

1 Reversal of response to artificial selection on 2 body size in a wild passerine bird

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41 **Author Contributions**

42 T.K. did the analyses and wrote the paper. T.H.R., H.J. and B.-E.S. came up with the
43 idea and study design. H.J., I.J.H., B.R., H.H. and T.K. did the parentage analyses.
44 S.E. advised the statistical analyses. All authors except S.E. performed fieldwork for the
45 study. All authors contributed to the intellectual content through comments and edits
46 when writing up the manuscript.

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11 Abstract

12 A general assumption in quantitative genetics is the existence of an intermediate pheno-
13 type with higher mean individual fitness in the average environment than more extreme
14 phenotypes. Here we investigate the evolvability and presence of such a phenotype in
15 wild bird populations from an eleven-year experiment with four years of artificial selec-
16 tion for long and short tarsus length, a proxy for body size. The experiment resulted
17 in strong selection in the imposed directions. However, artificial selection was counter-
18 acted by reduced production of recruits in offspring of artificially selected parents. This
19 resulted in weak natural selection against extreme trait values. Significant responses to
20 artificial selection were observed at both the phenotypic and genetic level, followed by a
21 significant return towards pre-experimental means. During artificial selection, the annual
22 observed phenotypic response closely followed the predicted response from quantitative
23 genetic theory ($r_{years} = 0.96$, $r_{cohorts} = 0.56$). The rapid return to pre-experimental means
24 was induced by three interacting mechanisms: selection for an intermediate phenotype,
25 immigration and recombination between selected and unselected individuals. The results
26 of this study demonstrates the evolvability of phenotypes and that selection may favour
27 an intermediate phenotype in wild populations.

28 Introduction

29 Natural selection is a key process for adaptation of contemporary wild populations to
30 changing environments (Endler, 1986). Understanding how and when selection shapes
31 phenotypic variation is vital to interpret and understand observable temporal and spatial
32 patterns in fitness related traits and to address evolutionary questions in management
33 (Arnold et al., 2001; Kinnison and Hendry, 2001; Estes and Arnold, 2007; Uyeda et al.,
34 2011; Bell, 2013; Sæther and Engen, 2015). Strong selection has repeatedly been shown
35 to cause rapid adaptation in heritable traits (Endler, 1980; Grant and Grant, 1995; Losos
36 et al., 1997; Reznick et al., 1997; Hendry and Kinnison, 1999; Reznick and Ghalambor,
37 2001; Darimont et al., 2009; Calsbeek and Cox, 2010). However, most of the time wild
38 populations are subject to weak phenotypic selection while experiencing considerable de-
39 mographic and environmental stochasticity in individual fitness (Kingsolver et al., 2001;
40 Hereford et al., 2004; Rice, 2008; Coulson et al., 2010; Kingsolver et al., 2012; Engen
41 and Sæther, 2014; Sæther and Engen, 2015; Morrissey, 2016; Hendry, 2017). This cre-
42 ates random variation in individual fitness among individuals and temporal variation in
43 individual fitness among years, which complicates detection of selection on traits and
44 conclusions on their adaptive significance (Arnold et al., 2001; Postma et al., 2007; Haller
45 and Hendry, 2014; Engen et al., 2012; Engen and Sæther, 2014; Sæther and Engen, 2015;
46 Hendry, 2017).

47 Basic features of phenotypic evolution was described by Simpson (1944), applying
48 theoretical concepts originally provided by Wright (1932), as movements along a n -
49 dimensional adaptive landscape, with variation in fitness for n quantitative traits (Simp-
50 son, 1944; Arnold et al., 2001; Hendry, 2017). Lande (1976, 1979) formalized this frame-
51 work and showed that the evolutionary response to selection on correlated traits, \mathbf{R} , can
52 be expressed by a multivariate extension of the breeder's equation $\mathbf{R} = \mathbf{G}\boldsymbol{\beta}$ (Lande,
53 1979). In this quantitative genetic model, \mathbf{G} is the additive genetic variance-covariance
54 matrix and $\boldsymbol{\beta}$ the vector of selection gradients, i.e. tangents on the adaptive landscape
55 in the direction of higher fitness. The model has been applied successfully to empirical
56 data in animal breeding and laboratory experiments (Hill and Caballero, 1992; Falconer

