proposed as a potentially influential contributor to population divergence  
634 (Irwin and Irwin 2005, Rolshausen et al 2009). Further study is required to better  
635 understand the role of migratory behavior in divergence among barn swallow  
636 populations, as well as the extent to which these seasonal movement patterns influence  
637 trait evolution.

638 Our analyses of IBA also revealed that population-level differences in color  
639 contribute to genomic differentiation. Plumage color in various ventral regions (throat,  
640 breast, vent) varies both within and among closely related populations of barn swallows  
641 (Scordato and Safran 2014). Both throat and breast coloration are variable among the

642 populations in this study, but in different ways. For example, males in our North  
643 American populations (Colorado, New York) have the darkest breast color, yet the  
644 lightest throat color compared to other populations. Ventral coloration has been the  
645 focus of several correlational and experimental studies in barn swallows. Together,  
646 these studies indicate that melanin-based ventral color varies among populations  
647 (Scordato and Safran 2014), is heritable (Hubbard et al. 2015), and relates to social  
648 mate selection and paternity allocation (Safran and McGraw 2004, Vortman et al 2011).  
649 Thus, patterns of ventral color differentiation are likely under divergent sexual selection,  
650 although this hypothesis requires further testing.

651         Patterns of IBA may be underlain by divergent selection on traits that directly  
652 impede gene flow or through the build up of an association between genome-wide  
653 divergence and trait divergence in allopatric populations. Whereas models of IBA have  
654 typically been applied to scenarios where gene flow is possible between populations,  
655 most of the eight populations in our study are geographically isolated from one another,  
656 with the exception of those within continental Europe (Czech Republic and Romania). It  
657 is possible that gene flow occurs among populations sampled from western and central  
658 Europe and Israel, though a formal analysis is needed to make direct inferences about  
659 recent historical or ongoing gene flow in these regions. Nevertheless, divergence in the  
660 absence of gene flow is an important aspect of avian diversification (e.g., Price 2008)  
661 and thus it is highly relevant to ask questions about associations between phenotypic  
662 and genomic divergence in order to better understand patterns and consequences of  
663 trait evolution in isolated, yet closely related populations. Still, ongoing studies of IBA in  
664 locations where divergent populations come into sympatry will be critical for inferring

665 whether trait divergence that may have evolved in isolation imposes a barrier to gene  
666 flow.  
667

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931 *Data accessibility*

932 Genome assembly: XXX

933 Phenotype and environmental data: Dryad XXX

934 Protocol and models: Dryad XXX

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936 *Authors' contributions:*

937 R.J.S. designed the study, R.J.S., M.R.W, J.K.H, B.R.J., T.A., H.K., Y.V., P.P., S.S., and S.C.,

938 collected data in the field, N.C and E.S.C.S. completed the draft assembly of the genome, B.R.J.

939 conducted most of the lab work, R.J.S., E.S.C.S., S.M.F., T. P. analyzed the data with input from

940 P.N., R.J.S. wrote the manuscript with input from E.S.C.S., P.N, T.P, and N.K. All authors gave

941 final approval for publication.

942

943 **Table 1.** Sampling locations across four subspecies of *Hirundo rustica*. Acronyms are  
 944 given for each location that match with locations on Figures 1 and 2.

945

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<b>Sampling location</b> [abbreviation for location on map]	<b>subspecies</b>	<b>Lat/Long</b>	<b>Final Sample size</b>	<b>Sampling dates: year (sample size)</b>
Boulder, Colorado, USA [CO]	<i>Hirundo rustica erythrogaster</i>	40.17, -105.10	144	2008 (50) 2009 (72) 2010 (22)
Czech Republic [CR]	<i>Hirundo rustica rustica</i>	49.06, 14.76	24	2010
Israel [IL]	<i>Hirundo rustica transitiva</i>	32.93, 35.54	45	2008 (3) 2009 (37) 2010 (5)
Ithaca, New York, USA [IA]	<i>Hirundo rustica erythrogaster</i>	42.44, -77.50	27	2002
Romania [RM]	<i>Hirundo rustica rustica</i>	46.75, 23.83	16	2010
Taiwan [TW]	<i>Hirundo rustica gutturalis</i>	25.09, 121.56	18	2010
Turkey [TR]	<i>Hirundo rustica rustica</i>	36.85, 31.16	50	2010 (50)
United Kingdom [UK]	<i>Hirundo rustica rustica</i>	50.50, -4.65	26	2009
<b>Totals</b>				
8 populations	4 subspecies		350	

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950 **Table 2.** Morphological and environmental traits measured in 350 individuals, across 8  
 951 sampling locations.

Types of Traits	Traits
Traits related to reproductive performance among sub-species of barn swallows	Tail streamer length, ventral color (throat breast and vent, % brightness)
Trait related to flight behavior and migratory distance	Wing length (mm)
Features of the environment that affect food availability	Mean, variability, and seasonal minima in breeding temperatures; elevation

