

1 **Micro-evolutionary response of spring migration timing in a wild seabird**

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13

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15 Secondary Theorem of Selection

16 **ABSTRACT**

17 Understanding the mechanisms by which populations can adapt to changing environmental
18 conditions is crucial for predicting their viability. In the context of rapid climate change,
19 phenological advance is a key adaptation for which evidence is accumulating across taxa. Among
20 vertebrates, phenotypic plasticity is known to underlie most of this phenological change, while
21 evidence for micro-evolution is very limited and challenging to infer. In this study, we quantified
22 phenotypic and genetic trends in timing of spring migration using 8032 dates of arrival at the
23 breeding grounds obtained from observations on 1715 individual common terns (*Sterna hirundo*)
24 monitored across 27 years, and tested whether these trends were consistent with predictions of a
25 micro-evolutionary response to selection. We observed a strong phenotypic advance in arrival
26 date, with birds arriving on average 9.34 days earlier over the study period. This phenotypic trend
27 translated into an advance in breeding values, which accounted for c. 7.4 % of the observed change
28 in the population. The Breeder's equation and Robertson's Secondary Theorem of Selection
29 predicted qualitatively similar evolutionary responses to selection, although the estimate from the
30 latter was uncertain, and those theoretical predictions were largely consistent with observed
31 genetic patterns. Overall, our study therefore provides rare evidence for micro-evolution to
32 underlie (part of) an adaptive response to climate change in the wild, and illustrates how a
33 combination of adaptive micro-evolution and phenotypic plasticity facilitated a shift towards
34 earlier migration in this natural population of common terns.

35 **IMPACT STATEMENT**

36 Empirical evidence for evolutionary change underlying vertebrate adaptation to current global
37 change is very rare. This may be due to phenotypic plasticity being the main mechanism underlying
38 adaptation, or to challenges associated with the empirical testing of genetic changes in the wild,
39 i.e., to data limitations. In this study, we tested whether an observed phenotypic advancement in
40 the timing of spring migration in a wild seabird population was due to an evolutionary response
41 (i.e., genetic change), and/or to phenotypic plasticity or other environmental (nongenetic) effects.
42 To do so, we applied a series of “animal models” to a long-term dataset from a pedigreed
43 population of common terns located at the North Sea coast of Germany and monitored over 27
44 years. We tested whether timing of migration underwent a genetic change towards earlier arrival
45 of birds on the breeding grounds, in response to increasing temperatures at the wintering grounds
46 in West Africa. We found an evolutionary response to selection favouring earlier arriving
47 individuals. Additionally, we showed that two different theoretical approaches predicted a
48 qualitatively similar evolutionary response, both in terms of direction and magnitude. As such, our
49 study provides a rare empirical case where estimated and predicted evolutionary responses are in
50 agreement and suggest an evolutionary response in the timing of avian spring migration. Our
51 results further demonstrated that adaptive evolution likely played a smaller role than phenotypic
52 plasticity, and overall, they highlight the importance of disentangling the relative contributions of
53 plastic and evolutionary changes to better understand adaptive processes and predict responses to
54 future changes.

55 **INTRODUCTION**

56 Phenotypic changes in response to climate change are common place across taxa (Parmesan and
57 Yohe 2003; Parmesan 2006; Hendry et al. 2008; Chen et al. 2011; Bellard et al. 2012; Radchuk et
58 al. 2019), but the mechanism(s) underlying these changes remain largely unknown (Lavergne et
59 al. 2010; Salamin et al. 2010; Hoffmann and Sgrò 2011; Gienapp and Brommer 2014; Merilä and
60 Hendry 2014). Where the mechanisms underlying the observed shifts have been explicitly
61 investigated, they have often been attributed in vertebrates to phenotypic plasticity (i.e., genotypes
62 express variable phenotypes under different environmental conditions), while evidence for micro-
63 evolution (i.e., genotypes that have a higher fitness increase in frequency in the population) is still
64 relatively rare (Hairston et al. 2005; Gienapp et al. 2008; Visser 2008; Hoffmann and Sgrò 2011;
65 Vedder et al. 2013; Gienapp and Brommer 2014; Merilä and Hendry 2014).

66 Animal models (*sensu* Kruuk 2004) are useful statistical tools to test for genetic trends and
67 rates of micro-evolutionary responses to selection by means of providing estimates of individual
68 breeding values (i.e., the expected effect of the genes that an individual passes on to its offspring,
69 Falconer and Mackay 1998; Lynch and Walsh 1998). The strength of this test, however, relies on
70 properly accounting for uncertainty in breeding value predictions and environmental factors
71 (Postma 2006; Hadfield et al. 2010). To our knowledge, there are only three long-term studies
72 from natural vertebrate populations that have found evidence for a genetic change underlying
73 phenotypic change (most likely) in response to climate change, while taking into account the
74 uncertainty in the breeding value estimation and while controlling for a temporal trend in the
75 phenotypic data (but see Merilä et al. 2001 and Walsh and Lynch 2018 for reviews on evolutionary
76 stasis, where no micro-evolutionary change was observed). First, based on ten years of data on an
77 alpine population of snow voles (*Chionomys nivalis*), Bonnet and colleagues (2017) showed a

78 genetic change towards a reduced body mass as an adaptive response to viability selection, likely
79 in response to a change in snowfall patterns. This genetic change represented a case of “cryptic
80 evolution” where the population did evolve but there was no observed phenotypic change in body
81 mass. Second, Evans and Gustafsson (2017) studied male plumage colouration in collared
82 flycatchers (*Ficedula albicollis*) over 34 years and showed a marked phenotypic decline, which
83 was mirrored by a decline in the mean breeding value of these males. This decline, which
84 accounted for 11% of the total phenotypic change, was likely driven by viability selection acting
85 on ornamentation being sensitive to the climatic conditions experienced at the breeding ground in
86 the preceding year. And third, using data obtained from a wild population of red deer (*Cervus*
87 *elaphus*), Bonnet and colleagues (2019) found that average parturition date had advanced by nearly
88 two weeks over the 45-year study period, with warmer temperatures during the previous rut season
89 tending to advance parturition date, and showed that a micro-evolutionary response accounted for
90 15% of the total change of -12.3 days (Bonnet et al. 2019).