57 and Mackay, 1996; Lynch and Walsh, 1998; Brakefield, 2003; Conner, 2003). For in-
58 stance, when artificial selection has been used to explore quantitative genetic constraints
59 (e.g. Beldade et al., 2002; Tigreros and Lewis, 2011; Bolstad et al., 2015) and predictions
60 about rates of adaptive phenotypic evolution (e.g. Lendvai and Levin, 2003; Teuschl et al.,
61 2007). In wild populations, the estimation of selection, genetic parameters and evolu-
62 tionary responses is more difficult for several reasons. For instance, environmental and
63 demographic stochasticity (Lande et al., 2003; Engen and Sæther, 2014), temporal envi-
64 ronmental changes (Merilä et al., 2001), a misidentified target of selection (Price et al.,
65 1988), selection on unmeasured genetically correlated traits (Lande and Arnold, 1983),
66 and gene flow between adjacent populations (Hendry et al., 2001; Hendry, 2017). Accord-
67 ingly, several empirical studies have reported an apparent lack of correspondence between
68 observed and predicted phenotypic changes in traits under directional selection (Merilä
69 et al., 2001; Brookfield, 2016). Many study populations have overlapping generations
70 with age structure, where an individual's contribution to population growth depends on
71 age-specific components of fecundity and survival (e.g. Reid et al., 2003). Fluctuations
72 in the age distribution of such populations may cause transient phenotypic changes if
73 phenotypes differ between age classes due to previous genetic drift or fluctuating selec-
74 tion (Lande, 1982; Coulson et al., 2003, 2006; Coulson and Tuljapurkar, 2008; Morrissey
75 et al., 2012; Engen et al., 2009, 2011, 2012, 2014). If not accounted for, such tempo-
76 ral changes may conceal responses to actual selection and cause erroneous estimates of
77 selection (Engen et al., 2014).

78 Selection experiments in the wild have a large potential to reveal novel insights into
79 adaptive evolutionary dynamics, by manipulating the observed link between phenotypes
80 and the environment (Arnold, 1983; Wade and Kalisz, 1990; Conner, 2003; Brakefield,
81 2003; Reznick and Ghalambor, 2005; Bell, 2008, 2010; Merilä and Hendry, 2014). There
82 are two basic approaches to manipulate selection in natural populations: (1) indirectly
83 by altering biotic or abiotic environmental factors or (2) directly by imposing artificial
84 selection. Both approaches have their advantages; the first offers control over the causal
85 agents of selection (e.g. Endler, 1980; Losos et al., 1997, 2001; Reznick et al., 1997;

86 Calsbeek and Smith, 2007; Calsbeek and Cox, 2010), while the second offers control over
87 the applied strength of selection. When the main interest is the evolvability of a specific
88 trait or a suite of traits within a population, the second approach is preferable (Wade and
89 Kalisz, 1990; Conner, 2003; Brakefield, 2003; Hansen and Houle, 2004, 2008; Fuller et al.,
90 2005; Bell, 2008, 2010; Merilä and Hendry, 2014). However, to our knowledge, only
91 two artificial selection experiments in wild vertebrate populations have been reported,
92 both on clutch size in birds (Flux and Flux, 1982; Postma et al., 2007). Flux and Flux
93 (1982) artificially selected for large clutch size in starlings *Sturnus vulgaris*. The response
94 was evident when comparing selected to unselected individuals. However, due to high
95 levels of gene flow there was only a marginal response in the population as a whole.
96 In a bidirectional experiment, Postma et al. (2007) artificially selected over eight years
97 for increased and decreased clutch size in two subpopulations of great tit *Parus major*.
98 Despite strong artificial selection, they found no clear evidence of evolutionary change
99 in mean clutch size at the phenotypic level. Large environmentally induced variation in
100 clutch size among years was believed to mask the response.