952

953 Table 3. Average pairwise  $F_{ST}$  (upper diagonal) and geographic distance (km; lower  
 954 diagonal).

955

<b>Sampling location</b>	Colorado	Czech Republic	New York	Israel	Romania	Turkey	Taiwan	UK
Colorado	<b>NA</b>	0.049	0.024	0.054	0.053	0.046	0.046	0.059
Czech Republic	853.948	<b>NA</b>	0.059	0.038	0.041	0.030	0.062	0.048
New York	2397.257	6726.714	<b>NA</b>	0.062	0.063	0.055	0.055	0.067
Israel	10959.901	2454.926	9169.197	NA	0.044	0.031	0.065	0.051
Romania	9131.466	700.843	7399.893	1828.448	<b>NA</b>	0.037	0.067	0.053
Turkey	10381.422	1864.741	8578.322	591.36	1255.353	<b>NA</b>	0.059	0.045
Taiwan	11321.578	9083.006	12305.315	8199.932	8582.409	8425.255	<b>NA</b>	0.073
UK	7337.099	1443.954	5361.685	3818.056	2144.723	3230.591	10145.824	<b>NA</b>

956

957 Table 4. Percentiles for male morphological traits across all populations used for  
 958 calculating  $\Delta P$  differences. Lower wing and tail streamer percentiles indicate shorter  
 959 wings and streamers, respectively. Higher color percentiles indicate *lighter* color.  
 960 Greater elevation percentiles are indicative of higher elevations at breeding locations.  
 961 Greater percentiles for variation, minima, and average temperatures are indicative of  
 962 more variable, hotter climates measured over the period of data collection (1965-2015).

Trait	Population Sampled								Range in percentiles)
	Czech								
	Colorado	Republic	Israel	New York	Romania	Taiwan	Turkey	UK	
<i>Wing length</i>	24.08	70.73	61.00	30.22	61.87	9.02	61.00	91.16	9.02-91.16
<i>Streamer length</i>	21.82	72.23	60.22	15.05	79.79	28.66	67.16	60.22	15.05-79.79
<i>Throat Color</i>	77.74	59.42	39.96	80.77	18.35	28.26	47.12	44.62	18.35-80.77
<i>Breast color</i>	21.03	58.63	27.61	21.99	58.72	80.77	55.41	70.39	21.03-80.77
<i>Vent color</i>	18.10	27.01	16.02	20.32	21.45	32.99	25.32	30.14	16.02-30.14
<i>Elevation</i>	100	62.5	75	37.5	50	12.5	87.5	25	12.5-100
<i>Var temp</i>	37.5	75	87.5	62.5	50	25	100	12.5	12.5-100
<i>Min temp</i>	62.5	25	12.5	87.5	50	100	87.5	37.5	12.5-100
<i>Mean temp</i>	62.5	25	12.5	75	50	100	87.5	37.5	12.5-100

963  
 964 Table 5. Results of ANOVAs to show geographic variation in morphological traits among  
 965 eight closely related populations of barn swallows.

Trait	F	P	Adj R-squared
Wing length	54.11 <sub>7,353</sub>	< 0.001	0.52
Streamer length	54.27 <sub>7,352</sub>	< 0.001	0.52
Throat Color	18.84 <sub>7,346</sub>	< 0.001	0.25
Breast color	28.16 <sub>7,345</sub>	< 0.001	0.37
Vent color	24.97 <sub>7,345</sub>	< 0.001	0.34

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970 **Table 6.** Results of Mantel tests. Tests of isolation by adaptation and phenotypic-  
 971 genetic distance correlations using traits related to mate selection (color, tail length) and  
 972 migratory behavior (wing length) and environmental traits including elevation and  
 973 various metrics of temperature during the breeding season. To control for multiple  
 974 testing, we have indicated the confidence intervals around the mantel test coefficient.  
 975

Variable	Association with genome wide $F_{ST}$			Association with geographic distance		
	Mantel $r$	CI	$P$ value	Mantel $r$	CI	$P$ value
Wing length	0.85	0.79-0.91	<b>0.002</b>	0.60	0.34-0.76	<b>0.010</b>
Tail length	0.67	0.60-0.81	<b>0.012</b>	0.59	0.38-0.85	<b>0.013</b>
Throat color	0.29	-0.08-0.66	0.091	0.48	0.27-0.78	<b>0.030</b>
Breast color	0.34	0.16-0.53	<b>0.023</b>	0.46	0.24-0.74	<b>0.031</b>
Vent Color	0.40	0.14-0.59	<b>0.037</b>	0.30	-0.11-0.54	0.094
Elevation	-0.04	-0.27-0.38	0.593	0.17	-0.07-0.35	0.202
Min breeding temp	0.07	-0.18-0.40	0.348	-0.01	-0.28-0.38	0.483
Mean breeding temp	0.20	-0.18-0.64	0.145	0.13	-0.14-0.54	0.274
Var breeding temp	0.07	-0.11-0.29	0.346	-0.05	-0.30-0.27	0.560

976 **Table 7.** Quantitative tests of variance partitioning (Figure 7): RDA analyses of the effects of  
 977 selection and geographic distance on genome-wide divergence. WING = pairwise differences in  
 978 wing length, COLOR = pairwise differences in throat color, DIST = pairwise differences in  
 979 geographic distance. Note that the matrix title before the symbol '|' indicates the variance due  
 980 to that matrix that contributed to genome-wide divergence, conditioned upon the matrices listed  
 981 after the '|' symbol.  
 982