91 Once genetic trends have been estimated, testing their consistency with predictions of a
92 micro-evolutionary response to selection can inform us about the occurrence and magnitude of
93 adaptive responses to selection, as well as our ability to model these responses. Traditionally, the
94 per-generation evolutionary response to selection in a given trait, R , can be predicted by $R = h^2S$,
95 also known as the Breeder’s equation, where h^2 represents the trait heritability, and S is the
96 directional selection differential (Falconer and Mackay 1998; Lynch and Walsh 1998). This
97 Breeder’s equation, however, tends to lead to biased predictions in studies from natural populations
98 due to the occurrence of correlated selection as well as confounding environmental effects (i.e.,
99 when selection is acting on the environmental rather than genetic component of a trait, the
100 Breeder’s equation will overestimate the trait’s response to selection) (Morrissey et al. 2010). It

101 therefore has been argued that micro-evolutionary responses to selection in natural populations are
102 better predicted by applying Robertson's Secondary Theorem of natural Selection (hereafter, STS)
103 (Morrissey et al. 2010), which postulates that the additive genetic covariance between a trait and
104 relative fitness is a direct measure of the expected per-generation evolutionary change in the mean
105 trait value, R (Robertson 1966; Price 1970). While STS is a strictly genetic approach to predicting
106 a response to selection, and therefore unbiased by unmeasured covariates, it is also agnostic
107 regarding the selection processes underlying the micro-evolutionary responses. Moreover, the
108 predictive power of STS can be hampered in the case of strong genetic drift or selection acting on
109 a genetically correlated trait (Morrissey et al. 2010, 2012a; Walsh and Lynch 2018). Hence, a
110 comprehensive understanding of the predicted evolutionary responses to selection will benefit
111 from evaluating the predictions from both the Breeder's equation and the Robertson's Secondary
112 Theorem of Selection, as both quantitative genetic models will provide complementary
113 information.

114 Along with poleward shifts in geographic distributions, phenotypic changes in response to
115 global warming are most apparent in phenology (Root et al. 2003; Menzel et al. 2006; Parmesan
116 2006; Thackeray et al. 2010; Poloczanska et al. 2013). As such, timing of migration, together with
117 timing of breeding, is a trait of special interest in the context of current global and climate change
118 (Parmesan and Yohe 2003; Charmantier and Gienapp 2014; Gienapp et al. 2014). Among birds,
119 most studied populations have shifted their phenology to earlier migration and breeding (review:
120 Lehikoinen et al. 2004; Gienapp et al. 2007; Gordo 2007; Charmantier and Gienapp 2014; meta-
121 analysis: Radchuk et al. 2019), although not all bird species share the same adaptive potential (e.g.,
122 Keogan et al. 2018). For instance, some birds such as long-distance migrants might be constrained
123 in their potential to keep track of changing environmental conditions by inherited circannual clocks

124 and migratory behaviours (Åkesson et al. 2017) and by reduced correlation between climate on the
125 wintering and the breeding grounds, making them more sensitive to rapid warming conditions.
126 With respect to the timing of migration, the date of arrival at the breeding grounds has received
127 most empirical attention, and an early arrival is expected to enable selection of high-quality
128 breeding sites (Kokko et al. 2004) and mates (Coppack and Both 2002; Ludwigs and Becker 2005),
129 and to favour an early initiation of breeding, which in turn provides offspring with additional
130 developmental time before autumn migration. Besides often being under strong directional
131 selection (favouring earlier arrival to the breeding sites), arrival date has also been found to harbour
132 an important genetic component. For instance, arrival date heritabilities ranged from 0.11 to 0.32
133 in three different studies from bird populations: barn swallows (*Hirundo rustica*, Teplitsky et al.
134 2011), common terns (*Sterna hirundo*, Arnaud et al. 2013), and great reed warbler (*Acrocephalus*
135 *arundinaceus*, Tarka et al. 2015), with collective evidence across four bird species revealing a
136 mean heritability of 0.43 (Pulido 2007).

137 Even though the prerequisites for a micro-evolutionary response are often met (non-zero
138 additive genetic variance and selection), to date we know of only two studies providing “indirect”
139 evidence for a micro-evolutionary response of the timing of migration in natural bird populations.
140 First, Teplitsky and colleagues (2011) used two long-term datasets on barn swallows from Spain
141 and Denmark to show that, for the Spanish population, the Breeder’s equation and Robertson’s
142 Secondary Theorem of Selection predicted qualitatively similar responses to selection (towards
143 earlier arrival dates). Teplitsky and colleagues (2011) did not, however, further estimate whether
144 their study population had undergone substantial phenotypic or genetic change across years,
145 precluding the comparison between observed and predicted responses to selection in timing of
146 migration. Second, Helm and colleagues (2019) replicated an experimental study of the annual

147 cycle of a long-distance migratory species, the pied flycatcher (*Ficedula hypoleuca*), after 21 years
148 of warming to show that migration timing had advanced by 9 days. This advance was also observed
149 in a nearby natural population of flycatchers, such that, altogether, their results supported a role of
150 micro-evolution in changing spring migration timing (Helm et al. 2019).

151 In this study, we applied a series of univariate and bivariate animal models to test for the
152 existence and magnitude of a micro-evolutionary response in the timing of spring arrival in a long-
153 distance migratory bird. We used 27 years of data from a natural, pedigreed population of common
154 terns (*Sterna hirundo*). We had two main objectives: (1) to quantify whether there was a detectable
155 genetic change in arrival date (i.e., a temporal trend in individual breeding values), and (2) to test
156 whether this trend was consistent with predictions of micro-evolutionary responses to selection
157 based on both the Breeder's equation and Robertson's Secondary Theorem of natural Selection.
158 Previous studies from the same study population found that arrival date was heritable (Arnaud et
159 al. 2013; Moiron et al. 2020) and under directional selection, with earlier arrival being associated
160 with a higher annual number of fledglings (Arnaud et al. 2013, but see Ezard et al. 2007) and a
161 higher probability of survival (Zhang et al. 2015b). Hence, although we acknowledge that
162 evolutionary stasis often occurs, even when these prerequisites are met (Merilä et al. 2001; Walsh
163 and Lynch 2018), we expected a micro-evolutionary response towards earlier arrival dates from
164 the wintering grounds. Finally, earlier work found evidence for individual, and, to a lesser extent,
165 additive genetic variance in plasticity of arrival date in response to sea surface temperature at the
166 main wintering areas (Moiron et al. *in revision*, see Dobson et al. 2017 for analyses on breeding
167 timing), highlighting the potential for multiple, non-mutually exclusive adaptive responses to be
168 in action in this population of common terns.