101 In the present study, artificial selection on tarsus length was applied in two wild pop-
102 ulations of house sparrows *Passer domesticus*, to examine the evolvability of a fitness
103 related trait and the degree to which observed trait values represent an adaptation to
104 prevailing environmental conditions. Tarsus length was selected in opposite directions in
105 the two populations for four subsequent years. Following the artificial selection, the pop-
106 ulations were monitored for another seven years. An unmanipulated control population
107 was monitored over the same period. The target of selection, tarsus length, is a herita-
108 ble trait commonly used as a proxy for structural body size in passerine birds (Jensen
109 et al., 2003, 2008; Rising and Somers, 1989; Senar and Pascual, 1997). The following
110 five objectives were addressed. First, total phenotypic selection was estimated and the
111 contribution from natural selection quantified. Second, variation in individual fitness was
112 compared among individuals with different selective ancestry. Third, annual changes in
113 tarsus length and other phenotypic traits were estimated. Fourth, the additive genetic
114 (co)variance of the traits and the annual change in breeding values were quantified. Fi-

115 nally, observed responses to selection were compared to predictions from quantitative
116 genetic theory.

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117 **Material and methods**

118 **Study system**

119 The study was conducted in three insular populations of house sparrows in northern
120 Norway. The islands, Hestmannøy (66°33'N, 12°50'E), Vega (65°40'N, 11°55'E) and Leka
121 (65°06'N, 11°38'E), are located along a north-south gradient, separated by 97 (Hestmannøy-
122 Vega) and 54 (Vega-Leka) km of ocean and small islands along the coastline (see map
123 in Hagen et al., 2013). Thus, the geographical distance and the sedentary nature of the
124 house sparrow ensured virtually no migration between the study populations (Altwegg
125 et al., 2000; Tufto et al., 2005; Pärn et al., 2012). All individuals in the populations
126 inhabit dairy farms and human settlements, where they breed in holes and cavities from
127 May until mid-August (Ringsby et al., 1998). The mean generation time of house spar-
128 rows in natural populations in this area has been found to be 1.97 years (Stubberud et al.,
129 2017).

130 In the years 2001-2012, individuals were captured and marked with a unique combina-
131 tion of a numbered metal leg ring from the Ringing Centre at Museum Stavanger and three
132 plastic colour leg rings. Individuals were either followed from the nestling stage or when
133 captured in mist nets during summer (May-August), autumn (late September-October;
134 all populations) or winter (February-March; Leka and Vega). Over 90 % (Hestmannøy)
135 and ~ 90 % (Leka and Vega) of the winter populations were marked at all times during
136 the study. At first capture, a small blood sample (25 μ L) was collected, which enabled
137 the construction of a genetic pedigree for each population. Parentage analyses were per-
138 formed in Cervus 3.0 software with 95 % confidence for parentage assigned (Marshall
139 et al., 1998; Kalinowski et al., 2007), based on genotyping putative parents and offspring
140 for 14 microsatellite markers (Jensen et al., 2004, 2008; Rønning et al., 2016).

141 The data was organized with pre-breeding census and two age classes: 1 year old and
142 2+ years old. Hence, annual individual survival was recorded as 1 if an individual in year
143 t was re-sighted (captured or observed) in year $t + 1$ (otherwise 0). Any emigrants from
144 the islands were treated as dead individuals. For each individual, the annual number of

145 offspring produced was recorded as the number of offspring born in year t that survived to
146 year $t + 1$ (i.e. recruits). House sparrows go through a complete post-juvenile and post-
147 breeding moult during autumn, after which ageing based on plumage is not possible.
148 Hence, ageing was either made on individuals marked before the post-juvenile moult
149 during summer or based on an assumption that all full-grown unmarked individuals were
150 born in the most recently completed breeding season. Individuals which we were unable
151 to age, were excluded from the analyses in the year they were marked. In addition, we
152 excluded a few individuals with missing traits and all individuals from one farm at each
153 experimental island, where we did not have access until the final years of the study.