<b>Model</b>	<b>F</b>	<b>P</b>	<b>Adjusted R-squared</b>
WING, COLOR, DIST	WING <sub>1,24</sub> = 78.36	< 0.001	0.73
	COLOR <sub>1,24</sub> = 7.03	0.010	
	DIST <sub>1,24</sub> = 0.03	0.878	
WING COLOR+DIST	WING <sub>1,24</sub> = 44.17	< 0.001	0.42
COLOR WING+DIST	COLOR <sub>1,24</sub> = 6.03	0.018	0.05
DIST WING+COLOR	ECO <sub>1,24</sub> = 0.03	0.878	0.00
WING+COLOR DIST	WING <sub>1,24</sub> = 38.66	< 0.001	0.40
	COLOR <sub>1,24</sub> = 6.03	0.021	
DIST+COLOR WING	DIST <sub>1,24</sub> = 1.04	0.321	0.05
	COLOR <sub>1,24</sub> = 6.03	0.024	
DIST+WING COLOR	DIST <sub>1,24</sub> = 32.15	< 0.001	0.70
	WING <sub>1,24</sub> = 44.17	< 0.001	

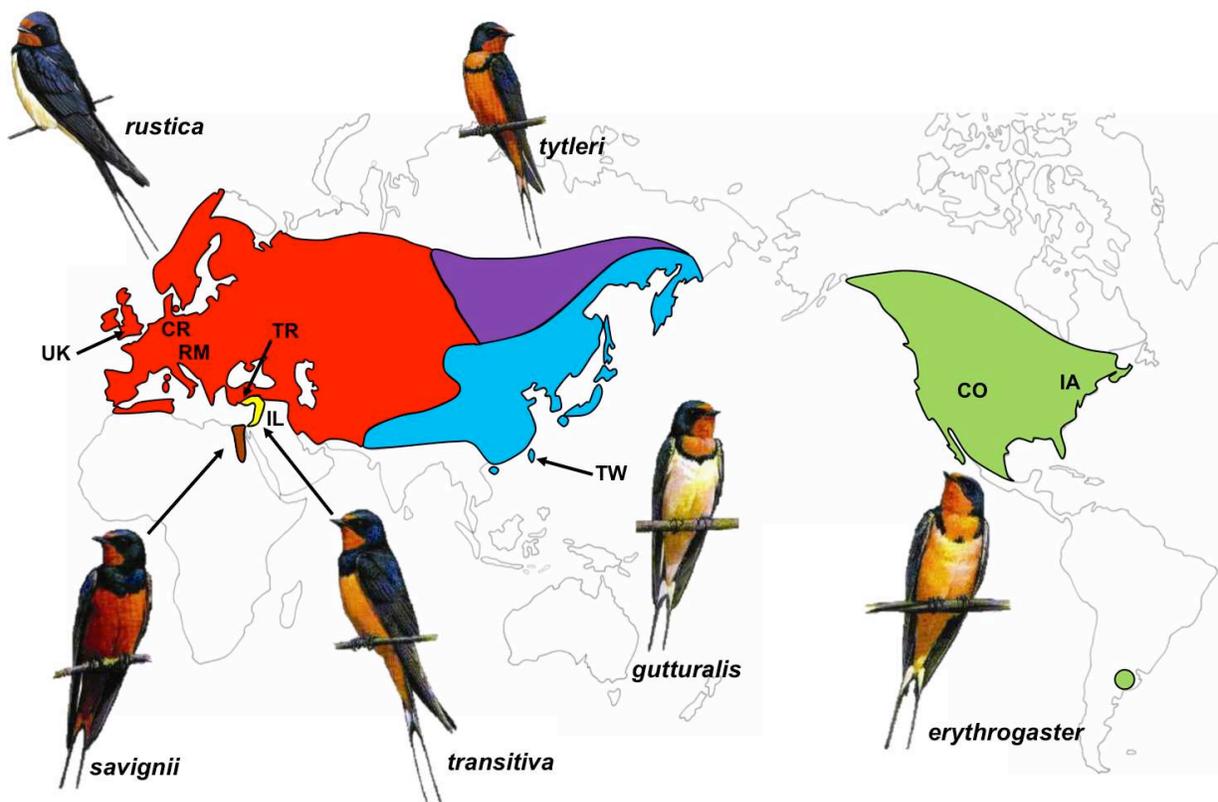
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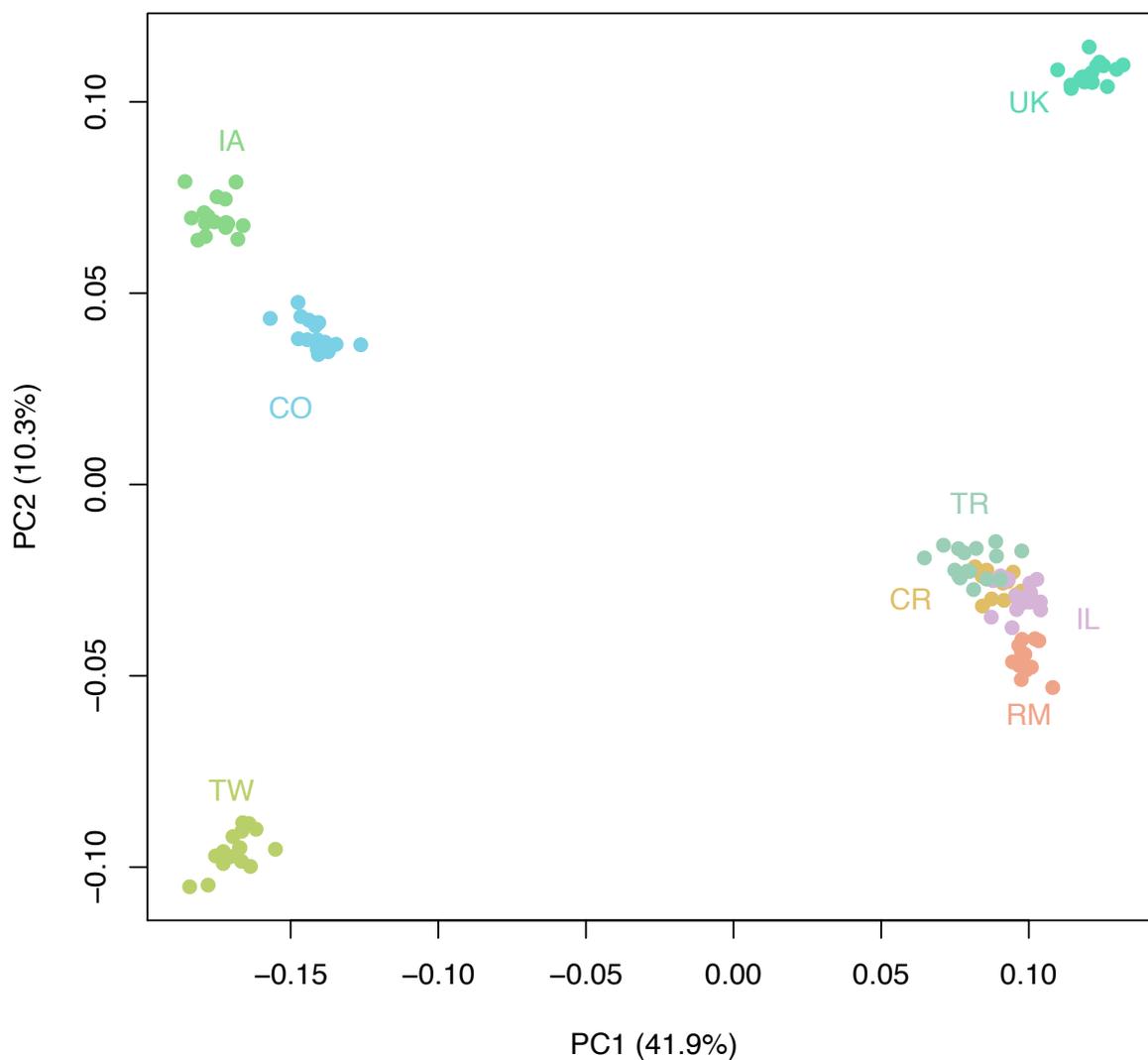
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987 **Figure 1.** *Hirundo rustica* complex breeding distribution map indicating eight sampling  
 988 locations. Cartoons of male phenotypes are shown for each subspecies (with  
 989 permission by artist Hilary Burn). Acronyms for sampling sites are as follows: CO =  
 990 Colorado, USA; CR = Czech Republic; IA = New York, USA; IL = Israel; RM = Romania;  
 991 TR = Turkey; TW = Taiwan; UK = United Kingdom.  
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993