169

170 **METHODS**

171 **STUDY SYSTEM AND DATA COLLECTION**

172 The data were collected as part of a long-term study of a common tern population located at the
173 Banter See on the German North Sea coast (53°36'N, 08°06'E). The colony consists of six concrete
174 islands, each of which is surrounded by a 60 cm wall. An individual-based study was initiated in
175 1992, when 101 adult birds were caught and marked with individually numbered subcutaneously
176 injected transponders. Since 1992, all locally hatched birds have similarly been marked with a
177 transponder shortly prior to fledging and the presence and reproductive performance of marked
178 individuals has been monitored with the help of antennae and following a standard protocol
179 (Becker and Wendeln 1997). Thanks to the automatic antenna system, affixed to the walls of the
180 colony site since 1994, it is possible to record the arrival date from the wintering grounds of every
181 marked bird in the population. From this, and confirmed by a tracking study, it is known that the
182 common terns of this population show highly repeatable migratory behaviour (Kürten et al. 2022).

183 As part of the standard protocol, breeding birds are identified using portable antennae
184 placed around each nest for 1–2 days during incubation, which is shared by both partners. Pairs
185 can rear up to three chicks per brood and produce replacement clutches after loss of eggs or chicks.
186 True second clutches are extremely rare (Becker and Zhang 2011). Chicks are ringed at hatching
187 and checked every 2–3 days throughout the breeding season until they fledge (at about 26 days;
188 Becker and Wink 2003) or perish.

189

190 **DATA SELECTION**

191 The phenotypic and fitness data used in this study were collected between 1994 and 2020. We
192 focused on arrival date at the breeding area from the wintering grounds in West Africa (Becker et

193 al. 2016; Kürten et al. 2022), a phenological trait that captures variation in the timing of spring
194 migration and that is defined as the day of first return to the breeding grounds (January 1st =1)
195 (Zhang et al. 2015a). Because no breeding can occur before a bird has arrived at the colony and
196 recovered from migration, we removed those observations of arrival date that were not at least 10
197 days earlier than the egg laying date (n = 696 observations from 390 birds). The resulting dataset
198 included 8032 observations of arrival date from 1715 individuals of known sex and age (mean
199 number of observations per individual = 4.68, range = 1–21). The overall mean (\pm SD) arrival date
200 across the 27 years (1994 - 2020) included in our dataset was 118 ± 12.63 (~April 28).

201 We used Adult Lifetime Reproductive Success (adult LRS) as our fitness measure, and
202 quantified it as the total number of fledglings locally produced in the lifetime of an adult individual
203 (Fig. S1). Although we cannot directly observe an individual's death, we can reliably assume it,
204 because breeders at the Banter See are highly site-faithful, evidenced by the resighting probability
205 of adult individuals that bred at least once being close to one (Szostek and Becker 2012), and 96%
206 of breeders not skipping recording by the antenna system for two or more consecutive years after
207 first reproduction (Bouwhuis et al. 2015; Zhang et al. 2015a). Based on this knowledge, we
208 removed birds that were observed in 2019 and/or 2020 because they are known to not be, or cannot
209 yet be assumed to be, dead. In addition to using data from individuals assumed to be dead (n = 946
210 individuals), however, we also included fitness data from those individuals that were still alive but
211 older than 10 years (n = 121) to avoid “cohort truncation” (i.e., to avoid excluding individuals “not
212 at random” with respect to fitness, Hadfield 2008; Morrissey et al. 2012). We chose this threshold
213 because the cumulative reproductive success of known dead birds at ages older than 10 and their
214 lifetime reproductive success is known to be highly correlated ($r > 0.8$, Moiron et al. 2021). This
215 age threshold also matches the biology of the population, as the mean lifespan is ~10 years

216 (Szostek and Becker 2012). The dataset for the analyses with adult LRS consisted of 7000
217 observations from 1067 individuals of known age and sex. Relevant details of the social pedigree
218 used in the study are in the Supplementary Material.

219

220 **STATISTICAL ANALYSES**

221 **Phenotypic trend and sources of phenotypic variation in arrival date**

222 We used annual mean arrival dates to test for a temporal (inter-annual) trend in phenology between
223 1994 and 2020. The model was run as a linear model with arrival date as a response variable
224 modelled assuming a Gaussian error distribution and year as a continuous fixed effect.

225 To estimate sources of phenotypic variation in arrival date and test for a temporal trend in
226 breeding values, we built a univariate animal model that decomposed the total phenotypic variance
227 of arrival date (V_P) into additive genetic (V_A), permanent environmental (V_{PE}), year of breeding
228 (V_{YR}), year of hatching (V_{COHORT}) and residual (V_R) variances. We fitted arrival date as the
229 response variable, and modelled it assuming a Gaussian error distribution. We fitted random
230 intercepts for year of breeding (to control for temporal variation over and above the linear trend
231 modelled in the fixed effect structure), year of hatching (to control for differences among cohorts),
232 individual identity linked to the relatedness matrix (to quantify additive genetic effects), and
233 individual identity not linked to the relatedness matrix (to control for repeated measures of
234 individuals across years and model permanent environmental effects). We did not fit parental or
235 maternal identity as random effects because 61.22% of individuals have an unknown mother, and
236 62.27% an unknown father, such that doing so would reduce the dataset (see Supplementary
237 Material for detailed information on the pedigree structure). As fixed effects, we fitted the linear
238 and quadratic effects of age (continuous variable measured in years, Ezard et al. 2007; Zhang et

239 al. 2015), sex (categorical variable), the linear effect of year of breeding (continuous variable, to
240 control for the linear temporal trend in phenology), and sea surface temperature at the main
241 wintering area (Kürten et al. 2022) (continuous variable) both as a main effect and in interaction
242 with the linear and quadratic effects of age. We used data of sea surface temperature between June
243 23 and July 27 the year prior to arrival at the coast of Guinea and Sierra Leone, determined based
244 on the approach developed by Haest and colleagues (2018, 2019, 2020) (Bouwhuis et al. *in prep*).
245 All continuous fixed effects (except age) were mean centred and variance standardized. We
246 estimated the heritability (h^2) conditional to the variance explained by fixed effects as the
247 proportion of the total phenotypic variance explained by the additive genetic variance.