154 Morphological measurements

155 Full-grown individuals were measured for tarsus length (± 0.005 mm), body mass (± 0.05
156 g), wing length (± 0.5 mm), bill length (± 0.005 mm) and bill depth (± 0.005 mm). The
157 measurements were performed by several different fieldworkers. After an initial period of
158 training, each fieldworker measured approximately 30 individuals together with T.H.R. or,
159 in some cases, another experienced fieldworker. Then all linear measurements were ad-
160 justed according to T.H.R. by adding mean differences when found significant ($P < 0.05$)
161 using paired t-tests. All traits, except tarsus length, display seasonal variation (Ander-
162 son, 2006). Hence, only measurements from the main sampling periods were used in the
163 analyses, i.e. summer for the Hestmannøy population and winter for the Leka and Vega
164 populations. Furthermore, within-individual age effects were investigated for body mass,
165 wing length, bill length and bill depth, using an extended data set over the years 1993-
166 2012 at Hestmannøy and 2001-2012 at Leka and Vega. Due to the difference in sampling
167 season, Hestmannøy was analysed separately. Traits were age-standardised by fitting a
168 linear mixed effects model with age and age² as explanatory variables, random intercepts
169 with year, cohort and individual identity, and an individual random slope to separate out
170 any between-individual variation (Bates et al., 2015; Schielzeth and Forstmeier, 2009).
171 The significance of each age variable was tested by likelihood ratio tests of nested models
172 (fitted using maximum likelihood). All traits with significant age effects were adjusted to

173 age 1, using predicted values from the model, before individual means were calculated.

174 Body mass scale with body size, measured as tarsus length, through an allometric
175 relationship $bodymass = b \times bodysize^k$, where k is the allometric exponent (Huxley,
176 1932). This relationship was linearised for each sex and population separately by log
177 transformation. Residuals from the log-log linear regressions were used as measures of
178 individual body condition in subsequent analyses (Schulte-Hostedde et al., 2005).

179 **Experimental procedure**

180 Each winter of the four years 2002-2005, opposing artificial selection on tarsus length
181 was imposed after a census in the Leka and Vega populations. During the experimental
182 manipulations ~ 90 % of individuals in each population were captured and kept in a
183 large aviary (abandoned cow barn) with *ad libitum* access to food (sunflower seeds, grain
184 feed for cattle, oats and slices of bread), water and perching branches. The ranges in
185 sample sizes during period of artificial selection (2002-2005) were 172-222 (Leka), 155-
186 352 (Vega) and 59-80 (Hestmannøy), while the ranges in the subsequent period (2006-
187 2012) were 89-216 (Leka), 102-330 (Vega) and 104-219 (Hestmannøy). Within each sex,
188 all individuals with tarsi longer (Leka) or shorter (Vega) than the limit of mean ± 0.3
189 SD were returned to their origins, while the remaining individuals were translocated to
190 populations located at least 70 km from the islands (see also Skjelseth et al., 2007). On
191 average, 56.4 % (Leka) and 62.9 % (Vega) of all captured individuals were removed at
192 each annual episode of artificial selection, such that the artificially selected individuals
193 constituted approximately 78 % of the breeding populations. The whole procedure took
194 between one and two weeks for each population. In the subsequent seven years (2006-
195 2012) on Leka and Vega, the same procedure was followed, except that all individuals
196 were returned to their origin. The Hestmannøy population was used as an unmanipulated
197 control, where individuals were returned directly to the place of capture after banding
198 and measurements. Henceforth, these populations are referred to as *high* (Leka, selected
199 for large body size), *low* (Vega, selected for small body size) and *control*. Each individual
200 in the *high* and *low* populations was assigned a selection category, selected, unselected,

intermediate or other, based on whether their parents had been artificially selected (Table 1). Our genetic parentage analyses had a very high probability of assigning a parent to an individual, given that the parent had been sampled. Hence, when no genetic parent had been assigned to an individual, its parents were assumed to not have been artificially selected.