994 **Figure 2.** Statistical summary of population genetic structure based on principal  
995 component axes one (PC1), and two (PC2) derived from genotype probabilities.  
996 Acronyms for sampling sites are as follows: CO = Colorado, USA; CR = Czech  
997 Republic; IA = New York, USA; IL = Israel; RM = Romania; TR = Turkey; TW = Taiwan;  
998 UK = United Kingdom.



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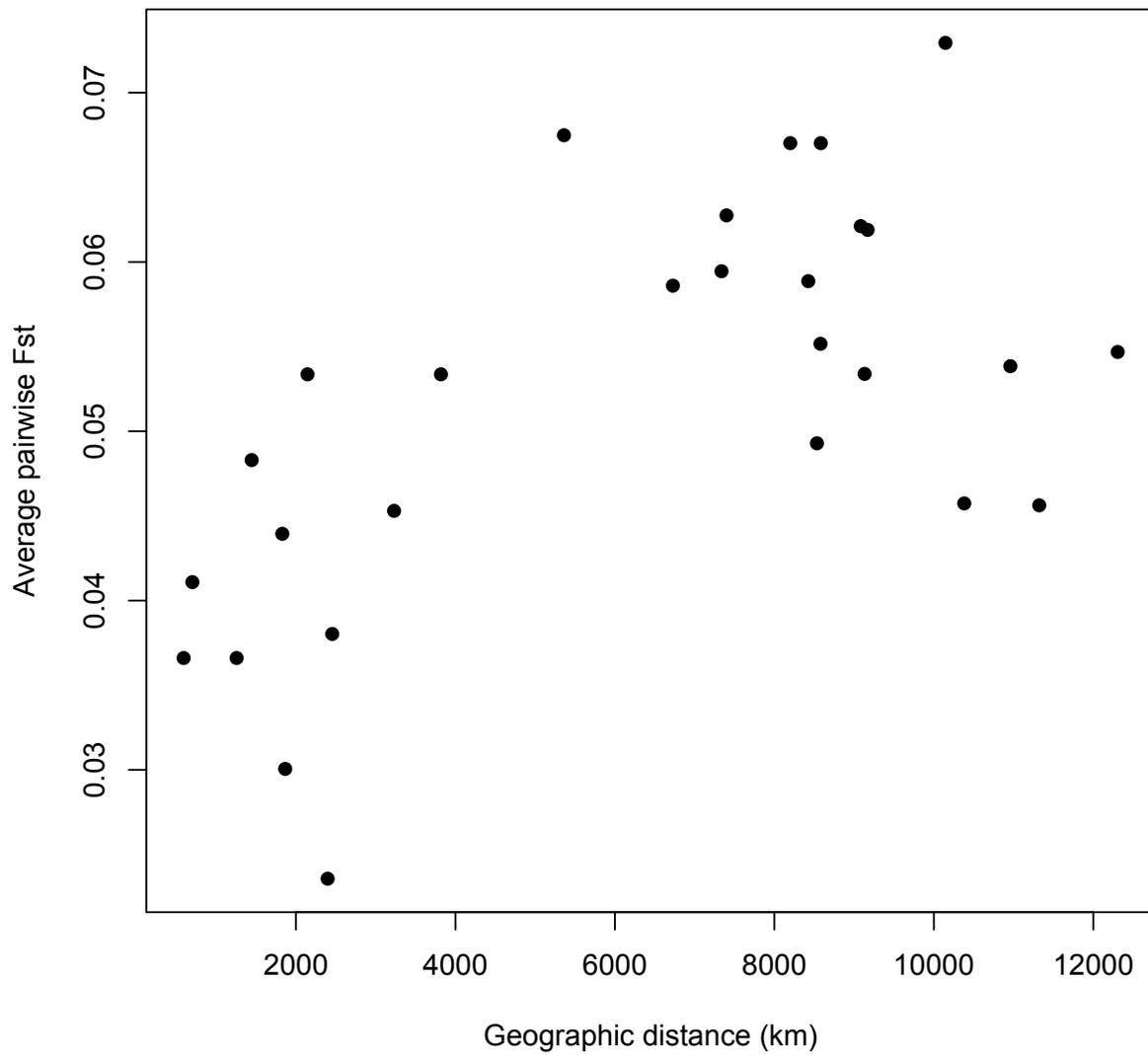
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1003 **Figure 3.** Average pairwise genetic distance ( $F_{ST}$ ) as a function of geographic distance  