248

249 **Testing for observed micro-evolutionary change and genetic drift**

250 We used the univariate animal model described above to extract the best linear unbiased predictors
251 (BLUPs) of breeding values for each individual and test for a temporal change in breeding values
252 in arrival date. To avoid biasing the estimate of the temporal trend in breeding values towards the
253 trend in the phenotypes (due to the environmental structure not being explicitly modelled, Postma
254 2006), we fitted year of breeding both as a random and fixed effect in our univariate animal model
255 following Bonnet and colleagues (2019), rather than as either a fixed effect as *per* Hadfield and
256 colleagues (2010), or a random effect as *per* Postma (2006). Fitting year of breeding as a covariate
257 in the univariate animal model used to obtain BLUPs is, however, a conservative approach because
258 it likely underestimates the rate of change in breeding values (Postma 2006; Hadfield et al. 2010).
259 Hence, we also fitted a univariate animal model with the same fixed and random effect structure
260 as detailed above but without year of breeding (linear and quadratic components) as a fixed effect

261 and extracted the BLUPs to test for a temporal change in breeding values in arrival date (see Table
262 S1).

263 To calculate the probability that the observed change in breeding values differed from a
264 scenario resulting from genetic drift, we simulated random breeding values for arrival date as *per*
265 Hadfield et al. (2010), and fitted a linear regression to these random breeding values to obtain the
266 temporal slopes due to drift for each posterior sample. Because the distribution of temporal slopes
267 due to random drift is expected to be centred at zero (sampling was neutral), we could then
268 calculate the proportion of the posterior distribution of these drift slopes that returned a regression
269 value more negative than the posterior mode of the observed temporal slopes.

270 Finally, the annual estimate of a micro-evolutionary response was converted to a per-
271 generation rate by multiplying by the population's generation time of ~7.95 years. We estimated
272 this generation time for common terns in our dataset as the mean age of parents of nestlings that
273 later recruit in the population (Charlesworth 1994), eight years being in line with the generation
274 time known for common terns (Nisbet et al. 2020). We also converted the annual estimate of micro-
275 evolution to units of standard deviation per generation by multiplying by generation time and
276 dividing by phenotypic standard deviations. This way, the unit for evolutionary change was
277 equivalent to a change measured in Haldanes (Hendry and Kinnison 1999).

278

279 **Predicting micro-evolutionary change**

280 To test for a micro-evolutionary response in arrival date, we fitted a bivariate animal model with
281 arrival date and adult LRS as response variables to estimate (1) the additive genetic covariance
282 between the two traits (direct test of Robertson's Secondary Theorem of Selection), and (2) the
283 individual covariance between adult LRS and arrival date (i.e., selection differential). Arrival date

284 was modelled assuming a Gaussian error distribution, while adult LRS was modelled assuming an
285 overdispersed Poisson error distribution with log-link function (Fig. S1). Because our fitness
286 measure follows a log-normal distribution, variance estimates for absolute fitness on the latent
287 scale data are equivalent to variance estimates directly on the data scale for relative fitness (de
288 Villemereuil et al. 2016; Bonnet et al. 2019; Morrissey and Bonnet 2019). The bivariate animal
289 model included random intercepts for individual identity linked to the relatedness matrix (V_A), for
290 individual identity not linked to the pedigree to account for repeated measures (V_{PE}), and for
291 among-cohort variation (V_{COHORT}) within an unstructured variance–covariance matrix allowing
292 the estimation of additive genetic, permanent environmental and cohort covariances (COV_A ,
293 COV_{PE} and COV_{COHORT} , respectively). We also fitted year of breeding as a random effect
294 associated only with arrival date. Additionally, adult LRS is only measured once per individual,
295 unlike the repeated annual measures on arrival date, and hence, V_{PE} and V_R in adult LRS are not
296 separable. To accommodate this difference in replication between the two traits, we fixed the
297 residual variance of LRS to a very small value (i.e., 0.001), allowing us to effectively model the
298 additive genetic and permanent environmental covariances between LRS and arrival date. We
299 modelled the same fixed effect structure as described for the univariate animal model.

300 We predicted the expected per-generation rate of micro-evolutionary change of arrival date
301 by first applying the univariate Breeder’s Equation (i.e., $R = h^2S$). The selection differential or
302 phenotypic covariance between focal trait and fitness, S , reported in the main text was calculated
303 as the sum of the additive genetic and permanent environmental covariances between arrival date
304 and adult LRS (but see Table S2 for a direct estimate of individual-level covariance between arrival
305 date and adult LRS, as opposed to the sum of the two covariance components). To obtain the full
306 posterior distribution for the evolutionary response to selection (R) and hence account for the

307 uncertainty in all estimated parameters, we multiplied the associated posterior distributions of h^2
308 and S instead of the point estimates. Second, Robertson's Secondary Theorem of natural Selection
309 states that the additive genetic covariance between a trait and relative fitness represents a direct
310 measure of the expected per-generation evolutionary change R in that trait (Robertson 1966; Price
311 1970). Both estimates of evolutionary response to selection (R) were reported as the posterior mode
312 and 95% Credible Intervals.