Data analysis

Phenotypic population differences

Differences in phenotype between populations in 2002, before the onset of the experiment, were explored using a multivariate analysis of variance (MANOVA). *Post hoc* tests for each trait were performed by separate analyses of variance (ANOVA). Tukey's range tests were applied to identify which populations differed phenotypically. Pairwise phenotypic correlations are shown in Table S1. Any sexual dimorphism in the traits was accounted for in the models by including sex as a categorical variable.

Analyses of directional selection

Analyses of directional selection were performed for each sex and population separately, and structured into two periods: (1) years 2002-2005 (with artificial selection) and (2) years 2006-2011 (without artificial selection). The demographic framework in the R package *lmf* was applied to analyse selection (Engen et al., 2012). This recently developed framework integrates evolutionary theory with an age-structured model for population dynamics, which accounts for overlapping generations and fluctuating age distribution in the estimation of selection (Engen et al., 2009, 2011, 2012, 2014). The annual absolute fitness of an individual j in age class i was defined by the individual reproductive value (Engen et al., 2009),

$$W_{ij} = J_{ij}v_{i+1} + B_{ij}v_1/2, \quad (1)$$

where J_{ij} is 1 if the individual survives (otherwise 0), B_{ij} is the number of recruits

225 produced and v_{i+1} and v_1 are age-specific reproductive values (Engen et al., 2009; Sæther
 226 and Engen, 2015). Defining fitness this way enables correct estimation of an individual's
 227 contribution to the total reproductive value next year, by accounting for both survival and
 228 reproduction (Engen et al., 2011, 2012; Metcalf and Pavard, 2007; Wilson and Nussey,
 229 2010; Sæther and Engen, 2015). However, additional insights into the selective processes
 230 could be obtained by analysing different fitness components separately. This was achieved
 231 by defining viability (W_{sij}) and fecundity (W_{fij}) fitness as the first and second additive
 232 component in equation 1 (Engen et al., 2011).

233 The age-specific reproductive values (\mathbf{v}), stable age distribution (\mathbf{u}) and deterministic
 234 multiplicative growth rate (λ) of a population are needed to calculate individual repro-
 235 ductive values and estimate selection gradients. These were obtained from the sex-specific
 236 mean projection matrix (\mathbf{I}), estimated separately for each population (Table S2) (Caswell,
 237 2001). With two age classes, 1 year old and 2+ years old, \mathbf{I} had age-specific fecundities
 238 (f_i) in the first row and age-specific survivals (s_i) in the bottom row. Age-specific fecun-
 239 dities and survivals for each sex and population were estimated as their means across the
 240 whole study period (Engen et al., 2011). In these calculations, experimentally removed
 241 individuals were excluded in the year they were removed. Then \mathbf{v} , \mathbf{u} and λ were esti-
 242 mated as the scaled left and right eigenvector, and the dominant eigenvalue of \mathbf{I} (Table
 243 S2) (Caswell, 2001). Eigenvectors were scaled according to $\sum u_i = 1$ and $\sum v_i u_i = 1$ (Engen
 244 et al., 2009). Conditioned on the sex-ratio at birth ($q =$ proportion of females) the growth
 245 rate of the male and female segment in each population has to be identical (Engen et al.,
 246 2010). Hence, we estimated the growth rate (λ_f) for females and set the growth for males
 247 equal to the females by scaling all male fecundities by a constant (c). The constant was
 248 estimated by solving the Euler-Lotka equation for the male segment of the population,
 249 $c(1 - q) \sum_{k=1}^{\infty} \lambda^{-k} l_k m_k = 1$, using Newton's method. Here, $l_k = \prod_{i=1}^{k-1} s_i$, $m_k = f_{i=k}$,
 250 $\lambda = \lambda_f$ and in house sparrows the sex ratio at birth does not deviate significantly from
 251 1:1 ($q = 0.5$, Anderson, 2006).