1004 (km) among pairwise comparisons of eight barn swallow populations, consistent with a  
1005 pattern of isolation by distance  
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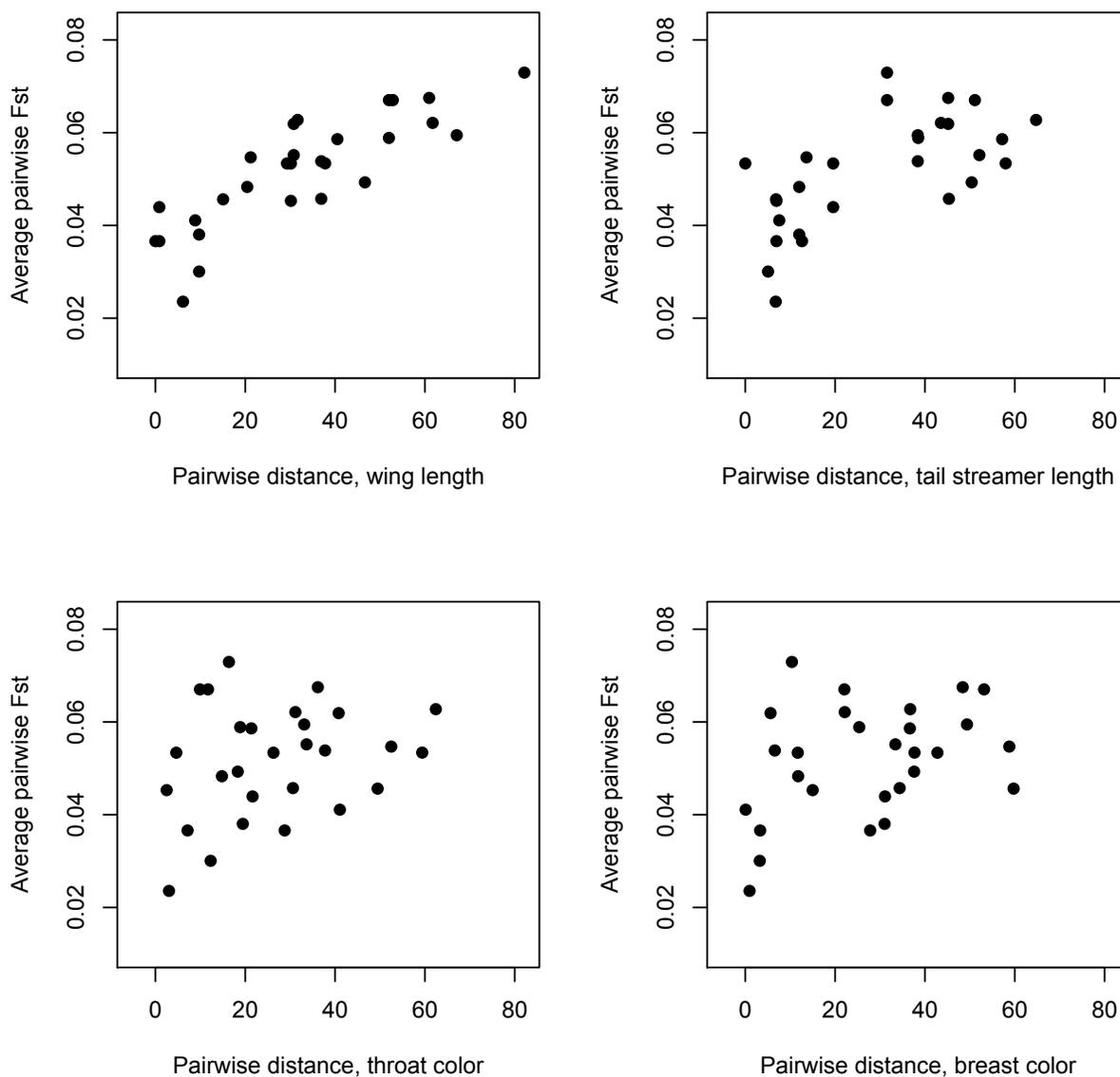
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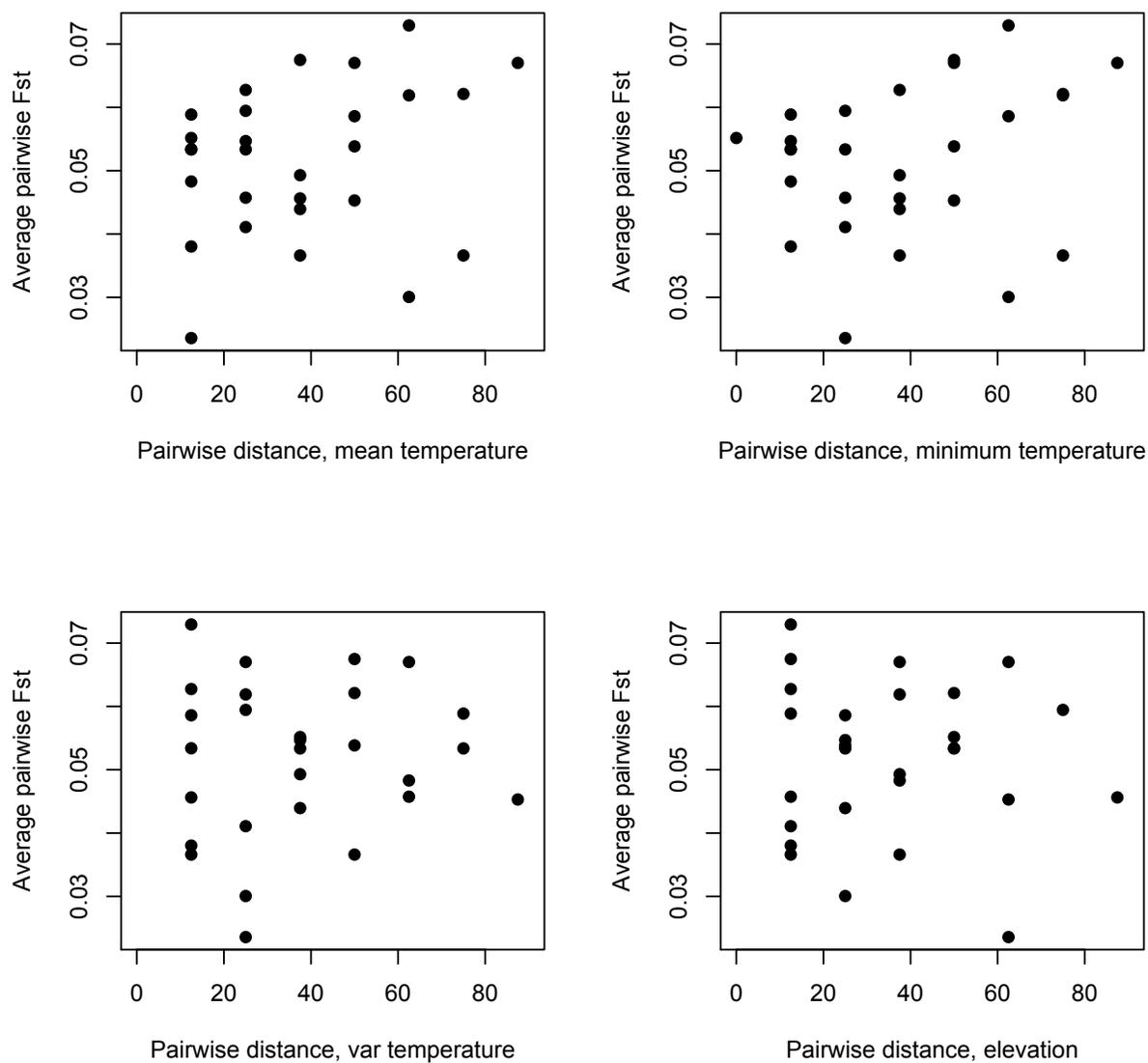
**Figure 4.** Average pairwise genetic distance ( $F_{ST}$ ) is strongly correlated with phenotypic distance among pairwise comparisons of eight barn swallow populations, consistent with a pattern of IBA. Table 6 shows results of mantel tests between  $F_{ST}$  and these four phenotype measures; all are significantly associated with the exception of throat color.