313

314 **Statistical model implementation**

315 We fitted all models using a Bayesian framework implemented in the statistical software R (v.
316 4.0.5, R Core Team 2021) using the R-package *MCMCglmm* (Hadfield 2010). We fitted
317 parameter-expanded priors (see Text S2). The number of iterations and thinning intervals for each
318 model were chosen so as to ensure that the minimum MCMC effective sample size for all
319 parameters was 1000. Burn-in was set to a minimum of 2000 iterations. The retained effective
320 sample sizes yielded absolute autocorrelation values lower than 0.1 and satisfied convergence
321 criteria based on the Heidelberger and Welch convergence diagnostic (Heidelberger and Welch
322 1981). We drew inferences from the posterior mode and 95% Credible Intervals (95% CI).

323

324 **RESULTS**

325 **Temporal trend in arrival date**

326 Between 1994 and 2020, annual mean arrival date advanced at a rate of 0.36 days per year (overall
327 change: -9.34 days, 95% CI: -10.43, -8.24; Fig. 1A), as did the individual's mean arrival date
328 regressed over their mean breeding year, with a rate of 0.09 days per year (overall change: -2.45
329 days, 95% CI: -4.51, -0.38).

330

331 **Sources of phenotypic variation in arrival date**

332 The mean population age increased steadily over the first 19 years of study, with a flattening-out
333 phase in the latter years (Fig. S2), and the linear and quadratic effect of age influenced the
334 population's arrival date (Table 1). Arrival date was also influenced by the sex of individuals, as
335 well as by sea surface temperature at the main wintering area and year of breeding in interaction
336 with age (Table 1).

337 Interpreting the variance components, common terns showed strong heritable differences
338 in arrival date: additive genetic effects accounted for 20.6% of the total phenotypic variance, while
339 permanent environmental effects accounted for an additional 10.6% (Table 1). Arrival date also
340 harboured important variance among years of breeding (7.6%) and cohorts (0.5%) (Table 1).

341

342 **Testing for observed micro-evolutionary change and genetic drift**

343 The best linear unbiased predictors (BLUPs) for breeding values of arrival date extracted from the
344 univariate animal model described above were tested for a trend over time (i.e., over an
345 individual's mean breeding year). The slope of the linear regression was estimated at -0.026 days
346 per year of breeding (95% CI: -0.062, 0.005), indicating that breeding values advanced a total of
347 ~0.687 days across the 27-year study (Fig. 1B). While the lower 95% CI of the annual estimate
348 was slightly overlapping zero, the fraction of the posterior distribution of the genetic slopes that
349 was greater than zero was only 3.3%. Re-expressed in units of phenotypic standard deviations per
350 generation, the estimated rate of micro-evolution corresponded to an observed evolutionary change
351 of -0.017 Haldanes (95% CI: -0.040, 0.003), equivalent to -0.211 days per generation (95% CI: -
352 0.500, 0.036).

353 Using a potentially less conservative approach to estimate breeding values for arrival date,
354 i.e., using BLUPs from a univariate animal model that did not include year of breeding as a fixed
355 effect (Table S1), the observed rate of evolution was of -0.035 days per year (95% CI: -0.072,
356 0.000). This estimate was of similar magnitude as that obtained using the more conservative
357 approach (Table 1). Finally, only 4.5% of the simulations of random drift generated an advance as
358 large as, or larger than, the change estimated from the linear regression of genetic change (Fig.
359 S3).

360

361 **Predicting micro-evolutionary change**

362 The phenotypic selection differential for arrival date quantified as the sum of additive genetic and
363 permanent environmental covariances from our bivariate animal model was negative and the
364 associated 95% credible intervals did not overlap zero ($S_{bivariate\ animal\ model} = -1.232$, 95% CI: -1.592,
365 -0.788), indicating that individuals with earlier arrival dates obtained, on average, higher adult
366 Lifetime Reproductive Success. We also estimated the selection differential for arrival date
367 directly as the individual-level covariance between arrival date and adult LRS from a bivariate
368 phenotypic model, instead of the sum of additive genetic and permanent environmental
369 covariances from a bivariate animal model. We did so because additive genetic (co)variances
370 involving fitness are notoriously difficult to estimate with high precision and accuracy.
371 Reassuringly, both estimates were similar in magnitude and direction ($S_{bivariate\ phenotypic\ model} = -$
372 1.234, 95% CI: -1.618, -0.805, Table S2).

373 Given that arrival date was heritable and under directional selection (Table S3), the
374 Breeder's equation predicted a rate of evolutionary change in arrival date of -0.030 days per year
375 (95% CI: -0.052, -0.018), equivalent to a total change of -0.788 days across the 27 years of study

376 (95% CI: -1.360, -0.469, Fig. 2), and translating into a response rate of -0.243 days per generation
377 (95% CI: -0.418, -0.144) and -0.019 Haldanes (95% CI: -0.032, -0.011).

378 The bivariate animal model of adult Lifetime Reproductive Success and arrival date
379 revealed a negative additive genetic covariance between arrival date and adult LRS ($COV_A = -$
380 0.474, 95% CI: -1.015, 0.247, Table S3), corresponding to a total evolutionary change in arrival
381 date of -1.541 days across the 27-year study period (95% CI: -3.298, 0.803; Fig. 2), or a predicted
382 evolutionary rate of -0.059 days per year (95% CI: -0.127, 0.031) and -0.037 Haldanes (95% CI: -
383 0.079, 0.019). While Robertson's Secondary Theorem of natural Selection also predicted a clear
384 advance in arrival date, the 95% CI of the evolutionary response estimate was wide and the lower
385 95% CI limit overlapped with zero (Table S3). Nevertheless, 89.3% of the posterior samples were
386 below zero, suggesting the additive genetic covariance between arrival date and adult LRS to be
387 very likely negative, but we might lack the statistical power to detect it with more certainty.
388 Additionally, this bivariate animal model showed the additive genetic variance in adult LRS to be
389 close to or not significantly different from zero (Table S3). While this is a pattern commonly
390 observed in fitness and fitness components (reviewed by Hendry et al. 2018), the close to zero
391 estimate might be due to either a true absence of additive genetic variance in fitness or a lack of
392 power to detect it with higher precision, being the latter case true for several fitness components
393 in this population (Moiron et al. 2021).

