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2 **Intraspecific genomic variation and local adaptation in a young hybrid**  
3 **species**

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25 **ABSTRACT**

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27 Hybridization increases genetic variation, hence hybrid species may have greater evolutionary  
28 potential once their admixed genomes have stabilized and incompatibilities have been purged.  
29 Yet, little is known about how such hybrid lineages evolve at the genomic level following  
30 their formation, in particular their adaptive potential. Here we investigate how the Italian  
31 sparrow (*Passer italiae*), a homoploid hybrid species, has evolved and locally adapted to its  
32 variable environment. Using restriction site-associated DNA sequencing (RAD-seq) on  
33 several populations across the Italian peninsula, we evaluate how genomic constraints and  
34 novel genetic variation have influenced population divergence and adaptation. We show that  
35 population divergence within this hybrid species has evolved in response to climatic variation,  
36 suggesting ongoing local adaptation. As found previously in other non-hybrid species,  
37 climatic differences appear to increase population differentiation. We also report strong  
38 population divergence in a gene known to affect beak morphology. Most of the strongly  
39 divergent loci among Italian sparrow populations do not seem to be differentiated between its  
40 parent species, the house and Spanish sparrows. Unlike in the hybrid, population divergence  
41 within each of the parental taxa has occurred mostly at loci with high allele frequency  
42 difference between the parental species, suggesting that novel combinations of parental alleles  
43 in the hybrid have not necessarily enhanced its evolutionary potential. Rather, our study  
44 suggests that constraints linked to incompatibilities may have restricted the evolution of this  
45 admixed genome, both during and after hybrid species formation.

46

47 **KEYWORDS:** Local adaptation, hybrid species, *Passer* sparrows, genomic incompatibilities,  
48 hybrid constraints, genome evolution.

49

50 **INTRODUCTION**

51

52 Hybridization is an evolutionary process that has been increasingly studied in the last decade  
53 (Abbott et al., 2013; Marques, Meier & Seehausen, 2019; Taylor & Larson, 2019). It can have  
54 a wide array of consequences, ranging from speciation reversal, reinforcement of prezygotic  
55 barriers to gene exchange, adaptive introgression and hybrid speciation. In particular, hybrid  
56 speciation – the formation of new species as a result of hybridization (Mallet, 2007) – can be  
57 seen as one of the most creative outcomes of hybridization. Especially the case of homoploid  
58 hybrid speciation (HHS) is thought to be rare given that reproductive isolation from the  
59 parental species does not automatically derive from differences in ploidy levels. Nevertheless,  
60 in the last decade, several compelling cases of HHS have been described in animals (Abbott et  
61 al., 2013; Mallet, 2007; Schumer, Rosenthal & Andolfatto, 2014). Mathematical models have  
62 addressed the mechanisms by which hybrid populations develop reproductive isolation from  
63 the parental lineages leading to HHS. Some studies suggest that geographic isolation of the  
64 hybrid from the parental taxa (Buerkle, Morris, Asmussen, & Rieseberg, 2000), as well as the  
65 genetic architecture and selection pressures on adaptive loci linked to incompatibility loci, is  
66 needed for the development of reproductive isolation from the parental taxa (Comeault,  
67 2018). Other studies argue that HHS can occur solely by the rapid development of  
68 reproductive barriers via sorting of genetic incompatibilities (Schumer, Cui, Rosenthal, &  
69 Andolfatto, 2015).

70

71 However, most of these theoretical and empirical studies focused on making a case for  
72 demonstrating HHS while little focus has been placed on analysing the evolutionary fate and

73 adaptive potential of hybrid species. In the long term, the establishment and success of a  
74 homoploid hybrid species only partially depends on the fast evolution of reproductive barriers  
75 that isolate it from its parental species and the purging of incompatibilities. Selection should  
76 also favour locally adapted allelic combinations to ensure the hybrid's ecological persistence  
77 and further adaptation to a potentially variable environment.

78

79 Genetic variability in hybrid lineages can be enhanced by the admixture process itself,  
80 through the generation of heterozygosity at loci that are differentially fixed in the parental  
81 species, novel re-arrangements of parental ancestry blocks, or the inheritance of parental  
82 standing genetic variation (Abbott et al., 2013). These processes can produce genetic variation  
83 in the hybrid that later may display a higher evolutionary potential than that found in non-  
84 hybrid species. Studies have shown that novel genetic combinations in hybrid lineages can  
85 substantially increase phenotypic variation and even lead to adaptive radiations (Keller et al.,  
86 2013; Meier et al., 2017; Rieseberg et al., 2003; Selz, Lucek, Young, & Seehausen, 2014).  
87 However, the evolutionary potential of a hybrid species can be hampered by genetic  
88 incompatibilities (i.e. Dobzhansky-Muller incompatibilities - DMIs) inherent to the formation  
89 of admixed genomes (Runemark, et al., 2018a; Schumer, et al., 2014b; Schumer et al., 2018;  
90 Trier, Hermansen, Sætre, & Bailey, 2014). Sorting of incompatibilities, originally arising and  
91 driving reproductive isolation between parental species, can generate symmetrical  
92 incompatibilities isolating the hybrid from both parental species (Schumer, et al., 2015;  
93 Buerkle, et al., 2000). However, this process may also constrain hybrid lineages long after  
94 hybridization has occurred, affecting their evolutionary potential (Eroukhmanoff et al., 2017;  
95 Runemark et al., 2018a). For instance, selection against DMIs can reduce the availability of  
96 variation responsive to adaptive evolution and hence, reduce population divergence and the

97 potential for local adaptation (Runemark, et al., 2018a). DMIs and incompatibilities in general  
98 often involve alleles at different loci that have never coexisted within the same genome.  
99 Accordingly, genomic variation in a hybrid species could be reduced at loci where alleles are  
100 differentially fixed between the parents, through purging of incompatible alleles. This may in  
101 turn constrain or fix certain genomic blocks through linkage with incompatibility loci and  
102 reduce the evolutionary potential in these genomic regions (Runemark et al., 2018a). Thus,  
103 the process of HHS includes both the sorting of incompatibilities and fixation of favourable  
104 genetic combinations to generate viable and functional genomes (Rieseberg et al., 2003;  
105 Runemark, et al., 2018b; Schumer, et al., 2014b). In this study we aim to provide insights to  
106 how admixture may ultimately constrain or facilitate adaptive divergence in a hybrid lineage  
107 and how genetic variation is generated and made accessible to selection.

108

109 In addition to constraints inherent to admixed genomes, hybrid lineages experience the same  
110 challenges as non-hybrid species do. The examination of factors that may mediate population  
111 differentiation (i.e. environmental variation or geography) in conjunction with inference  
112 regarding the role of drift and selection is therefore crucial to understand population  
113 divergence (Prunier, Colyn, Legendre, Nimon, & Flamand, 2015; Seeholzer & Brumfield,  
114 2018; Wang, 2013). Heterogeneity in abiotic factors such as climate and geography can  
115 determine patterns of population genomic divergence, either through geographic isolation  
116 (Isolation by distance, IBD) where gene flow is limited due to physical distance and  
117 geographic barriers (Meirmans, 2012; Slatkin, 1993; Wang, 2013; Wang & Bradburd, 2014),  
118 or through ecological isolation (isolation by environment IBE) (Shafer & Wolf, 2013; Wang  
119 & Bradburd, 2014), where individuals locally adapting to divergent habitats remain separated,  
120 facilitating genomic differentiation. Specific selective pressures, like those in IBE, could

121 result in differential changes in phenotypic traits that can also contribute to population  
122 genomic divergence; a process that in time could lead to isolation by adaptation (IBA)  
123 (Edelaar, Alonso, Lagerveld, Senar & Björklund, 2012; Nosil et al., 2008).

124

125 In the absence of geographic isolation, genetic and phenotypic population divergence can be  
126 hampered by gene flow (Hendry & Taylor, 2004; Räsänen & Hendry, 2008; Stuart et al.,  
127 2017), limiting local adaptation, although the directionality of causation of these processes is  
128 debatable. The opposite process can also occur; local adaptation may constrain gene flow,  
129 favouring divergence between populations and even lead to ecological speciation (Gosden,  
130 Waller & Svensson, 2015; Räsänen & Hendry, 2008, Nosil, 2012). In the specific case of  
131 hybrid lineages, it has also been argued that incompatibilities could reduce gene flow between  
132 hybrid populations (Bierne, Welch, Loire, Bonhomme, & David, 2011), especially when  
133 genes under ecological selection are coupled with DMI loci (Seehausen, 2013), which in turn  
134 may facilitate local adaptation (Eroukhmanoff, Hermansen, Bailey, Sæther & Sætre, 2013;  
135 Trier et al., 2014).

136

137 In this study we investigate how the homoploid hybrid Italian sparrow (*Passer italiae*) has  
138 evolved since its formation. We focus on how constraints and novel genetic variation, linked  
139 to admixture, have impacted its genomic evolvability, limiting or favouring its adaptive  
140 potential and ultimately its population divergence. The Italian sparrow is a homoploid hybrid  
141 species resulting from past hybridization between the house sparrow (*Passer domesticus*) and  
142 the Spanish sparrow (*Passer hispaniolensis*) (Hermansen et al., 2014; Trier et al., 2014). This  
143 hybridization event likely occurred when the house sparrow spread into Europe alongside  
144 agriculture, approximately 6 kyr BP (Hermansen et al., 2011; Ravinet et al., 2018; Elgvin et

145 al., 2017). It is possible that this hybrid species originated through a period of multiple  
146 hybridization events (Runemark, et al., 2018a; Elgvin et al., 2017) with rapid evolution of  
147 reproductive barriers from both parental species (Trier et al., 2014; Hermansen et al., 2014),  
148 despite some localized ongoing gene flow in parts of Italy and Europe. In mainland Italy the  
149 genome is admixed with a slightly higher contribution from the house sparrow (Elgvin et al.,  
150 2017). It is reproductively isolated from its parental species, with strong post-zygotic barriers  
151 associated with mito-nuclear and sex-linked incompatibilities (Elgvin et al., 2017; Trier et al.,  
152 2014).