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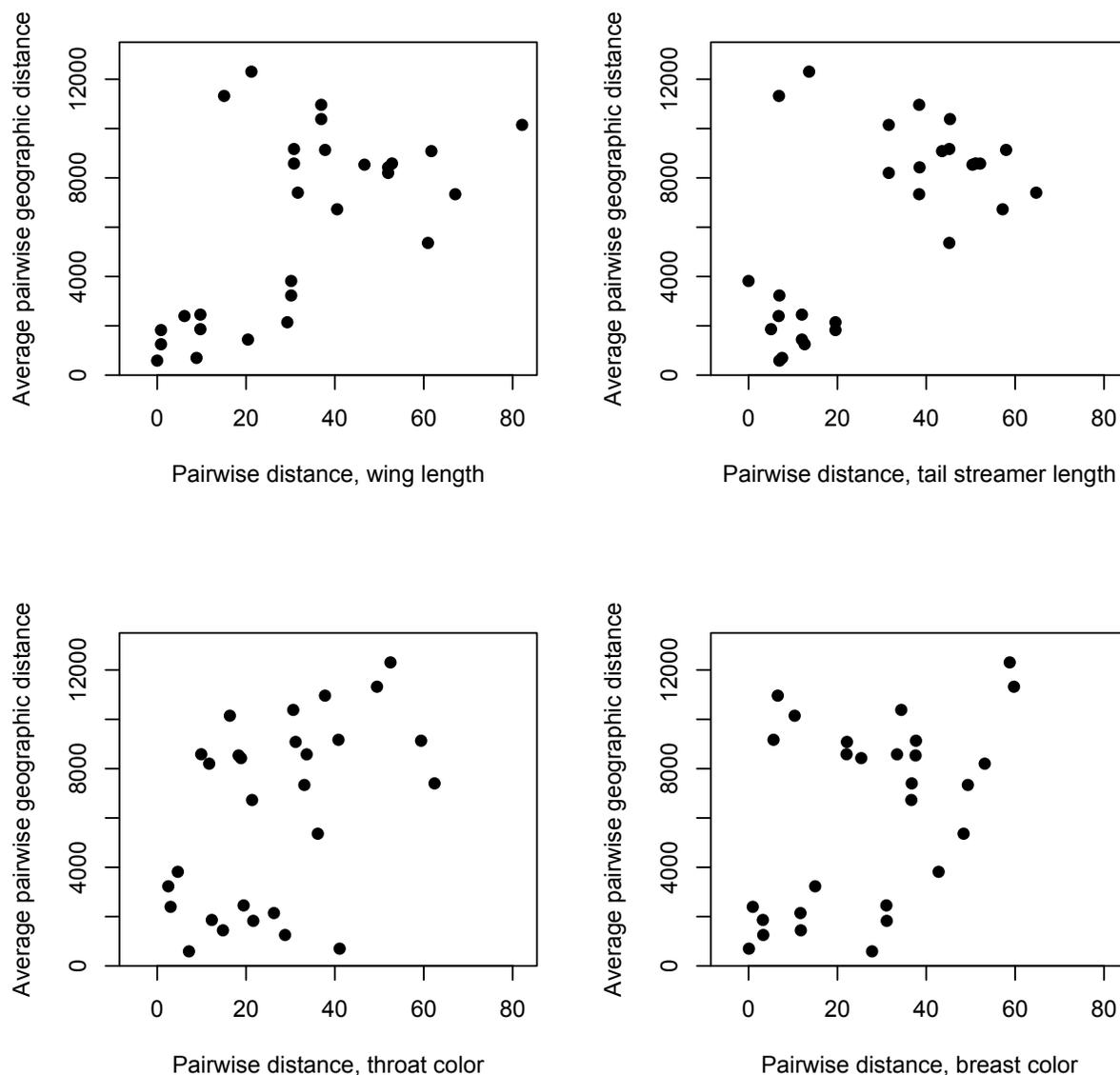
1024 **Figure 5.** Average pairwise genetic distance ( $F_{ST}$ ) as a function of environmental  
1025 distance among pairwise comparisons of eight barn swallow populations. Table 6 shows  
1026 results of mantel tests between  $F_{ST}$  and environmental distance; that there are no  
1027 statistically significant associations between genetic distance and these environmental  
1028 variables.