394

395 **DISCUSSION**

396 In this study, we investigated a case of expected evolutionary change in the timing of spring
397 migration of a seabird to further our understanding of the evolutionary dynamics of phenology in
398 natural populations, whilst also unravelling the selection pressures that underlie micro-

399 evolutionary changes. We did so by using long-term phenotypic data from a natural population of
400 common terns located at the North Sea coast of Germany and applying a series of quantitative
401 genetic models that allowed incorporating relatedness information from a social pedigree.

402 Previous studies from the common tern population at the Banter See found that the timing
403 of migration from the wintering grounds was heritable ($h^2 = 0.06 \pm 0.03$, Arnaud et al. 2013), and
404 under directional selection (Arnaud et al. 2013, but Ezard et al. 2007). In the current study we
405 corroborated those earlier findings using a considerably longer and larger dataset, finding a
406 heritability of 20.6% and a phenotypic selection differential of -1.2. Given that the requisites for a
407 potential evolutionary change were therefore met, we expected an evolutionary response to
408 selection with advanced arrival dates from the wintering grounds, i.e., a change in mean phenotype
409 towards earlier arrival dates at the genetic level, although evolutionary stasis is widespread (see
410 Table 1 in Merilä et al. 2001 and Table 20.3 in Walsh and Lynch 2018), and phenotypic plasticity
411 and/or other sources of environmental (non-genetic) variation might also lead to phenotypic
412 divergence (e.g., Gienapp et al. 2008; Gienapp and Merilä 2014; Bonnet et al. 2017, 2019). Indeed,
413 a previous study from this population of common terns has found support for individual, and, to a
414 lesser extent, additive genetic variance in plasticity of arrival date in response to an important
415 climatic factor at the main wintering areas (Moiron et al. *in revision*), although that was not the
416 case for timing of breeding (Dobson et al. 2017).

417 We found average arrival dates to have advanced 9.34 days across the 1994 to 2020 study
418 period (95% CI ranged from -10.43 to -8.24, Fig. 1A). Likewise, there was a genetic trend in the
419 average breeding values for arrival date (Fig. 1B). Taking into account the parameter uncertainty
420 and early criticisms on the use of predicted breeding values (Postma 2006; Hadfield et al. 2010),
421 we found that the expected selection response ranged from -0.062 to 0.005 days per year. Whereas

422 the 95% credible intervals of the observed genetic change slightly overlapped with zero, the
423 posterior probability of this trend being greater than zero was only 3.3% (Fig. 2). Additionally, the
424 probability of random genetic drift to generate a trend in breeding values as strong as, or stronger
425 than, the observed genetic trend was only 4.5%, indicating that an evolutionary response to
426 selection was a much more likely explanation for the observed temporal change in arrival date than
427 genetic drift. Translating the estimated annual rate of evolution to Haldanes (in units of phenotypic
428 standard deviations per generation), the estimated response corresponds to an evolutionary rate of
429 -0.017 Haldanes (95% CI: $-0.040, 0.003$), being this estimate slightly lower than both the median
430 rate of 0.035 Haldanes reported by Kinnison and Hendry (2001) for studies of less than 80
431 generations, and the average rate of -0.066 Haldanes reported by Gienapp and colleagues (Gienapp
432 et al. 2007) for several traits related to the timing of migration in birds.

433

434 **Theoretical predictions of evolutionary change**

435 Because we estimated all parameters associated with the Breeder's equation and Robertson's
436 Secondary Theorem of Selection, e.g., heritability, selection differential, and genetic covariance
437 of phenotype and relative fitness, using a single Bayesian animal model, our approach allows for
438 a quantitative comparison of both evolutionary change predictions, while also taking forward the
439 uncertainties in the estimated parameters and subsequent calculations. The predicted rate of
440 evolutionary change in arrival date based on the Breeder's equation was -0.019 Haldanes (95%
441 CI: $-0.032, -0.011$), indicating that common terns were expected to advance the timing of their
442 spring migration over time. The predicted rate of evolutionary change in arrival date based on the
443 STS was -0.037 Haldanes, similarly indicating a response to selection towards earlier spring

444 migration dates, although the associated 95% credible intervals for this estimate were partially
445 overlapping zero (95% CI: -0.079, 0.019).

446 Altogether, the theoretical predictions of evolutionary change based on Breeder's equation
447 and Robertson's Secondary Theorem of Selection were largely concordant, both in direction and
448 strength, although the STS estimate was associated with substantial statistical uncertainty. In
449 addition, these two theoretical predictions of evolutionary change were qualitatively similar to the
450 estimated genetic change (i.e., the posterior distributions for the three parameters largely
451 overlapped, Fig. 2), and in line with the observed temporal trend in mean phenotype (Fig. 1A). As
452 such, our results provide a rare case of consistency in theoretical predictions of evolutionary
453 change, that are additionally consistent with observed genetic patterns.

454 Additionally, and given that we found the evolutionary predictions from the Breeder's
455 equation and STS to be in substantial agreement, our finding implies that the Breeder's equation
456 might be a good predictor of evolutionary change in our study system, where the assumption of
457 causality most likely holds, at least, when assuming a static environment (Queller 1992; Rausher
458 1992; Kruuk et al. 2002, 2003; Morrissey et al. 2010, 2012a). It also implies that our selection
459 estimate might be mostly unbiased (i.e., the individual-level covariance between arrival date and
460 relative fitness is mostly caused by arrival date, and there are no "missing traits", Queller 1992;
461 Hadfield 2008; Morrissey et al. 2010, 2012b), shedding light on the true form of natural selection
462 acting on spring migration timing and suggesting a lack of genetic constraints that might interfere
463 with an evolutionary responses to selection. However, it is important to note that, while our three
464 estimates of evolutionary response largely agreed, conclusions must be drawn with care as we
465 cannot readily neglect the potential for the apparent agreement in evolutionary predictions to be
466 coincidental. Multiple factors can potentially bias evolution estimates (e.g., "missing fraction" in

467 fitness estimate or indirect selection from genetically-correlated traits), most often affecting
468 evolutionary predictions based on the Breeder's equation, but also on STS (Merilä et al. 2001;
469 Kruuk et al. 2002, 2003; Morrissey et al. 2012a; Walsh and Lynch 2018).