252 All k traits were centred by the global mean across years prior to analyses. Then
 253 directional selection gradients were estimated for each year and age class separately, us-

ing multiple regressions of absolute fitness on the trait values (Lande and Arnold, 1983; Engen et al., 2012). Annual selection gradients ($\boldsymbol{\alpha}_t = (\alpha_{0t}, \alpha_{1t}, \dots, \alpha_{kt})$) were given as the weighted average of age-specific gradients, $\alpha_{mt} = \sum_i u_i \alpha_{imt}$, where $m = (0, 1, \dots, k)$ (Engen et al., 2011). Then, assuming no fluctuating selection, the temporal mean selection gradients $\boldsymbol{\alpha} = \mathbf{E}\boldsymbol{\alpha}_t$ were estimated according to procedures in Engen et al. (2012). In addition to estimating the total directional selection (due to artificial and natural selection), we also estimated natural selection separately for the artificially selected individuals. Natural selection was separated into total, fecundity and viability selection. In this model, the growth rate λ is a measure of the expected individual reproductive value (i.e. the mean absolute fitness), with annual estimates given by, $\lambda_t = \sum_i u_i \mathbf{E}W_{it}$.

The directional selection coefficients ($\boldsymbol{\alpha}$) were estimated using absolute fitness. Hence, the standard SD-scaled selection gradients ($\boldsymbol{\beta}_\sigma$) were calculated by $\boldsymbol{\beta}_\sigma = \lambda^{-1} \boldsymbol{\alpha} \odot \boldsymbol{\sigma}$, where $\boldsymbol{\sigma}$ is the vector of trait standard deviations (averaged over all years) and \odot denotes element-wise multiplication (Engen et al., 2012). Statistical significance of temporal mean selection gradients was assessed using a multinormal bootstrap procedure for 10000 bootstrap replicates (Engen et al., 2012). 95 % confidence intervals were calculated from the estimated bootstrap distributions.

Demographic and environmental stochasticity, and selection are integral parts in the applied demographic framework for estimating selection. The demographic and environmental variance for the population were estimated as $\sigma_d^2 = \sum_i u_i \sigma_{di}^2$, where $\sigma_{di}^2 = \text{Evar}(W_i|z, \varepsilon_t)$ and $\sigma_e^2 \approx \text{var}(\alpha_{0t})$, where α_{0t} is the intercept in year t (Engen et al., 2012).

Variation in individual fitness

The difference in survival and production of recruits among selected, unselected and intermediate individuals (see Table 1) in the years 2003-2012 were analysed using mixed effects logistic and Poisson regression models, fitted using the R package *lme4* (Bates et al., 2015). As the proportion of selected individuals increases over years, an environmental (year) effect could not be estimated directly in the analyses without conflating it with fitness consequences from the experiment. Hence, a year effect (slope) was esti-

282 mated for each of the two dependent variables with only unselected individuals. Among
283 unselected individuals, there was no significant trend during the years 2003-2009 in re-
284 cruit production in the *high* population ($b_{year} = -0.03 \pm 0.03$, $\chi^2 = 1.01$, $df = 1$, $P =$
285 0.314), but a slight decrease in the *low* population ($b_{year} = -0.07 \pm 0.03$, $\chi^2 = 6.94$, $df =$
286 1 , $P = 0.008$). Survival rates did not show any significant temporal trend across years in
287 unselected individuals in either population (*high*: $b_{year} = 0.03 \pm 0.05$, $\chi^2 = 0.29$, $df = 1$,
288 $P = 0.587$, *low*: $b_{year} = -0.03 \pm 0.05$, $\chi^2 = 0.31$, $df = 1$, $P = 0.577$).

289 The significant decrease in recruit production in the *low* population was accounted
290 for in subsequent analyses by fitting it as a covariate with known effect (i.e. offset).
291 In addition, a random intercept associated with individual identity was estimated, age
292 and sex were included to account for differences in survival and fecundity between ages
293 and sexes, and two-way interactions to estimate age- and sex-specific differences among
294 selection categories were included. The significance of the terms of interest were tested
295 using likelihood ratio tests of pairs of nested models fitted with maximum likelihood,
296 where twice the difference in log-likelihood is χ^2 -distributed with $df_1 - df_2$ degrees of
297 freedom.