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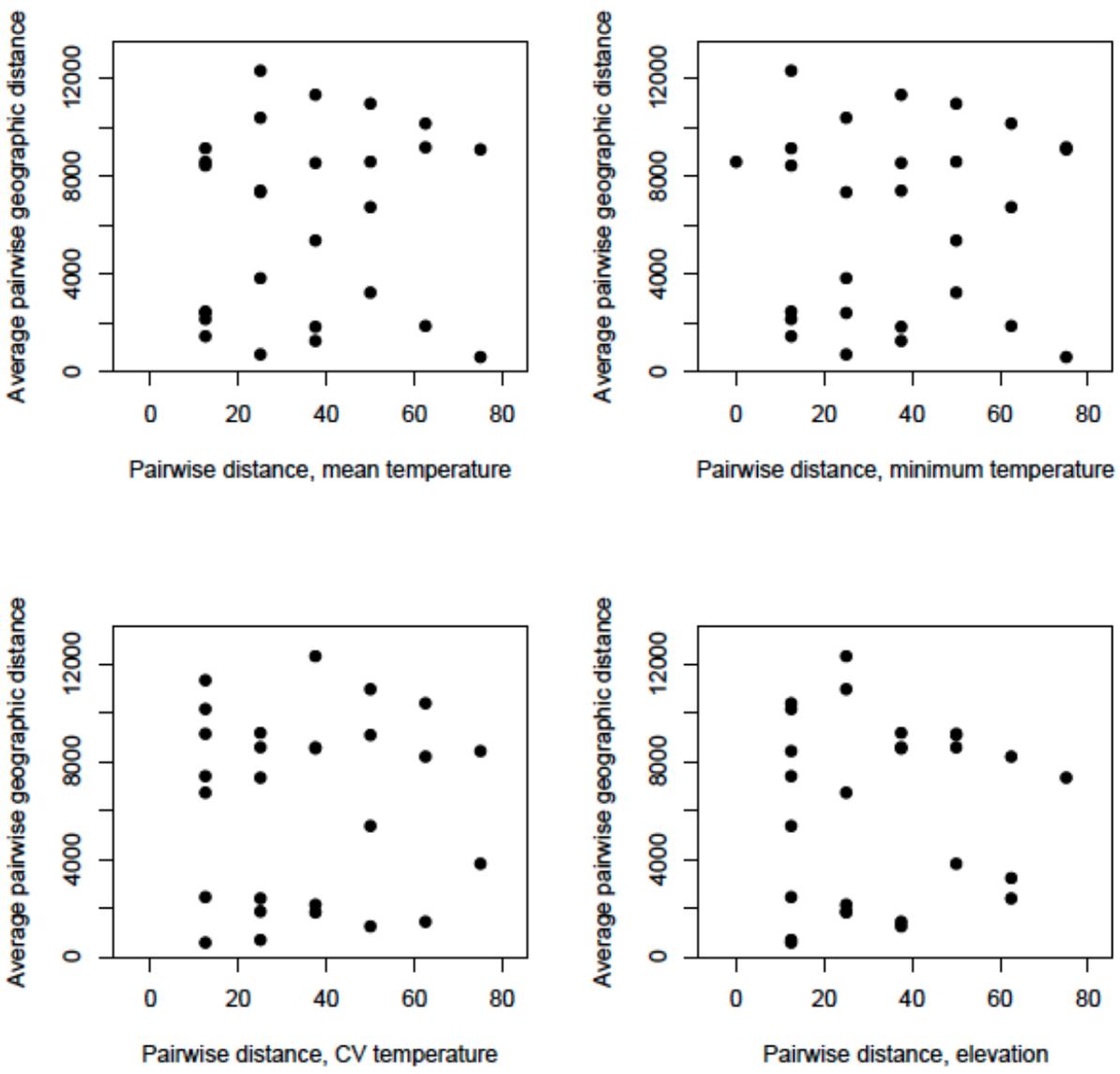
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1035 **Figure 6.** Average pairwise geographic distance as a function of phenotypic distance  
 1036 among pairwise comparisons of eight barn swallow populations indicate spatial  
 1037 autocorrelation between morphological and geographic distance. Table 6 shows results  
 1038 of mantel tests between geographic distance and phenotype distance; all of these  
 1039 associations are statistically significant.  
 1040