153  
154 Patterns of population divergence and local adaptation at the genomic level have not yet been  
155 investigated in the Italian sparrow, nor the extent to which genomic constraints might have  
156 affected population divergence in this species. We limited our study to mainland populations  
157 across the Italian peninsula, excluding populations from Mediterranean islands as they are  
158 likely influenced by separate, independent hybridization events (Runemark, et al., 2018a). We  
159 assessed population divergence and the role of climatic variation on genomic divergence. Our  
160 results suggest that genetic divergence within the Italian sparrow is driven by climatic  
161 variation. We report patterns of isolation by environment (IBE), which appears to be driven  
162 primarily by temperature. We identify some outlier loci of adaptive divergence associated  
163 with precipitation and beak height variation. To determine the nature of the genomic  
164 divergence patterns found in the hybrid species, we examined the ancestry of the hybrid  
165 genome and genomic divergence in its parental species. Our results demonstrate that most loci  
166 involved in local adaptation in the hybrid species are little differentiated between the parental  
167 species, suggesting that incompatibilities may play a role in constraining population  
168 divergence. Conversely, loci involved in local adaptation within each parent species seem to

169 have previously been under divergent selection between the parental taxa, which is consistent  
170 with the natural history of both species (Ravinet et al., 2018). Overall, genomic divergence  
171 and local adaptation seem to be highly polygenic both in the hybrid and the parent species,  
172 albeit different loci are involved in adaptive intraspecific divergence.

173

## 174 **METHODS**

175

### 176 **Study species and sampling**

177

178 The Italian sparrow is distributed across the Italian peninsula and a few Mediterranean  
179 islands. Of its parental species the house sparrow has a wider native distribution, extending  
180 throughout large parts of Eurasia, whereas the Spanish sparrow is located around the  
181 Mediterranean Sea and eastwards to Central Asia (Summers-Smith, 1988). We concentrated  
182 on the mainland distribution of the Italian sparrow sampling several populations across the  
183 Italian peninsula.

184

185 Birds were caught using mist nets. Blood samples were obtained by puncturing the left  
186 brachial vein and stored in standard Queen's lysis buffer. Individuals were released  
187 immediately after sampling. All relevant sampling permits were obtained from the regional  
188 authorities.

189

190 We sampled a total of 131 (68 males and 63 females) Italian sparrows from 8 populations  
191 across Italy (Fig.1A, Table.S1). These populations are geographically well spread  
192 representing most of the mainland distribution of the Italian sparrow. In addition, we sampled



193 82 Spanish sparrows (51 males and 31 females) from Spain, Italy, Kazakhstan and Sardinia  
194 and 75 house sparrows (49 males, 26 females) from Norway, Switzerland, Spain and France.  
195 Per location between 13 and 27 individuals were sampled (Table.S1).

196

### 197 **DNA extraction and sequencing**

198

199 Genomic DNA was purified from blood samples using Qiagen DNeasy 96 Blood and Tissue  
200 Kits (Qiagen N.V., Venlo, The Netherlands) according to the manufacturer's instructions. The  
201 protocol was slightly modified by adding 125 ul of blood stored in Queen's lysis buffer and  
202 warming the Qiagen Elution Buffer (EB) to 40°C to increase yield of DNA. DNA isolates  
203 were stored in EB. Double digestion of the genomic DNA for ddRAD sequencing was  
204 performed using EcoRI and MseI restriction enzymes following the protocol by Peterson,  
205 Weber, Kay, Fisher & Hoekstra (2012). Genomic DNA was digested and ligated to respective  
206 adapters comprising EcoRI and MseI restriction overhangs. Molecular identifier tags were  
207 added with PCR amplification. Resulting individual sample libraries were pooled and library  
208 pools were size selected for fragments between 500-600bp with gel electrophoresis and  
209 extraction of the respective size range. The size selected library pools were then sequenced  
210 using an Illumina Nextseq500 machine and the 1x75bp sequencing format. On average, 2.4  
211  $\times 10^6$  single reads were produced per sample. Library preparation, sequencing, demultiplex  
212 and trimming of the adapters were performed by Ecogenics GmbH (Balgach, Switzerland)  
213 ([www.ecogenics.ch](http://www.ecogenics.ch)).

214

### 215 **Mapping to reference genome and variant calling**

216

217 RAD sequences were quality checked by FASTQC (Andrews, 2010) and mapped to the house  
218 sparrow reference genome, assembled by Elgvin et al. (2017), with BWA-MEM (v 0.7.8) (Li  
219 & Durbin, 2009) using the default parameters with the exception of using the -M flag  
220 allowing Picard compatibility for further analysis. Bam files were sorted by coordinates using  
221 Picardtools (v 1.72) SortSam (<https://broadinstitute.github.io/picard/>). Identification of indels  
222 and local realignment was run using Genome Analysis Tool kit (GATK)'s  
223 RealignerTargetCreator and IndelRealigner (Auwera et al., 2014; Mckenna et al., 2010) with  
224 default parameters. We validated bam files with the Picardtools (v 1.72) ValidateSamFile  
225 tool.

226

227 From the realigned bam file a set of variants were called by GATK (v 3.7) HaplotypeCaller  
228 using the following cut off for filtering: a Phred based mapping quality score of 10, soft  
229 clipping of the last 5bp without the need to soft clip both ends (-rf OverclippedRead --  
230 filter\_is\_too\_short\_value and --do\_not\_require\_softclips\_both\_ends). The resulting individual  
231 genomic variant files (gVCF) were then combined by CombineGVCFs and merged using the  
232 GenotypeGVCFs tools. As our analyses were based on single nucleotide polymorphisms  
233 (SNPs), all indels were excluded using the GATK's SelectVariants tool. Variants in unplaced  
234 scaffolds were removed using SelectVariants. Individuals with a proportion of missing data  
235 greater than 0.75 were excluded at this early stage before further filtering.

236

237 SNPs were subsequently filtered by quality using vcftools v. 0.1.14 (Danecek et al., 2011) as  
238 follows: proportion of missing data < 0.8, genotype quality > 20, Depth of coverage > 10 and  
239 minor allele frequency of 0.02. Finally, non-variant sites present after filtering and excluding

240 missing-data-individuals, were removed using GATK's SelectVariants with the -env  
241 parameter.

242

243 After filtering we obtained a final VCF file including the Italian sparrow and its parental  
244 species (288 individuals, 131 Italian, 82 Spanish and 75 house sparrows) containing 2737  
245 high-quality SNPs and with mean proportion of per individual missing data of 0.13. This  
246 dataset was used to identify genomic divergence among species.

247

248 Within-species analyses were conducted using species-specific VCF files by selecting the  
249 correspondent samples, merging individual genomic variant files (gVCF) and genotyping  
250 using the GenotypeGVCFs and finally recalling variants within species. Filtering was  
251 conducted as described above. The Italian sparrow-only VCF file contains 131 individuals  
252 and 4387 SNPs from 8 localities. VCF files for each parental species were additionally  
253 filtered by minor allele frequency of 0.01. The house sparrow-only VCF includes 75  
254 individuals across 4 localities and 6503 high-quality-SNPs and a Spanish sparrow VCF file  
255 with 1320 SNPs across 82 individuals from Spain, Kazakhstan and two localities in Italy;  
256 Fontanarosa in the Gargano peninsula and Sardinia. The average proportion of individual  
257 missing data for these species-specific VCF files are 0.12, 0.12 and 0.13 for the Italian-only,  
258 house-only and Spanish-only files, respectively.

259

### 260 **Investigating population divergence within the Italian sparrow**

261

262 To evaluate population structure and divergence in the hybrid species we used a SNP set  
263 containing 4387 loci identified across 8 Italian localities (N=131). We ran admixture analysis

264 and principal component analysis (PCA) using glPca in the R package ADEGENET 2.0  
265 (Jombart, 2008). We used vcftools (Danecek et al., 2011) and PLINK v. 1.9 (Chang et al.,  
266 2015) to transform the VCF file into format files (MAP, RAW, PED and BED) required by  
267 ADEGENET.

268

269 To assess the potential for isolation by distance among these Italian sparrow populations at  
270 different locations we used a multiple (and univariate) matrix regression with randomization  
271 (MMRR and UMRR, respectively) approach (Prunier et al., 2015; Wang, 2013), correlating  
272 geographic distance and genomic divergence (mean pairwise  $F_{ST}$ ) across all pairwise  
273 comparison of Italian sparrow populations. This method is described in the next section.

274

275 We used Tajima's D statistics to investigate signals of selection and potentially recent  
276 demographic change, which may have occurred post-hybridization. We also calculated mean  
277 values of Tajima's D, nucleotide diversity and  $F_{ST}$  for the Italian sparrow, using vcftools v.  
278 0.1.14 (Danecek et al., 2011).

279

280 To identify regions of divergence in the hybrid species, genome scan analyses were  
281 performed across the genome for the 8 populations of Italian sparrows. We calculated  
282 windowed  $F_{ST}$  and nucleotide diversity using a sliding window of 100kb in size with 25-kb  
283 steps. Nucleotide diversity was estimated retaining non-variant sites and avoiding minor allele  
284 frequency filtering. We also calculated Tajima's D on non-overlapping windows of 100kb,  
285 given that linkage disequilibrium (LD) tends to decay within this distance in sparrows (Elgvin  
286 et al., 2017) using a VCF file without minor allele frequency filtering to avoid bias by  
287 removing rare variants.

288

289 **Selection, local adaptation and environmental variation**

290

291 The Italian peninsula varies considerably in climate, thus we investigated whether genomic  
292 divergence covaried with environmental variation. Pairwise differences in climatic variables  
293 were regressed with the pairwise genetic distance between populations. We analysed five  
294 climatic variables obtained from the global climate data server, WorldClim (v. 2.0,  
295 <http://www.worldclim.org>) (Hijmans, Cameron, Parra, Jones & Jarvis, 2005), BIO1=Annual  
296 Mean Temperature, BIO4=Temperature Seasonality (standard deviation \*100),  
297 BIO12=Annual Precipitation and BIO15=Precipitation Seasonality (Coefficient of Variation).  
298 Values were retrieved using the R packages RGDAL (v 1.3-4, Bivand et al., 2017) and SP (v  
299 1.2-4) (Pebesma & Bivand, 2005), with a resolution of 1km. Geographic distance was  
300 obtained with the function spDistsN1 from the R package SP (v 1.2-4) and altitudinal data  
301 was gathered from the R package RASTER (v 2.6-7) (Hijmans, 2014) and SP (v 1.2-4) using  
302 the getData function. We also analysed phenotypic distance in two beak traits, mean beak  
303 height (BH) and beak length (BL), in each population.