298 **Observed phenotypic change**

299 Annual arithmetic mean phenotypes in age-structured populations are subject to tran-
300 sient temporal fluctuations due to fluctuations in the age distribution and variation in
301 mean phenotype among age classes (Engen et al., 2014). Thus, phenotypic changes in
302 each trait following artificial selection were explored by estimating annual weighted means
303 and 95 % confidence intervals with weights \mathbf{u} . The weighting accounted for the effect of
304 fluctuating age distribution on phenotypic means (Engen et al., 2014, 2012). Piecewise
305 regression for each population was used to estimate the change in annual weighted mean
306 phenotype across the years 2002-2012, with a breakpoint in 2006. Sex was included to
307 account for any sexual dimorphism. These rates of responses to selection result from the
308 partial transmission of selection to recruiting individuals and survival of adults, with the
309 final response achieved when all individuals under selection have stopped reproducing.

310 Corresponding analyses were performed on cohort arithmetic means across the cohorts
 311 2000-2011 with a breakpoint in cohort 2005, to investigate annual changes in recruited off-
 312 spring separately. These means will be subject to transient temporal fluctuations due to
 313 fluctuations in age distribution and age-specific phenotypic means among parents. Each
 314 cohort consisted of offspring with two, one or no artificially selected parents (see Table 1).
 315 Hence, phenotypic changes across the cohorts 2000-2005 were also analysed separately
 316 within selected, intermediate and unselected offspring. Permutation tests were used to
 317 test whether slopes were significantly different from zero, and bootstraps were performed
 318 to estimate standard errors of the estimated slopes. In both cases 10000 iterations of the
 319 models were performed.

320 Quantitative genetic analyses

321 Analyses of additive genetic effects included phenotypes from 1141, 1404 and 554 in-
 322 dividuals sampled from the *high*, *low* and *control* population over the years 2002-2012.
 323 Multivariate Bayesian animal models were constructed with all five traits to estimate
 324 additive genetic effects (breeding values), and the \mathbf{G} -matrices with additive genetic vari-
 325 ances and covariances (Lynch and Walsh, 1998; Kruuk, 2004; Hadfield, 2010). As sample
 326 sizes did not allow for separate analyses of females and males, models were constructed
 327 with sex as a categorical fixed effect. For each trait, phenotypic variation (σ_P^2) was sepa-
 328 rated into additive genetic variance (σ_A^2), cohort variance (σ_C^2) and residual variance (σ_R^2),
 329 such that $\sigma_P^2 = \sigma_A^2 + \sigma_C^2 + \sigma_R^2$. The cohort effect ensured that estimated breeding val-
 330 ues were unbiased with respect to any systematic environmental variation in phenotypes
 331 (Postma, 2006).

332 Models were fitted using *MCMCglmm* version 2.22.1 (Hadfield, 2010) with Gaussian
 333 distribution and identity link function. Prior to analyses, all traits were standardized
 334 by their standard deviation across all individuals to improve model mixing and ease
 335 construction of priors. The resulting \mathbf{G}_σ -matrices have heritabilities on the diagonal and
 336 genetic correlations in off-diagonal elements. Priors for the fixed effects were the normal
 337 distribution with zero mean and large variance (10^{10}), while a parameter expanded prior

338 was used for the variance components by specifying $V = \mathbf{I}_5$, $\text{nu} = 5$, $\text{alpha.mu} = \mathbf{0}_5$ and
339 $\text{alpha.V} = \mathbf{I}_5 \times 100$. Here \mathbf{I}_n is the identity matrix and $\mathbf{0}_n$ is a zero vector with dimensions
340 n . Care was taken to ensure good mixing of the chains and that specified priors did not
341 have exaggerated influence on posterior distributions, by examining the sensitivity of
342 the models to different choices of priors. In the analyses, runs with a burn-in period of
343 3000 and a thinning interval of 500 ensured low autocorrelation (generally < 0.1) for a
344 total of 1000 independent random samples from the stationary posterior distribution. All
345 estimates are reported as the posterior mode and 95% credibility intervals (CI).