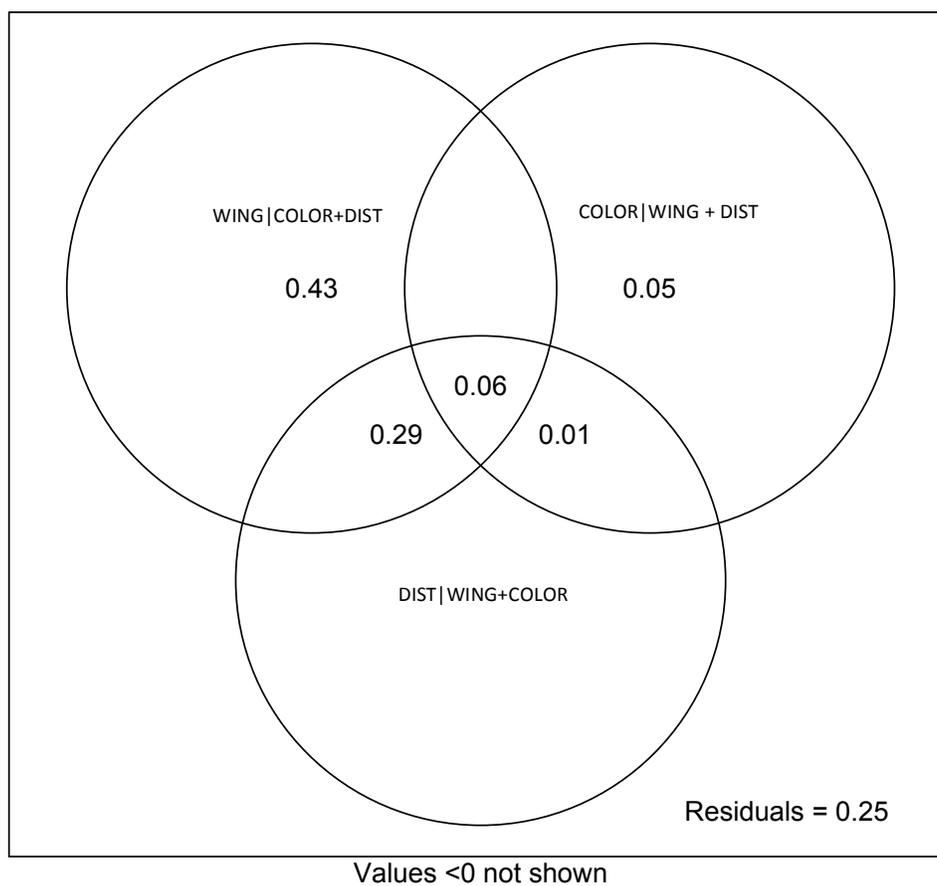


1041

1042 **Figure 7.** Average pairwise geographic distance as a function of environmental distance  
1043 among pairwise comparisons of eight barn swallow populations. Table 6 shows results  
1044 of Mantel tests between geographic distance and environmental distance; none of these  
1045 associations are statistically significant.  
1046



1047 **Figure 8.** Variance partitioning plot to represent the extent to which natural selection (WING),  
1048 sexual selection (COLOR), and geographic distance (DIST) explain variance in pairwise,  
1049 genome-wide differences in  $F_{ST}$ . Note that the matrix title before the symbol '|' indicates the  
1050 variance due to that matrix that contributed to genome-wide divergence, conditioned upon the  
1051 matrices listed after the '|' symbol.  
1052  
1053  
1054



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