304

305 To test for associations between environmental factors, geographic, altitudinal and phenotypic  
306 distances and genome-wide divergence we used univariate and multiple matrix regression  
307 with randomization (UMRR and MMRR respectively) approaches (Wang, 2013) and a  
308 modification implemented by Prunier et al., (2015), including commonality analysis (CA) to  
309 account for multicollinearity (non-independence) among environmental factors. Data were Z-  
310 transformed (i. standardization by subtracting the mean and dividing by the standard

311 deviation) to make regression coefficients of the predictor variables comparable (beta  
312 weights, Prunier et al., 2015).

313

314 MMRR is a multiple regression analysis on distance matrices used to quantify the  
315 contribution of environmental and geographic factors to patterns of genetic divergence  
316 (Wang, 2013). It allows the quantification of isolation by distance (IBD), isolation by  
317 environment (IBE) and even isolation by adaptation (IBA) when a phenotypic variable is  
318 included as predictor. One advantage of the method is that it not only resolves whether the  
319 dependent and independent variables are correlated but also quantifies the change and  
320 directionality (regression coefficients,  $\beta_n$ ) that the dependent variable (genomic distance) has  
321 with respect to multiple independent variables, i.e. geographic and environmental distances  
322 (Wang, 2013). The fit of the model is determined by the coefficient of determination ( $R^2$ ).  
323 Given the non-independent nature of the variables, the significance (*p-values*) of the  
324 variable's effects ( $\beta_n$ ) and fit of the model ( $R^2$ ) are estimated by randomized permutations of  
325 rows and columns of the dependent variable matrix (for more details see Wang, 2013).  
326 However, strong multicollinearity among predictors is still a limitation of this approach.  
327 Regression coefficients ( $\beta_n$ ), fit of the model ( $R^2$ ) and their significance can be affected by  
328 multicollinearity among explanatory variables (Kraha, Turner, Nimon, Zientek & Henson,  
329 2012; Nimon & Reio, 2011; Prunier et al., 2015). To overcome this caveat an incorporation of  
330 variance-partitioning procedures via commonality analysis (CA) can be used, implemented by  
331 Prunier et al., (2015). This method (CA) developed originally by Newton & Spurrell (1967)  
332 decomposes the model coefficients into unique (U) and common (C) variance components  
333 (Campbell & Tucker, 1992 in Prunier et al., 2015; K. F. Nimon & Oswald, 2013), allowing  
334 identifying the magnitude of collinearity and the unique (U) effect that a predictor variable

335 has on the dependent variable. The common (C) effect represents the proportion of variance,  
336 in the dependent variable, explained by the collinearity of the predictor evaluated and another  
337 explanatory variable; while the unique component (U) quantifies the variance explained by  
338 the unique effect of the predictor (Prunier et al., 2015).

339

340 CA allows determining unique (U) and common (C) contributions of each predictor to the  
341 response variable (pairwise  $F_{ST}$ ) while accounting for collinearity among predictors. The total  
342 effect ( $T=U+C$ ) of each predictor corresponds to the total effect that a predictor has to the  
343 variance explained by the model, independently of collinearity with other predictors, and the  
344 total variation a specific predictor accounts for is determined by  $T/R^2$ , which would be a  
345 portion of the variation explained by the model.

346

347 These methods have been shown to provide a better resolution of the effects of environment,  
348 geographic distance and phenotype, allowing us to identify patterns of IBD, IBE and IBA  
349 (Seeholzer & Brumfield, 2018). This approach is ideal for our analysis given the nature of our  
350 data. We are interested in understanding whether genomic divergence and gene flow within  
351 the Italian sparrow is linked to climatic, geographic and phenotypic variation. We ran UMRR  
352 and MMRR with 1000 permutations to estimate significance. We also performed variance-  
353 partitioning analysis by CA, 95% coefficient intervals of the commonality coefficient were  
354 calculated by bootstrapping 1000 replicates, as implemented by Seeholzer and Brumfield  
355 (2018).

356

357 We used pairwise geographical distance, altitudinal difference, climate disparity per  
358 environmental factor and pairwise mean phenotypic distance as predictor matrices and a

359 genomic distance matrix (pairwise  $F_{ST}$ ) as the dependent variable. As the number of predictor  
360 variables cannot be greater than the number of populations analysed in the MMRR analysis,  
361 two models were run. In model 1 only geographic and climate variables were used as  
362 predictors, while in model 2, altitude and one of the temperature variables were replaced by  
363 the phenotypic variables.

364

365 To identify SNP candidate loci under selection we ran an outlier analysis using Bayescan (v.  
366 2.1 – Foll & Gaggiotti, 2008), for the Italian sparrow and its parental species independently.  
367 Bayescan is a Bayesian approach based on the multinomial-Dirichlet model that uses  
368 differences in allele frequency to identify candidate loci under selection by decomposing  $F_{ST}$   
369 coefficients into population ( $\beta$ ) and loci ( $\alpha$ ) components; a reversible-jump MCMC evaluates  
370 models with and without selection and calculates posterior probabilities of the parameters  
371 under the different models (Foll & Gaggiotti, 2008).

372

373 Associations of genomic divergence and environmental (and phenotypic) variation can differ  
374 across the genome. Therefore we also evaluated such associations at the locus level (SNP), in  
375 the hybrid taxon, performing outlier analyses with BayeScEnv, version 1.1 (de Villemereuil  
376 & Gaggiotti, 2015). We used the same environmental variables ran on MMRR as predictors,  
377 including beak height and length. BayeScEnv, as Bayescan, is a genome-scan software based  
378 on Bayesian inference. To account for population structure it uses the F-model and to control  
379 for multiple testing, it returns false discovery rate statistics (Posterior Error Probability (PEP),  
380 q-value). This method allows the incorporation of environmental information, so that the  
381 associations between allele frequencies and environmental variables can be evaluated.

382



383 We ran BayeScEnv using the default parameters. As in Bayescan, the parameters  $\beta$  used in  
384 the neutral model as well as the locus-specific effect using  $\alpha$  are estimated. However a third  
385 model of local adaptation, estimating the parameter  $g$ , uses the environmental differentiation  
386 information. Significantly associated loci were determined by setting a FDR significance  
387 threshold of 5% for the correlation q-value of  $g$  (de Villemereuil & Gaggiotti, 2015).

388

389 To identify candidate genes associated to local adaptation we used the house sparrow  
390 annotation file developed by Elgvin et al. (2017). In the house sparrow linkage decays at  
391 approx. 100kb (Elgvin et al., 2017), thus we selected genes contained in regions at a  
392 maximum of 100kb distance from the Bayescan/BayeScEnv outlier loci. To further assess  
393 signatures of selection at the gene level we identified all genes from the house sparrow  
394 annotation file and calculated values of  $F_{ST}$ , Tajima's D, Pi and Dxy per-gene across the  
395 whole genome. Later we assessed whether our candidate genes show extreme values of the  
396 population statistics in comparison to the other genes genome-wide. For this per-gene analysis  
397 we used WGS data from the house and Spanish sparrow (2 populations per species) and 3  
398 Italian sparrows populations; data retrieved from Elgvin et al., (2017) and Ravinet et al.,  
399 (2018).

400

#### 401 **Investigating genomic constraints to population divergence linked to hybridization**

402

403 To determine the nature of the genomic divergence patterns in the hybrid species, and how  
404 they differ from non-hybrid species, we compared population genomic parameters of the  
405 parental species to the Italian sparrow. We also estimated ancestry patterns in the Italian

406 sparrow looking to shed light on the source of the genomic variation found in this hybrid  
407 lineage.

408

409 To identify how highly divergent loci in the hybrid are distributed, for instance whether they  
410 are located in genomic regions of high parent species divergence or not, we selected the top  
411 1% loci with the highest  $F_{ST}$  among all 8 Italian sparrow populations across the different  
412 localities and estimated ancestry as well as hybrid-parent  $F_{ST}$  and between-parents (SH)  $F_{ST}$   
413 values for these same loci. Similarly, we extracted the top 1% loci with the highest  $F_{ST}$  among  
414 house sparrow populations and among Spanish sparrow populations and as for the hybrid  
415 species, hybrid-parent  $F_{ST}$  and between-parents (SH)  $F_{ST}$  values were estimated for these  
416 highly variable loci. We also compared the observed patterns of Tajima's D between species.  
417 As for the hybrid, Tajima's D for the parental species was estimated using VCF files that  
418 were not filtered for minor allele frequency.

419

420 To evaluate whether loci involved in population divergence within the Italian sparrow  
421 correspond to loci of high or low genetic differentiation between the Italian and Spanish (IS  
422  $F_{ST}$ ) sparrows, Italian and house (IH  $F_{ST}$ ) sparrows or Spanish and house (SH  $F_{ST}$ ) sparrows,  
423 we performed logistic regressions on the probability of being an Italian  $F_{ST}$  outlier. In these  
424 models, the outlier status (outlier/non-outlier) of each locus (SNP) is the response variable,  
425 while additive and interaction effects of pairwise  $F_{ST}$  between the three species were tested as  
426 predictors.

427

428 We also used whole genome resequencing (WGS) data from Elgvin et al., (2017) and Ravinet  
429 et al., (2018) to estimate ancestry for the Italian sparrow genome. A total of 54 genomes were

430 used, a single population per parental lineage (10 Spanish sparrows from Kazakhstan and 14  
431 house individuals from Norway) and 3 Italian sparrow populations (Crotone, Guglionesi and  
432 Rimini) with 10 genomes per population. Data was phased prior to analysis (see Ravinet et  
433 al., 2018) and ancestry estimates were performed using the software LOTER (Dias-alves,  
434 Mairal, & Blum, 2018), a software package for local ancestry inference (LAI) that uses a  
435 copying model based on an optimization problem where switches of parental haplotypes are  
436 penalized by the regularization parameter  $\lambda$ . A final ancestry estimate is found by averaging  
437 results from different values of  $\lambda$  and several runs of the algorithm. Moreover, this package  
438 does not require statistical or biological parameters (i.e. recombination rate) to be specified,  
439 making it more accessible to non-model species.