470

471 **Global warming as underlying driver of phenotypic change**

472 Temperatures in the wintering grounds of the studied population have significantly increased since
473 1994 (sea surface temperature change in West Africa between 1994 to 2020: $\beta = 0.66$, 95% CI:
474 0.16, 1.16), and such warming has been associated with earlier spring phenological events (Dobson
475 et al. 2017), making climate change-related warming an obvious mechanism underlying the
476 population's change in migration timing. However, given the observational nature of our study
477 and the complex life cycle of long-distance migratory species experiencing environmental
478 conditions in both hemispheres, we cannot test this hypothesis directly and identify the true
479 biological cue. We can, however, speculate about the potential benefits of earlier arrival from the
480 wintering grounds. Such early arrival could facilitate access to higher-quality breeding sites
481 (Kokko et al. 2004) and mates (Coppack and Both 2002; Ludwigs and Becker 2005), and an early
482 initiation of breeding. Early breeding, in turn, would allow for the production of replacement
483 clutches in cases of predation, accidental egg or chick loss, or poor food availability for rearing
484 offspring early in the season, and such replacement clutches indeed are known to significantly
485 contribute to lifetime reproductive success (Becker and Zhang 2011). As such, selection for earlier
486 timing of breeding may translate into selection for earlier timing of spring migration, by penalizing
487 late breeders. Additionally, evolutionary responses in arrival date likely are aided by assortative
488 mating for migration timing (Bearhop et al. 2005; see also Moiron et al. 2020).

489

490 **Relative contributions of evolutionary and plastic responses to phenotypic change**

491 The genetic change in response to selection as estimated from the breeding values trend and as
492 predicted by the Breeder's equation and Robertson's STS (ranging from -1.541 to -0.687 days over
493 the 27 years of study), was much smaller than the observed change in the population's phenotype
494 (change = -9.34 days over the study period). This disagreement is somewhat expected because
495 phenotypic divergence can arise from several mechanisms of both genetic and environmental
496 (nongenetic) nature.

497 The estimated genetic change in the timing of spring migration represented c. 7.4% of the
498 population's phenotypic change, subsequently indicating that the observed advance in migration
499 timing was mostly underpinned by other sources of environmental (nongenetic) variation, such as
500 phenotypic plasticity. Indeed, a recent study identified a plastic response of arrival date to climate
501 change, with birds arriving earlier at the breeding ground following warmer years at the main
502 wintering area (Moiron et al. *in revision*). Altogether, our findings therefore are fully in line with
503 general reports of phenotypic plasticity underlying the vast majority of phenotypic responses to
504 climate change (Parmesan 2006 and references therein; Vedder et al. 2013; Charmantier and
505 Gienapp 2014; Merilä and Hendry 2014), but also highlight that the existence of phenotypic
506 plasticity does not readily imply a lack of a micro-evolutionary response (or preclude their
507 simultaneous existence) (e.g., Bonnet et al. 2019).

508

509 **Concluding remarks**

510 Altogether, our study not only provides evidence for an advance in spring migration which
511 matched the widely reported phenotypic shifts in phenology in response to climate change found
512 in multiple vertebrate groups, including other migratory bird species (e.g., Root et al. 2003; Menzel

513 et al. 2006; Parmesan 2006; Thackeray et al. 2010; Poloczanska et al. 2013; Haest et al. 2021,
514 meta-analysis: Radchuk et al. 2019, but see Keogan et al. 2018), but also adds to the slowly
515 growing evidence of micro-evolutionary responses of spring phenology to climate change in
516 natural animal populations (Bonnet et al. 2019; Helm et al. 2019).

517 While plastic responses to climate change are widely reported across taxonomic groups,
518 evidence of micro-evolutionary responses to selection is still uncommon and considerably more
519 challenging to accrue, mostly because the data collection for documenting micro-evolution in the
520 wild remains a long-term daunting task. In this study, we provided rare evidence of an evolutionary
521 change in spring phenology, likely in response to global warming, while also illustrating how both
522 plastic and genetic changes can simultaneously facilitate phenotypic divergence in natural
523 populations. Our work further emphasizes the importance of maintaining long-term individual-
524 based studies of natural populations to unravel the diversity of biological responses to climate
525 change, and, generally, to understand selective patterns and evolutionary dynamics of phenotypic
526 traits in the wild.

527

528 **AUTHOR CONTRIBUTIONS**

529 M.M. conceived the study with input from S.B., A.C. and C.T. M.M. designed and conducted the
530 statistical analyses, and wrote the manuscript. S.B. manages the tern data and collated the
531 phenotypic dataset. B.H. contributed the sea surface temperature dataset. All authors contributed
532 to discussions and editing of the final paper.

533

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539

540 **CONFLICT OF INTEREST**

541 The authors declare no conflict of interest.

542

543 **DATA ARCHIVING**

544 Data will be archived in the Dryad Digital Repository upon acceptance of the manuscript.

545

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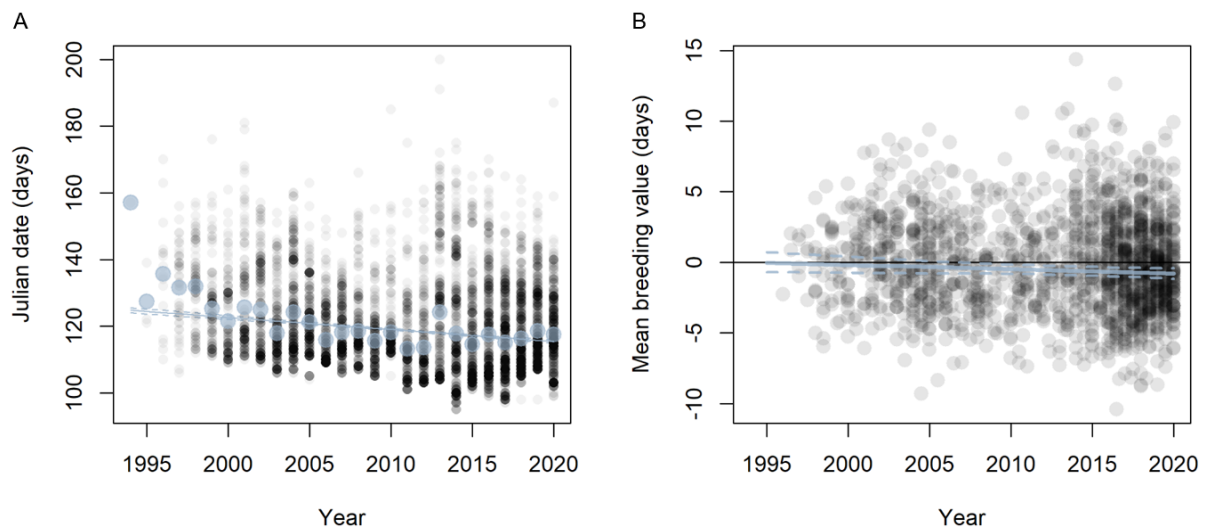
770 **TABLES**