440

441 Following this, we identified ancestry estimates for the 4387 RAD loci found across the 8  
442 Italian sparrow populations. When it was not possible to identify the ancestry estimate of a  
443 specific RAD locus its value was instead taken from the closest identified locus within a  
444 100kb window. We calculated house sparrow ancestry proportion across all 8 populations of  
445 the Italian sparrow. Also, given that LOTER assigns a specific ancestry estimate (house or  
446 Spanish ancestry) for each haplotype we weighted those estimates using the parental allele  
447 frequency difference (AFD), calculated from the WGS data, as a measure of certainty. Thus,  
448 the sign of the estimate symbolizes parental ancestry (negative values for house ancestry and  
449 positive for Spanish ancestry) and the value represents the degree of AFD between parental  
450 species. Values of zero show loci where alleles are segregating equally in the parental  
451 lineages, while values of 1 (or -1) occur on loci that are differentially fixed between the  
452 parents.

453

454 Evolution of recombination rate variation across the genome may have an effect on patterns  
455 of differentiation within and among species (Burri et al., 2015; Ortiz-Barrientos & James,  
456 2017; Ortiz-Barrientos, Engelstädter & Rieseberg, 2016). Therefore, we evaluated whether  
457 there was a correlation between recombination rate (estimates taken from a linkage map from  
458 Elgvin et al (2017)) and genomic differentiation ( $F_{ST}$ ) among populations for each of the  
459 species (house, Spanish and Italian sparrows) respectively.

460

## 461 RESULTS

462

### 463 **Genomic landscape of population divergence in the Italian sparrow**

464

465 As found in previous studies (Elgvin et al., 2017; Hermansen et al., 2011) our results support  
466 the mosaic nature of the hybrid Italian sparrow genome (Fig.1C, 1D). To evaluate the  
467 genomic variation among populations of the Italian sparrow we performed a PCA and  
468 admixture analyses from 8 locations across the Italian peninsula (N=131 individuals, 4387  
469 SNPs. Fig.1B, S1), covering a wide range of its mainland geographic distribution (Fig.1A).  
470 We found no evidence for genome-wide population structure, only moderate among-  
471 population clustering.

472

473 Estimated parameters of population divergence among Italian sparrows also showed a  
474 moderate genome-wide population divergence (mean  $F_{ST}$  across all 8 localities = 0.013,  
475  $\pi=2.595 \times 10^{-6}$ , Table.S2). Nonetheless, it was possible to identify regions of higher divergence  
476 in autosomes, with maximum  $F_{ST}$  values of  $\sim 0.17$  across populations and high nucleotide  
477 diversity (Fig.S2A, S2C).

478 Genome wide average of Tajima's  $D$  for the Italian sparrow was negative, as well as for the  
479 parental species, however, there is a significant difference between species. 1%  $F_{ST}$  outliers  
480 between Italian populations had higher nucleotide diversity than the genome wide average  
481 and interestingly these loci also showed elevated nucleotide diversity in the parental species,  
482 especially in the house sparrow (Table.S2).

483

#### 484 **Selection, local adaptation and environmental variation**

485

486 To further understand the genetic differentiation found among populations of the hybrid we  
487 tested patterns of IBD, IBE and IBA using the distances of several climatic factors and  
488 phenotypic traits, as well as altitudinal and geographic distances as predictor variables. We  
489 ran UMRR and MMRR models (Wang, 2013) and variance partitioning through commonality  
490 analyses (CA) (Prunier et al., 2015; Seeholzer & Brumfield, 2018). We found no evidence for  
491 IBD in our dataset (Table.1, Table.2, Table.S3). In UMRR (Table.1) geographical distance  
492 (GEO) showed a non-significant relationship ( $R^2=0.053$ ,  $\beta=0.004$ ) to genetic differentiation  
493 among populations. Its contribution in the multivariate model (MMRR) was non-significant  
494 ( $\beta=0.003$ ,  $P=0.34$ ) and under the commonality analysis the unique ( $U=0.03$ ) and common  
495 ( $C=0.02$ ) effects were considerably small (Table.2). Isolation by environment (IBE) appeared  
496 to be a more determining factor. Results from UMRR and MMRR yielded evidence that  
497 climate is driving genetic differentiation within the Italian sparrow, suggesting adaptation to  
498 climate (or some unmeasured factor correlate of climate). In particular, temperature  
499 seasonality explained a significant proportion of the genetic variation, (Table.1, Fig.S3), with  
500 a  $R^2=0.163$  and  $\beta$  weight of  $\beta=0.007$ . The multivariate model including all the climatic  
501 factors, altitude and geographic distances as predictors (MMRR – model 1, Table.2),

502 explained 25% of the inter-population variation in  $F_{ST}$  within the Italian sparrow ( $R^2=0.25$ ).  
503 Consistent with the results from UMRR, temperature seasonality yielded the highest  $\beta$  weight,  
504 with a considerable explanatory power ( $\beta=0.007$ ) (Table.2), accounting for 8% of the  
505 variation explained by the model. However, variance partitioning (CA) showed its unique  
506 contribution was almost negligible, meaning the interaction with other variables (collinearity)  
507 had a larger effect ( $U=0.003$ ,  $C=0.2$ , Table.2, Fig.S4).

508

509 While mean annual temperature explained a considerable amount of the variance (A.TEMP,  
510 Table.2) most of it fell into the unique factor ( $U=0.14$ ) and its beta weight was non-significant  
511 ( $\beta=0.001$ ,  $P=0.89$ ). Mean annual precipitation showed similar results (A.PREC,  $T=24\%$ ,  
512 Table.2). This suggests that there is collinearity between climatic factors. Unique (U) and  
513 common (C) contributions to the variation, estimated by CA (Table.2, Fig.S4), showed mean  
514 annual temperature ( $T=0.16$ ) and mean annual precipitation ( $T=0.06$ ), as the major  
515 contributors, accounting for 64% and 24% of the variation explained by the model,  
516 respectively (Table.2). However, beta weights for these predictors were not significant.  
517 Moreover, when removing mean annual temperature from the model (MMRR – model 2,  
518 Table.S3) temperature seasonality was no longer significant ( $P=0.1$ ), supporting the  
519 collinearity effect among climatic variables.

520

521 Finally, evaluating IBA, incorporating beak morphology as predictors, the univariate (UMRR,  
522 Table.1) and multivariate (MMRR, Table.S3) models showed that these phenotypic traits do  
523 not explain a significant amount of the genomic divergence among Italian sparrow  
524 populations. The univariate models for each of the beak traits showed a non-significant

525  $R^2=0.036$  ( $P > 0.34$ ), and in the multivariate model (MMRR - model 2, Table.S3) beta weights  
526 were low ( $\beta=0.001$  for BEAK.H and  $\beta=0.002$  BEAK.L) and non-significant.

527

528 To determine whether highly divergent genomic regions are associated with environmental  
529 factors and identify potential genes associated to local adaptation to climate we used a  
530 genome scan approach implemented by the software BayeScEnv (de Villemereuil &  
531 Gaggiotti, 2015). Five loci were found to be under selection through correlation with  
532 environmental variables. On chromosome 5 two outlier loci were associated with mean  
533 annual precipitation. One of these displayed values of Tajima's  $D=-0.833$  and  $F_{ST}=0.136$   
534 among Italian sparrow populations. A locus on chromosome 15 (with values of  $F_{ST}=0.172$   
535 among Italian populations) was also found to associate significantly with mean annual  
536 precipitation (Fig.2A) while presenting high, although non-significant, q-values of  $g$  for mean  
537 annual temperature and altitude (Fig.S5A and S5C). Consistently, divergence between species  
538 pairs for these loci was low (Table.S4). Similarly, chromosome 3 and 2 contained one outlier  
539 locus each (with across Italian localities  $F_{ST}=0.050$  and  $F_{ST}=0.084$ , respectively) associated to  
540 precipitation seasonality (Fig.2B). We also found three candidate loci under selection related  
541 to beak morphology, associated with population divergence in beak height (Fig.2C,  
542 Table.S4).

543

544 Further, we used the software Bayescan (Foll & Gaggiotti, 2008) to identify loci under  
545 selection across the Italian sparrow populations, independently on whether they are associated  
546 to specific environmental factors or phenotypic traits. We also performed the same analysis in  
547 each of the parental species to evaluate whether the hybrid lineage presents similar loci under  
548 selection as those in the parental taxa. Three outlier loci were identified as under selection in

549 the Italian sparrow; one locus on chromosome 6, a second locus on chromosome 20 and  
550 another in chromosome 15. The latter was previously identified as associated with mean  
551 annual precipitation by BayeScEnv (Fig.3A).

552

553 Within the putative regions under selection (i.e. 100kb around the outlier loci) we identified  
554 potential genes of interest that may be associated to climatic variation (Table.S4). To further  
555 assess signatures of selection, specifically at the gene level, we used WGS data from Elgvin et  
556 al., (2017) and Ravinet et al., (2018) to calculate per-gene population statistics ( $F_{ST}$ , Tajima's  
557  $D$ ,  $\Pi$ ,  $d_{xy}$ , (Table.S4)). The gene GDF5 was identified as a 5% gene- $F_{ST}$  outlier (one-tailed  
558 test) presenting a gene- $F_{ST}=0.047$  (Table.S4). The GDF5 gene, also known as BMP-14,  
559 involved in bone and cartilage development, encodes a growth differentiation factor protein  
560 related to the BMP (bone morphogenetic protein) gene family (Reddi & Reddi, 2009), a gene  
561 family involved in skeletal and jaw development (Bleuming et al., 2007; Cerny et al., 2010;  
562 Kaucka & Adameyko, 2019).

563

564 In the house sparrow (75 individuals, 6503 SNPs, 4 localities) 8 candidate loci on  
565 chromosomes 1, 5 and 8 were inferred to be significantly under selection (Fig.3B, Table.S4).  
566 Similarly, in the Spanish sparrow (1320 SNPs across 82 individuals from 4 localities) 8  
567 candidate loci (on chromosomes 1, 2, 3 and 5) were also identified using Bayescan (Fig.3C,  
568 Table.S4).

569

570 Only one of the outlier loci was simultaneously identified by both genome scan approaches  
571 (Bayescan and BayeScEnv) for the Italian sparrow. The lack of overlapping outlier loci under  
572 selection among the three species may be due to differential selective pressures acting in the



573 hybrid and its parental species. However, further work specifically investigating these loci is  
574 necessary to properly assess the role of selection in generating this pattern.

575

### 576 **Hybrid constraints to population divergence**

577

578 We compared population genomic parameters between the house, Spanish and Italian  
579 sparrows and estimated ancestry of the hybrid loci to determine whether genomic constrains  
580 are playing an important role in the genomic divergence of the hybrid species or whether  
581 genomic variation, boosted by the hybridization event, facilitates population structuring. We  
582 also looked to identify differences in genetic variation patterns between the hybrid and its  
583 parent species.

584

585 Population divergence in the house sparrow, with a maximum value of  $F_{ST}=0.33$  across all  
586 chromosomes (mean  $F_{ST}=0.019$ ) and mean nucleotide diversity of  $\pi=2.996 \times 10^{-6}$  (Fig.S2D,  
587 S2F, Table.S2) was similar to that in the Spanish sparrow (mean  $F_{ST}=0.021$ ,  $\pi=1.642 \times 10^{-6}$ ),  
588 with a maximum  $F_{ST}$  of 0.34 (Fig.S2G, S2I, Table.S2). In contrast, divergence in the Italian  
589 sparrow was lower, with a maximum  $F_{ST}$  value of  $\sim 0.17$  (mean  $F_{ST}=0.013$ ,  $\pi=2.595 \times 10^{-6}$ ;  
590 Fig.S2A, S2C, Table.S2).

591

592 Ancestry estimates showed that the probability of being an Italian  $F_{ST}$  outlier was not related  
593 to the ancestry proportion across Italian sparrow populations (Logistic regression  
594 estimate=0.229,  $P=0.56$ , Fig.5A). However, Italian outlier loci (1%  $F_{ST}$  outliers) segregated  
595 for alleles from both parents, as most of the genome presents a mosaic pattern (Fig.5B, S6,  
596 S7). Yet, outlier positions showed low allele frequency differentiation (AFD) between the

597 parents, as the majority of weighted ancestry values for outlier loci were around zero  
598 (Bartlett's test of homogeneity of variances using absolute values of weighted ancestry:  
599  $\text{Chi}^2=806789.041$ ,  $P=0.00$ , Fig.5B), also supporting the low SH  $F_{ST}$  values in highly  
600 divergent loci in the Italian sparrow, in comparison with the parental taxa (Fig.4A, 5C). In  
601 contrast, inherited parental blocks that are differentially fixed (regions with weighted ancestry  
602 values of 1 or -1 and with high values of between-parent-species-differentiation (SH  $F_{ST}$ ))  
603 showed lower levels of genetic differentiation within the Italian sparrow (Fig.4A, 5B).

604

605 Moreover, in the additive model, where both comparisons of the hybrid and each of the parent  
606 species are evaluated ( $IH F_{ST} + IS F_{ST}$ ), the probability of being an  $F_{ST}$  outlier within the  
607 Italian sparrow decreased with Italian-Spanish ( $IS F_{ST}$ ) genetic divergence (Table.3,  
608  $P=0.0127$ ). A negative, yet non-significant, correlation was also found between the highly  
609 divergent regions within the hybrid species and between parental species genetic divergence  
610 ( $SH F_{ST}$ ,  $P=0.0926$ ) (Table.3).

611

612 Additionally, none of the highly divergent regions within the hybrid lineage differed  
613 substantially from both of the parental species simultaneously, indicating that private alleles  
614 do not account for most of the population differentiation in the hybrid species (Fig.4B).  
615 Furthermore, the majority of private alleles have extremely low frequencies and were  
616 removed from the analysis when applying MAF filtering.

617

618 In contrast to the patterns found for the highly divergent regions in the Italian sparrow, 1%  
619  $F_{ST}$  outliers within each of the parental species present high parental genomic divergence  
620 (high SH  $F_{ST}$  values, Fig.5C). The 1% outlier loci of within house sparrow  $F_{ST}$  showed higher

621 divergence between the parental species Spanish-House (SH  $F_{ST}$ ) than those within the hybrid  
622 species, and the same pattern was found for the Spanish sparrow (Fig.5C). Furthermore,  
623 highly divergent loci within each of the parental species did not correspond to those found  
624 within the hybrid Italian sparrow (Fig.4C, 4D).

625  
626 We find some evidence suggesting that recombination rate could explain part of the genomic  
627 divergence pattern found within the Italian sparrow ( $R^2= 0.00085$ ,  $P= 0.033$ ) and within the  
628 Spanish sparrow ( $R^2= 0.003211$ ,  $P= 0.026$ ). However, despite significance, extremely low  
629 level of variation in divergence between loci is explained by recombination rate (exemplified  
630 by the low  $R^2$ , 0.3 % at most). For the house sparrow, there was not significant correlation  
631 ( $R^2= -9.67e-07$ ,  $P= 0.319$ ) (Fig.S8).

632  
633 We also found an overall higher proportion of negative genome-wide Tajima's D in the  
634 Italian sparrow (Fig.S2B) as well as in the house (Fig.S2E) and Spanish sparrows (Fig.S2H).  
635 However, the hybrid species differed significantly from the parental species (Table.S2).

636

## 637 **DISCUSSION**

638

639 Little is known about how a newly formed hybrid species evolves beyond just a handful of  
640 generations. The majority of genomic variation in a hybrid lineage will be derived from  
641 admixture, standing genetic variation inherited from the parental species and novel mutations  
642 after hybridization. This variation may ultimately facilitate rapid divergence, whereas genetic  
643 incompatibilities may constrain hybrid genome evolution (Runemark, et al., 2018a), including  
644 their potential for local adaptation. Purging of incompatibilities can remove adaptive variation

645 in regions in physical linkage to DMIs (Schumer et al., 2018). In this study we investigated  
646 the extent to which populations of a relatively young hybrid lineage have diverged in  
647 response to climatic variation. We further investigated to what extent divergence in the hybrid  
648 occurs at loci where variation is generated by admixture itself, in turn fuelling local  
649 adaptation.

650

### 651 **Population divergence in the Italian sparrow**

652

653 We found moderate, but significant genome-wide population divergence, in line with what  
654 has been previously found using neutral markers (Eroukhanoff et al., 2013), and consistent  
655 with ongoing gene flow between populations of Italian sparrows across the Italian peninsula,  
656 although other scenarios could also explain this pattern. The young age of this hybrid lineage,  
657 thought to be of approx. 6.000 years (Hermansen et al., 2011; Ravinet et al., 2018), may  
658 explain this pattern, as there may not have been sufficient time for populations to strongly  
659 diverge. Given the hybrid nature of the Italian sparrow, genomic constraints may also be an  
660 important factor in its evolution, hampering population divergence. Consistently, we found  
661 negative values of Tajima's D suggesting that regions in the genome are experiencing  
662 purifying selection, potentially linked to purging of incompatibilities. Nonetheless, genetic  
663 variation may also have been maintained by balancing selection, as we found regions  
664 harbouring high nucleotide diversity and loci exhibiting high divergence among populations,  
665 suggesting that there is room for variation in the hybrid genome. Also, variation in  
666 recombination rate could in part explain some of genomic differentiation identified.

667

668 Interestingly, this general pattern of differentiation was comparable but somewhat lower than  
669 the pattern of population divergence ( $F_{ST}$ ) we report for within each of the parent species.  
670 Yet, it is difficult to draw further conclusions on the within-species divergence in the parental  
671 lineages since the populations sampled are separated by greater geographic distances than  
672 those of the hybrid species, which likely affects relative divergence.

673  
674 Tajima's D differed between the hybrid and the parental lineages; however, all three species  
675 exhibited a negative genome wide average. In the house sparrow, this result supports recent  
676 work demonstrating a population expansion about 6 Kya (Ravinet et al., 2018). A negative  
677 Tajima's D in the Italian sparrow could also suggest recent population expansion that could  
678 mask the high nucleotide diversity expected from the hybridization event itself. Tajima's D in  
679 the Italian sparrow has been found to be negative overall and positive values were mostly  
680 located in regions of novel divergence, putatively under balancing selection (Elgvin et al.,  
681 2017).

682  
683 Loci of high differentiation among Italian sparrow populations had higher nucleotide diversity  
684 in the parental species than the corresponding genome wide average. One explanation could  
685 be that hybrid genetic variation has its origin in standing genetic variation inherited from the  
686 parental species, maintained by balancing selection and divergent natural selection following  
687 hybridization, leading to population differentiation possibly through the selection of variants  
688 playing a role in local adaptation to climate (Guerrero & Hahn, 2017). Tajima's D in these  
689 outlier loci was negative, yet higher than the genome wide average. However, there are a  
690 variety of processes, including demography, purifying selection and the break up of parental  
691 blocks, that can have confounding effects on the Tajima's D patterns observed in these regions,

692 therefore, it is difficult to conclusively identify the processes that could have generated this  
693 pattern.

694

695 **Selection, local adaptation and environmental variation**

696

697 Assessing genomic patterns across a spatially heterogeneous distribution, in correlation with  
698 factors that can play a role in genomic divergence, can help us elucidate the processes that  
699 have determined population differentiation in hybrid lineages. It can also give insights to the  
700 adaptive potential of the species (local adaptation and gene flow reduction) or whether  
701 genomic differentiation is essentially a result of genetic drift, where patterns of genetic  
702 variation are shaped by low gene flow (Prunier et al., 2015; Seeholzer & Brumfield, 2018;  
703 Wang, 2013).