771 **Table 1.** Estimates from a univariate animal model used to investigate sources of phenotypic
 772 variation in arrival date in a natural population of common terns. All fixed effects, except age,
 773 were mean centred and variance standardized. Estimates represent posterior modes with associated
 774 95% Credible Intervals.

Fixed effects	β (95% CI)	
Intercept	156.856	[155.578 , 158.93]
Sex [female]	-1.621	[-2.243 , -0.987]
Age [linear]	-8.51	[-8.765 , -8.21]
Age [quadratic]	0.373	[0.354 , 0.389]
Sea surface temperature	-1.278	[-2.731 , 0.368]
Sea surface temperature x age [linear]	0.16	[-0.043 , 0.366]
Sea surface temperature x age [quadratic]	-0.011	[-0.022 , -0.001]
Year	-3.89	[-5.427 , -2.483]
Year x age [linear]	1.299	[1.096 , 1.606]
Year x age [quadratic]	-0.108	[-0.123 , -0.095]
Random effects	σ^2 (95% CI)	
Additive genetic	19.918	[14.396 , 27.284]
Permanent environment	8.266	[2.766 , 13.558]
Cohort	0.474	[0.129 , 1.699]
Year	6.922	[3.68 , 13.552]
Residuals	52.782	[50.67 , 54.354]
Heritability	0.206	[0.163 , 0.301]

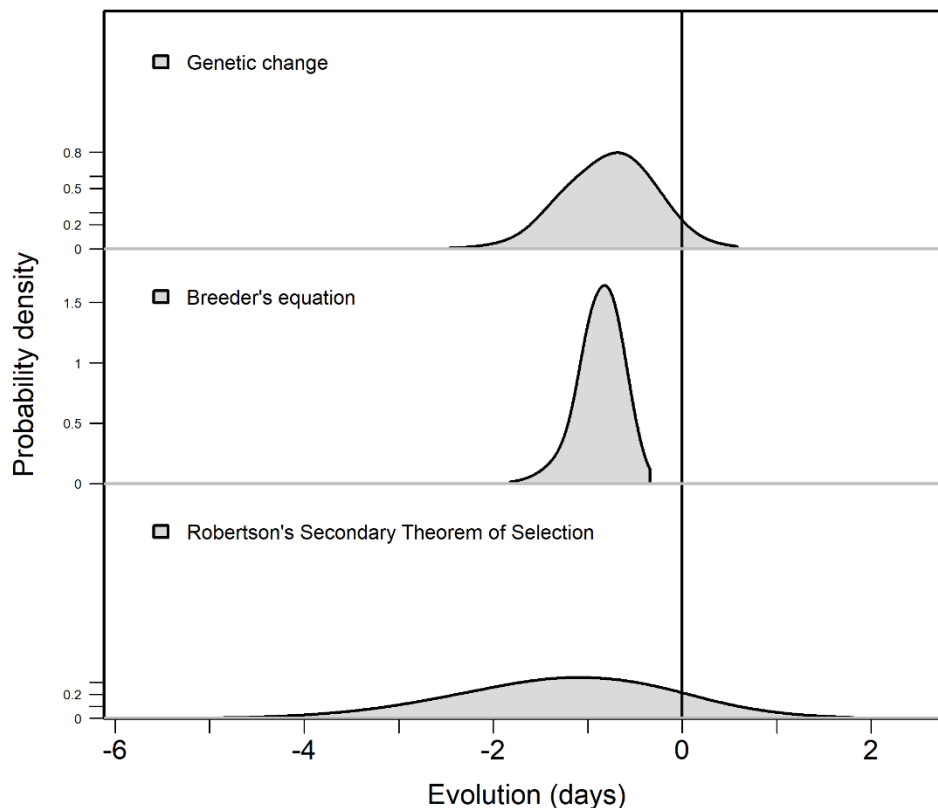
775 **FIGURES**

776 **Figure 1.** Temporal phenotypic trend in mean arrival date across years (A) and in individual
777 breeding values in arrival date across an individual's mean breeding year (B) measured in days.
778 Breeding values were extracted from the univariate animal model of arrival date (Table 1). In (A),
779 large silver dots represent annual means, small grey dots individual observations, and silver lines
780 the slope and associated 95% confidence interval of the phenotypic trend across the 27 years of
781 study. In (B), grey dots represent individual breeding values of arrival date; silver lines are the
782 slope and associated 95% confidence interval of the temporal trend in breeding values.



783

784 **Figure 2.** Posterior distributions for the estimated and predicted evolutionary response of arrival
785 date over the 27-year study period. From top to bottom, “genetic change” was measured as the
786 temporal change in individual breeding values for arrival date across the 27-year study period, the
787 evolution estimate from the “Breeder’s equation” was quantified using a bivariate animal model
788 from which we extracted the estimates of selection differential (individual-level covariance
789 between arrival date and adult lifetime reproductive success) and heritability of arrival date, and
790 the evolution estimate from “Robertson’s Secondary Theorem of Selection” was estimated as the
791 additive genetic covariance between arrival date and adult lifetime reproductive success. All
792 estimates presented here were converted to an evolutionary change (“evolution”) measured in days
793 of change over the 27-years study period. Parameter estimates are summarized in the main text.
794 All distributions have the same area and axes scales.



795