704

705 To assess adaptive divergence and gene flow, we evaluated IBE, IBA through beak  
706 divergence and IBD. We did not find evidence for IBD or IBA, but the significant correlation  
707 between genetic distance and climatic variation is consistent with IBE. Our results suggest  
708 that climatic differences, with temperature as the main factor, likely contribute to reduced  
709 gene flow between populations in the Italian sparrow, possibly as a result of local adaptation.  
710 Previously, precipitation has been found to correlate with beak morphology variation in this  
711 species (Runemark, Fernández, Eroukhmanoff, & Sætre, 2018b), and could indirectly be  
712 mediating gene flow between phenotypically divergent populations (Eroukhmanoff et al.,  
713 2013). Differential changes in phenotypic traits responding to selective pressures can have an  
714 effect on local adaptation that may sometimes lead to IBA (Edelaar, Alonso, Lagerveld, Senar  
715 & Björklund, 2012). However, when directly evaluating beak trait variation as a predictor of

716 overall genomic differentiation among populations of the Italian sparrow we did not find  
717 evidence for IBA.

718

719 Patterns of adaptive divergence with ongoing gene flow have also been extensively reported  
720 in species of non-hybrid origin (de Leon, Bermingham, Podos, & Andrew, 2010; Marques et  
721 al., 2016; Martin et al., 2013; Raeymaekers et al., 2017), which suggests that despite the  
722 possibility of constraints reducing the evolvability of this hybrid species (Runemark, et al.,  
723 2018a), there is also potential for adaptive divergence leading to local adaptation, as in non-  
724 hybrid lineages. In fact, theory suggests that incompatibilities could facilitate local adaptation  
725 by the coupling of genes under ecological selection and DMI loci (Seehausen, 2013). For  
726 example, if genomic incompatibilities become trapped in environmentally divergent habitats,  
727 coupling with loci involved in local adaptation may occur, which could potentially facilitate  
728 diversification within the hybrid lineage (Abbott et al., 2013; Bierne, Gagnaire & David,  
729 2013; Butlin & Smadja, 2018; Seehausen, 2004). This coupling mechanism, more prone to  
730 arise in hybrid lineages around regions of interspecific incompatibilities, could facilitate rapid  
731 local adaptation in comparison to other processes of diversifying selection in non-hybrid  
732 species (Eroukmanoff et al., 2013; Seehausen, 2013). To the best of our knowledge, there  
733 are no empirical studies that report such linkage between DMIs and regions under natural  
734 selection. However, our results and previous studies (e.g. Runemark, et al., 2018a) show that  
735 genomic constraints play an important role in the formation of the admixed Italian sparrow  
736 genome.

737

738 Here, we present for the first time direct evidence for the role that environmental variation has  
739 in mediating genomic variation in a hybrid species, a phenomenon well described in non-

740 hybrid species (Wang & Bradburd, 2014). We also report loci where high levels of adaptive  
741 genetic differentiation has occurred, some of which are covarying directly with climate  
742 variation, suggesting that they are situated in genomic regions linked to local adaptation. For  
743 example on chromosome 20 an outlier locus for adaptive divergence between Italian sparrow  
744 populations (via Bayescan) was found to be in the vicinity of the GDF5 gene (growth  
745 differentiation factor 5, also known as BMP14 (NCBI), a gene also identified as a 5%  $F_{ST}$   
746 outlier in the per-gene analysis based on whole genome-resequencing data. This gene is  
747 known to be involved in jaw development in vertebrates (Bleuming et al., 2007; Cerny et al.,  
748 2010; Kaucka & Adameyko, 2019) and related to the BMP (bone morphogenic protein) gene  
749 family (Buxton, Edwards, Archer & Francis-West, 2001; Francis-West et al., 1999a; Francis-  
750 West, Philippa, Parish, Lee & Archer, 1999b). The BMP gene family has a fundamental role  
751 in craniofacial development and beak shape and size variation in Darwin's finches  
752 (Abzhanov, Protas, Grant, Grant & Tabin, 2004; Lamichhaney et al., 2016).

753

754 The beak is a trait known to be under strong selective pressure (Lamichhaney et al., 2016;  
755 Lamichhaney et al., 2015). Beak size has been shown to be a crucial trait underlying the  
756 survival of Darwin's finches after a drought (Lamichhaney et al., 2016) and beak traits in  
757 general act as drivers of major evolutionary shifts in Darwin's finches (Almén et al., 2016;  
758 Chaves et al., 2016; Lamichhaney et al., 2016; Lamichhaney et al., 2015). Beak shape  
759 variation has been found to respond to environmental divergence affecting food availability in  
760 the medium ground finch (*Geospiza fortis*) (Grant & Grant, 2003; Grant & Grant, 2014).  
761 Thus, climatic factors could be considered a reasonable proxy for food availability in  
762 sparrows (Runemark, et al., 2018b). It is possible that divergence of genes associated with  
763 beak morphology may reflect an adaptive response to variation in food resources found in



764 environmentally different habitats. However, further analyses need to be conducted in order to  
765 determine the true underlying mechanisms of divergence between population both at the  
766 genetic and phenotypic level.

767

### 768 **Hybrid constraints to population divergence**

769

770 Evaluating patterns of ancestry and divergence in the hybrid genome can provide important  
771 insights on whether population differentiation is facilitated by novel genetic variation or  
772 hampered by genomic constraints linked to hybrid incompatibilities. Genomic variation  
773 within a hybrid lineage can be generated by novel genetic combinations through  
774 rearrangements of parental blocks, potentially generating novel epistatic interactions, or  
775 through heterozygosity at parental divergent loci. In this case, highly differentiated loci within  
776 the hybrid taxon can be expected to be located in regions where the parental species have  
777 diverged strongly. On the other hand, negative epistatic interactions between inherited  
778 parental blocks (in particular if these interactions involve genetic incompatibilities) may lead  
779 to strong stabilizing selection on loci fixed for compatible alleles and, through linkage  
780 disequilibrium, on other loci situated in their vicinity. Thus, inherited parental genomic blocks  
781 would be expected to be highly conserved, as these are more likely to harbour candidate loci  
782 for genetic incompatibilities. This type of genetic constraint on hybrids could reduce the  
783 evolutionary potential of the hybrid species to diverge at the population level. However, this  
784 may depend on variation in recombination rate across the genome, which is also known to  
785 affect the extent of purging and population divergence.

786

787 We found that a large proportion of the hybrid genome presents a mosaic pattern where  
788 polymorphic sites seem to be generated either by the inheritance of differential parental  
789 alleles or standing genetic variation already present in the parents. We also report that genetic  
790 variation present in loci that are not divergent between parental species accounts for most of  
791 the high genomic differentiation found within the hybrid at the population level and that some  
792 of this variation may play a role in local adaptation. Furthermore, loci where the parent  
793 species are fixed for different alleles or have highly divergent allele frequencies seem to be  
794 preferentially fixed for one parental allele across Italian sparrow populations (also evidenced  
795 by Runemark et al., 2018a). This supports the hypothesis of constraints biasing evolution to  
796 loci that are not differentiated between the parental species and hence are less likely to be  
797 incompatible. Although we note in this case we have no direct evidence that such loci have  
798 any fitness effects on hybrids.

799

800 Despite the potentially constrained nature of the hybrid genome, the Italian sparrow has been  
801 able to diverge and locally adapt as a response to environmental variation. Our results on  
802 ancestry estimates suggest that inheritance of parental standing genetic variation is a plausible  
803 source of the genetic divergence found in the hybrid species. This variation could be neutral  
804 in the parental species, as it seems to not be involved in population divergence in either parent  
805 species. Additionally, genomic variation generated in the hybrid (i.e. private alleles) does not  
806 seem to contribute to population structure.

807

808 Interestingly, patterns of population divergence within the hybrid taxon and each of its  
809 parental species seem to differ, suggesting that the admixed nature of the hybrid species may  
810 be somewhat restricted compared to its non-hybrid parental species. In contrast to the hybrid

811 species, intraspecific genomic variation in the parental lineages is located mainly in regions of  
812 parental divergence. Additionally, there is no overlap of outlier loci under selection among  
813 the three species. This could suggest that differential selective pressures may be operating in  
814 addition to specific genomic constraints in the hybrid species. However, an important factor  
815 to be considered in admixed genomes is the inheritance of traces of different evolutionary  
816 histories as well as the individual evolutionary path that the hybrid species has taken since its  
817 formation (and eventual further introgression with parent species). Thus, processes other than  
818 differential selective pressures could generate this pattern.

819  
820 These results provide a new perspective on how hybridization may impact adaptive evolution,  
821 more specifically on how novel genomic variation evolves and is utilized in a hybrid lineage  
822 post hybrid speciation, not only through genomic rearrangements linked to admixture and  
823 incompatibilities.

824

## 825 **CONCLUSION**

826

827 Genetic variation within the Italian sparrow appears to be driven by climatic variation,  
828 temperature being the main factor; we find evidence for isolation by environment (IBE),  
829 which could facilitate ongoing local adaptation. Our study supports previous findings  
830 suggesting that local adaptation nonetheless can occur, albeit in a biased and constrained  
831 manner. Indeed, genetic differentiation in the hybrid species is mainly found in loci that are  
832 not divergent between the parental species and hence possibly less prone to be incompatible  
833 in the hybrid. This suggests that purging of incompatibilities could be an important element in  
834 the evolution of this species. Standing genetic variation inherited from the parental species is

835 a likely explanation for much of the genomic variation in the hybrid species, and some of the  
836 variation may be involved in subsequent local adaptation. In contrast, we find little or no  
837 evidence that novel variation (private alleles - new mutations occurring after HHS) has been  
838 important in local adaptation. Coupling of incompatibilities and loci under natural selection  
839 may also have facilitated the rapid genomic divergence observed in the Italian sparrow and its  
840 effect on gene flow. However, studies addressing these hypotheses directly are necessary to  
841 assess causality.

842

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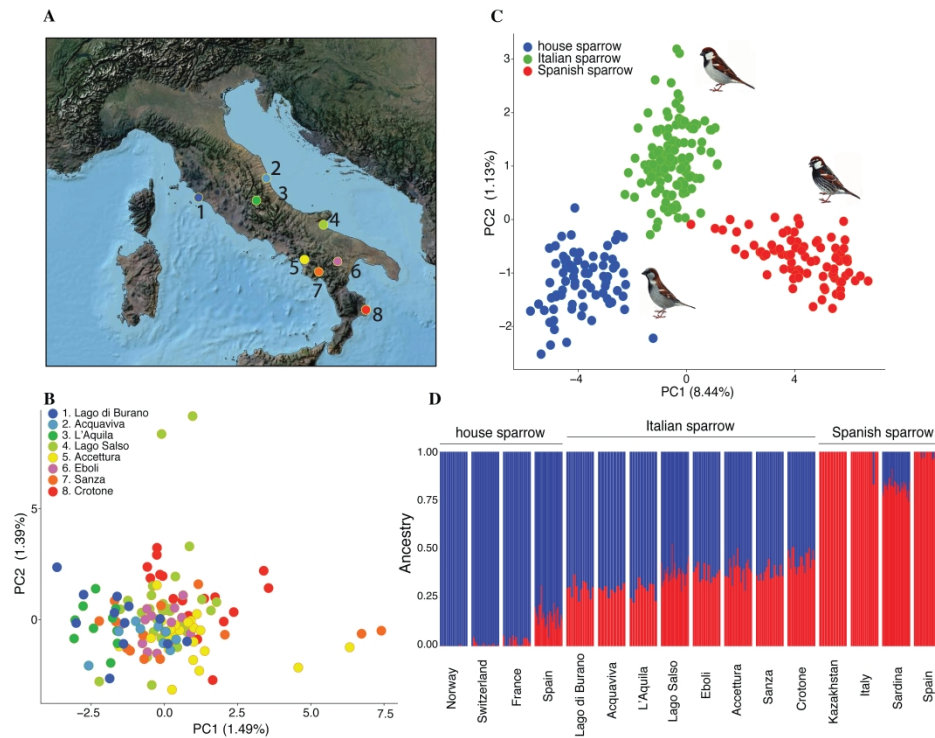
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1116 **DATA ACCESSIBILITY:** Genomic data produced in this study has been deposited at the  
1117 NCBI Sequence Read Archive under BioProject PRJNA680598 BioSample accessions  
1118 numbers SAMN16886216- SAMN16886520 (raw RADseq reads in fastq format). VCF files,  
1119 scripts to process the genomic data as well as scripts used for the statistical analysis and other  
1120 final dataset generated have been deposited in the Dryad Digital Repository at DOI  
1121 (doi:10.5061/dryad.q573n5th7).

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1123 **AUTHORS' CONTRIBUTIONS:** A.C. and F.E. designed the study; A.C., F.E., M.R.  
1124 analysed the data; A.C. conducted laboratory work; A.C., F.E. and G-P.S. collected field data;  
1125 A.C. wrote the manuscript. F.E., M.R. and G-P.S. contributed and commented on earlier  
1126 drafts of the manuscript.

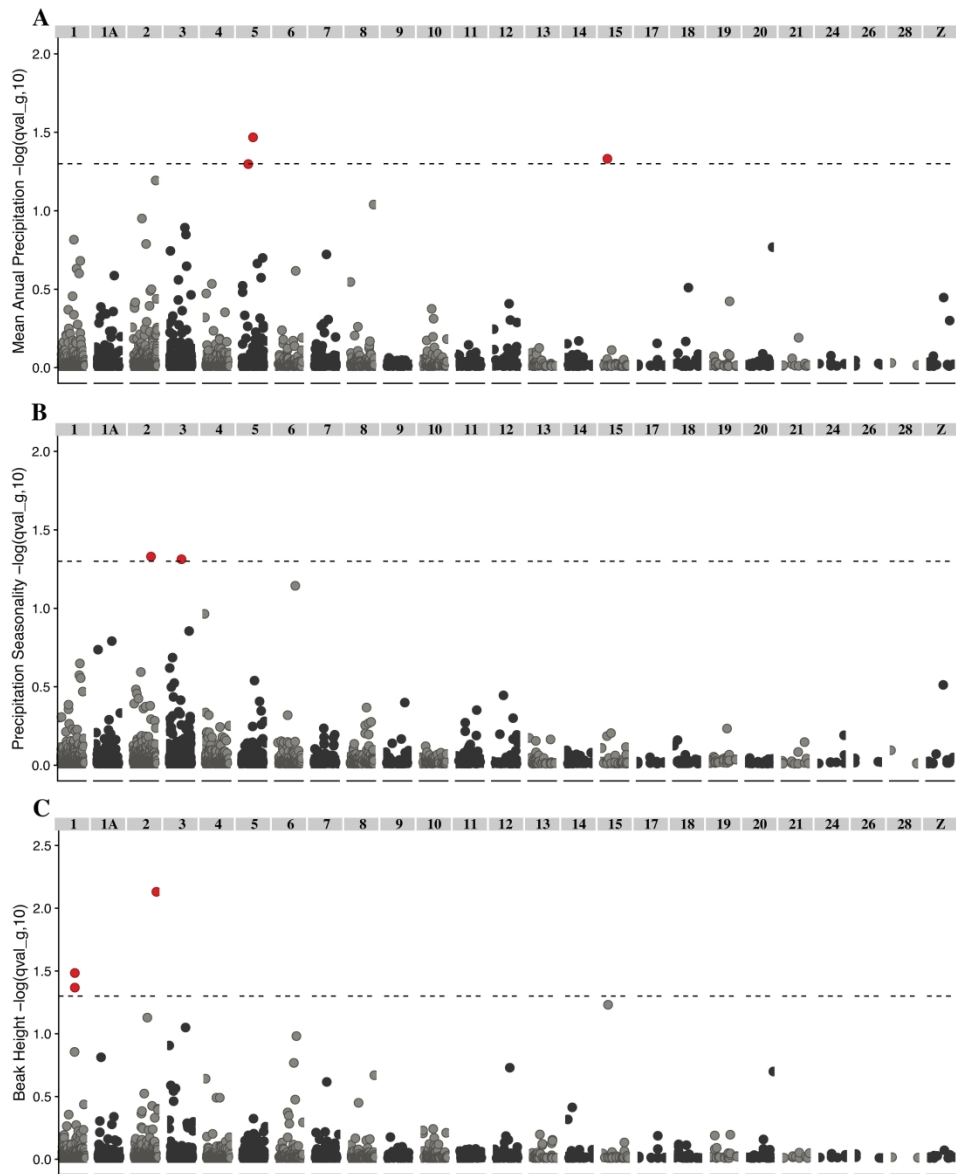
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1128 **COMPETING INTERESTS**

1129 The authors declare that they have no competing interests.

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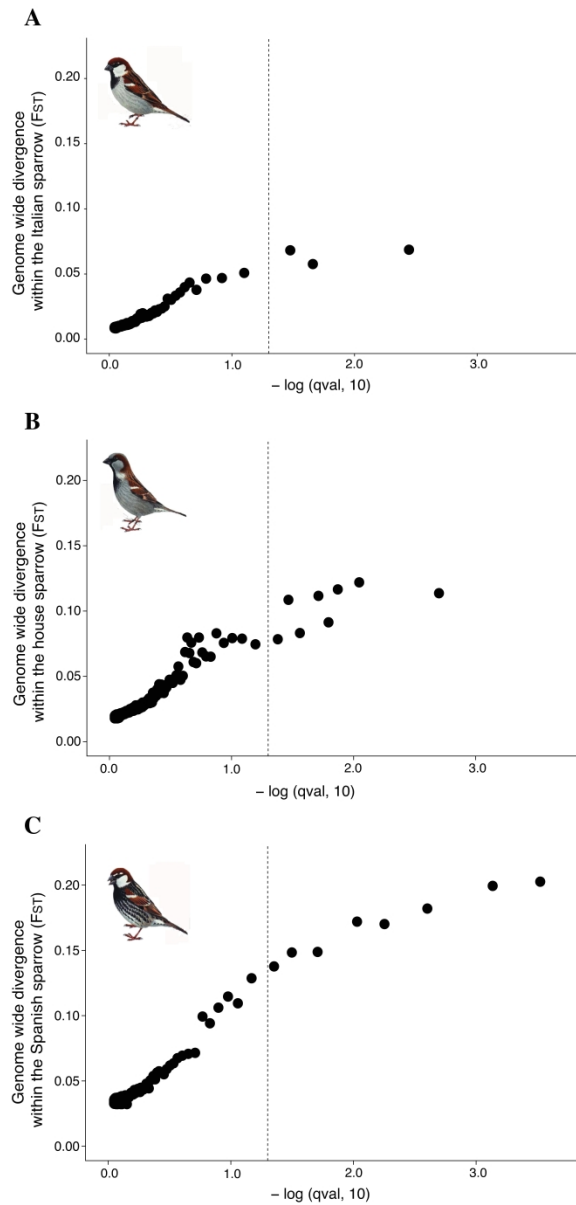


**Figure 1. A. Geographic distribution of sampled Italian sparrow populations B. Principal component analysis (PCA) to explore genetic divergence within the Italian sparrow (8 Italian populations, 131 individuals and 4387 SNPs). C. PCA assessing the three focal species. Spanish sparrow (red), house sparrow (blue) and Italian sparrow (green), and D. Admixture analysis based in a VCF file containing 288 individuals (131 Italian, 82 Spanish and 75 House sparrows) and 2737 high-quality SNPs. Localities are ordered following latitudinal distribution.**



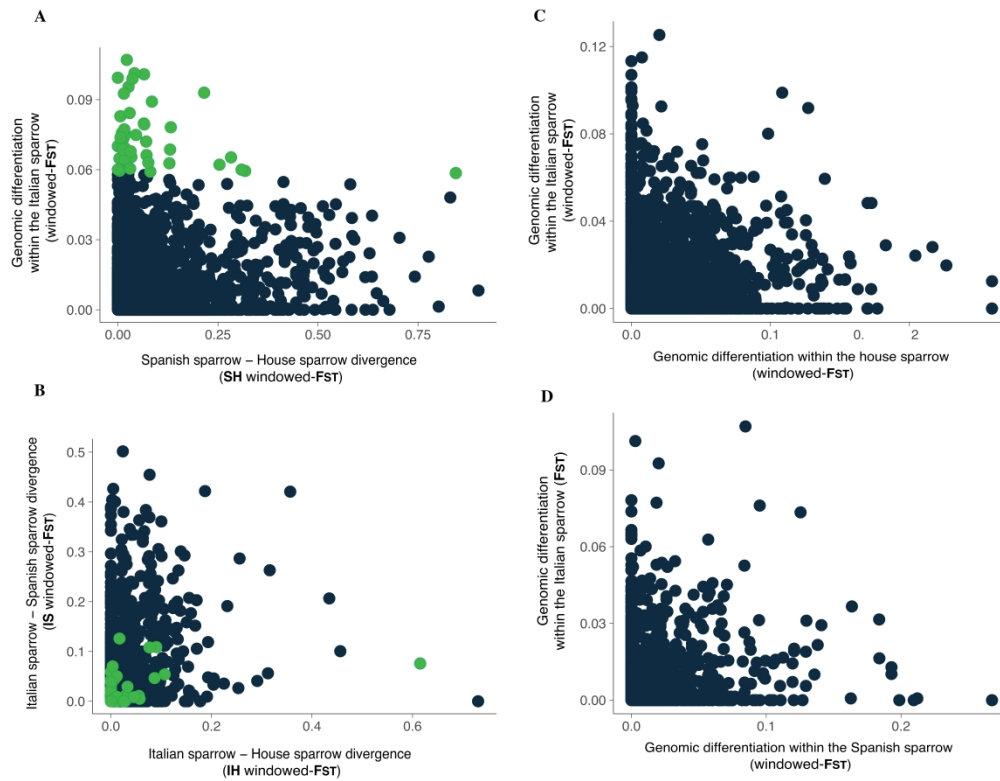
**Figure 2. Outlier analysis of local adaptation to climate (BayeScEnv).** Manhattan plots of correlation  $q$ -values for genetic divergence (SNPs) within the Italian sparrow showing association to climatic factors and one phenotypic trait. Significance level (FDR-corrected) is set at a  $q$ -value of  $< 0.05$  ( $-\log_{10} = 1.3$ ). **A.** Mean Annual Precipitation **B.** Precipitation seasonality and **C.** Beak height.

300x379mm (300 x 300 DPI)



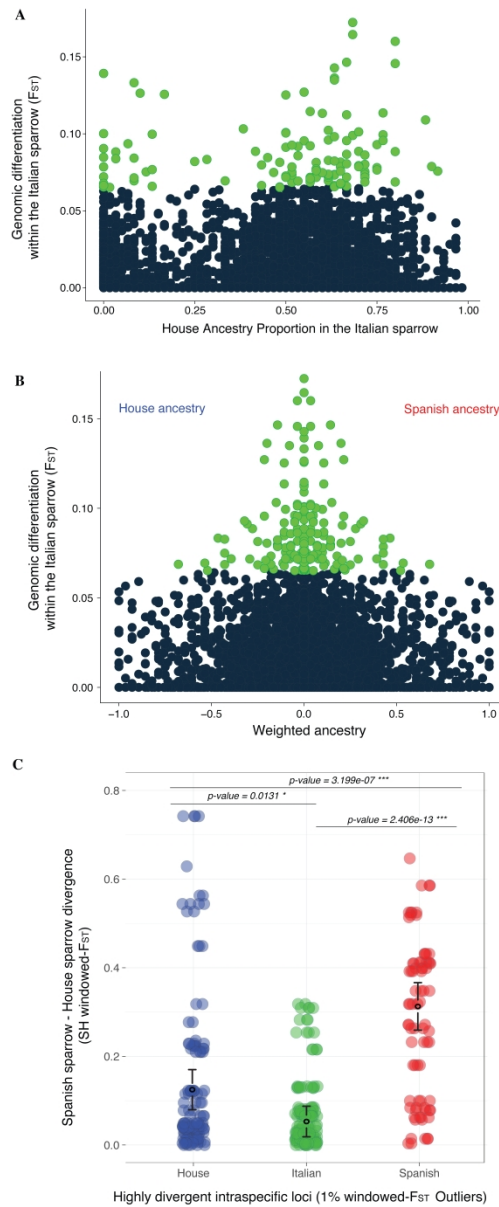
**Figure 3. Outlier analysis (BayeScan).** Correlation q-values for genetic divergence (SNPs). Significance level (FDR-corrected) is set at a q-value of < 0.05 ( $-\log_{10} = 1.3$ ). **A.** within the Italian sparrow **B.** within the house sparrow and **C.** within the Spanish sparrow.

351x731mm (600 x 600 DPI)



**Figure 4. Hybrid constraints to population divergence.** **A.** Genomic differentiation within the Italian sparrow (windowed-F<sub>ST</sub>) and divergence of its parental species (SH windowed-F<sub>ST</sub>) **B.** Genomic divergence of the Italian sparrow and each of its parental species (Italian - House sparrow divergence (IH windowed-F<sub>ST</sub>) and Italian - Spanish sparrow divergence (IS windowed-F<sub>ST</sub>)), with highlighted within-Italian-sparrow-F<sub>ST</sub> outliers in green. Genomic differentiation within the Italian sparrow v.s. genomic differentiation within each of the parental species, **C.** the house and **D.** the Spanish. sparrows.

831x644mm (600 x 600 DPI)



**Figure 5. A. Ancestry proportion v.s. Italian sparrow  $F_{ST}$ .** House ancestry proportion calculated across all Italian sparrow populations. **B. Weighted ancestry v.s. intraspecific  $F_{ST}$  in the Italian sparrow.**

Ancestry weighted by parental allele frequency difference. Green points represent the 1% Italian  $F_{ST}$  outliers. Negative values correspond to loci with house ancestry, while positive ones show Spanish ancestry.

**C. Parental genomic divergence ( $SH$  windowed- $F_{ST}$ )** presented on the intraspecific 1% windowed- $F_{ST}$  outlier loci from the three focal species (house sparrow  $F_{ST}$  outliers in blue, Italian sparrow in green and Spanish sparrow in red).

1 **Table1. Univariate matrix regression with randomization (UMRR).** 8 populations of the Italian sparrows. Pairwise  $F_{ST}$  between populations  
 2 as the response variable. Predictor variables are as following: Annual mean temperature (A.TEMP), Temperature seasonality (TEMP.S),  
 3 Altitude (ALT), Geographic distance (GEO), Annual mean precipitation (A.PREC), Precipitation seasonality (PREC.S), Beak height (BEAK.H)  
 4 and Beak length (BEAK.L)  
 5

	$R^2$	$\beta$	$t$	$p$ -value
TEMP.S	0.163	0.007	2.251	0.048 *
A.PREC	0.061	-0.004	-1.302	0.260
GEO	0.053	0.004	1.201	0.264
BEAK.L	0.036	0.003	0.991	0.358
BEAK.H	0.036	0.003	0.991	0.341
PREC.S	0.031	0.003	0.910	0.422
ALT	0.020	-0.002	-0.732	0.513
A.TEMP	0.001	-0.001	-0.172	0.864

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8 **Table 2. Multiple matrix regression with randomization (MMRR) and coefficients from Commonality Analysis (CA) – MODEL 1.**  
 9 Unique (U), common (C) and total (T) variance partitioning coefficients of each predictor variable to genomic divergence (Pairwise FST), in  
 10 parentheses the per cent contribution of the predictor to the total variance explained by the model (100 \* partition coefficient (U, C or T) / R2).  
 11 Pairwise FST between 8 populations of the Italian sparrow as the response variable. Predictor variables are the following: Annual mean  
 12 temperature (A.TEMP), Temperature seasonality (TEMP.S), Altitude (ALT), Geographic distance (GEO), Annual mean precipitation (A.PREC),  
 13 Precipitation seasonality (PREC.S).  
 14

15 **MODEL 1: Fst ~ GEO + A.TEMP + A.PREC + TEMP.S + PREC.S + ALT** **R<sup>2</sup> = 0.25**

Predictor	$\beta$	<i>t</i>	<i>p-value</i>	Unique (U)	Common (C)	Total (T)
GEO	0.003	0.93	0.34	0.03 (12%)	0.02 (8%)	0.05 (20%)
A.TEMP	0.001	0.15	0.89	0.14 (56%)	0.03 (12%)	0.16 (64%)
A.PREC	-0.004	-1.16	0.32	0.05 (20%)	0.01 (4%)	0.06 (24%)
TEMP.S	0.007	1.96	0.05 *	0.003 (0%)	0.02 (8%)	0.02 (8%)
PREC.S	-0.003	-0.60	0.57	0.001 (0%)	0.00 (0%)	0.00 (0%)
ALT	-0.002	-0.29	0.78	0.01 (4%)	0.02 (8%)	0.03 (12%)

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22 **Table 3. Logistic Regression on the probability to be an Italian sparrow  $F_{ST}$  Outlier.** Top 1% intraspecific  $F_{ST}$  outlier loci selected from a  
 23 vcf file including the three focal species (131 Italian, 82 Spanish and 75 house sparrows). Outlier loci were identified in a dataset of 2737 shared  
 24 SNPs between the three species. Outlier status (Italian  $F_{ST}$  outlier) used as response variable.  $F_{ST}$  Outlier threshold=0.06275, Genomic  
 25 divergence between parental species (Spanish – House (SH  $F_{ST}$ )) and between the hybrid lineage and each of its parents (Italian – House (IH  
 26  $F_{ST}$ ), Italian – Spanish (IS  $F_{ST}$ )), additive and interaction effects, are used as predictors.  
 27

Model	Predictor	Parameter Estimate	Std. Error	p-value	
<b>Italian <math>F_{ST}</math> Outlier</b>	~ SH $F_{ST}$	SH $F_{ST}$	-2.1391	1.2719	0.0926
	~ IH $F_{ST}$ + IS $F_{ST}$	IH $F_{ST}$	1.9886	2.4321	0.4135
IS $F_{ST}$		-7.6170	3.0571	0.0127 *	
~ IH $F_{ST}$ * IS $F_{ST}$	IH $F_{ST}$	3.2764	2.6091	0.2092	
	IS $F_{ST}$	-5.9250	3.5668	0.0967	
	IH $F_{ST}$ : IS $F_{ST}$	-49.4795	66.2468	0.4551	

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