346 For each trait and population, the temporal change in mean breeding value was anal-
347 ysed across years 2002-2012 and cohorts 2000-2011. Piecewise regression was used with
348 annual weighted mean breeding value (weights \mathbf{u}) and a breakpoint in year 2006, or
349 arithmetic cohort mean breeding value and a breakpoint in cohort 2005. To account for
350 uncertainty in the estimated breeding values, these analyses were performed for each re-
351 alization of the MCMC chain to obtain a full posterior distribution for temporal change
352 (Hadfield et al., 2010). Thus, posterior modes for temporal change could be calculated
353 with credibility intervals to assess whether the changes were significantly different from
354 zero. We also quantified whether estimated slopes differed significantly from slopes ex-
355 pected under genetic drift. This was done by simulating random breeding values down
356 the pedigree for each realization of the \mathbf{G}_σ -matrix in the MCMC chain, using the *rbv*
357 function in the *MCMCglmm* package (Hadfield, 2010). The probability of obtaining a
358 slope of the magnitude observed or larger was then calculated as a two-tailed test using
359 the posterior distribution of the slope under genetic drift.

360 **Response to selection**

361 To assess the agreement between observed phenotypic changes and predictions from quan-
362 titative genetic theory, the relationships between annual predicted and observed responses
363 to selection were explored. Because we could only estimate the \mathbf{G} -matrix with sexes com-
364 bined, observed and predicted responses were averaged across sexes. The annual observed
365 phenotypic response to selection was calculated for each trait by subtracting the weighted

366 mean of parents at time t from the weighted mean at time $t + 1$, with weights \mathbf{u} . At time
 367 $t + 1$, both adults which survived and recruiting offspring from known parents are included
 368 to calculate the weighted mean. To investigate the response in offspring separately, the
 369 observed phenotypic response in recruits were calculated by replacing the weighted mean
 370 at time $t + 1$ by the arithmetic mean of recruiting offspring from known parents. The
 371 response in recruits will vary temporally due to fluctuations in the age distribution of
 372 parents, and will only capture the partial response because the final response will be
 373 achieved when all individuals under selection have stopped reproducing.

374 The annual predicted phenotypic response to selection (\mathbf{R}_t) averaged across females
 375 (f) and males (m) was calculated as

$$\mathbf{R}_t = \frac{(\mathbf{G}_\sigma \boldsymbol{\beta}_{\sigma_{tf}}) \odot \boldsymbol{\sigma}_{tf} + (\mathbf{G}_\sigma \boldsymbol{\beta}_{\sigma_{tm}}) \odot \boldsymbol{\sigma}_{tm}}{2}, \quad (2)$$

376 where \mathbf{G}_σ is the variance-standardized additive variance-covariance matrix, $\boldsymbol{\beta}_{\sigma_{tf}}$ and
 377 $\boldsymbol{\beta}_{\sigma_{tm}}$ are the vectors of variance-standardized selection gradients, and $\boldsymbol{\sigma}_{tf}$ and $\boldsymbol{\sigma}_{tm}$ are the
 378 vectors of phenotypic standard deviations. Analyses were performed using the statistical
 379 software R version 3.3.3 (R Core Team, 2016).

380 Results

381 In 2002, before the onset of artificial selection, there were significant phenotypic differ-
382 ences between the three populations (MANOVA: $F_{10,698} = 20.84$, $P < 0.001$, ANOVAs:
383 all $P < 0.001$, Table S3). Tarsus length was shorter in the *low* population than in the
384 *high* (mean difference = -0.27, $P = 0.004$) and *control* (mean difference = -0.44, $P <$
385 0.001) population, while the *high* and *control* populations did not differ significantly
386 (mean difference = -0.18, $P = 0.277$).

387 Phenotypic selection

388 The artificial selection resulted in strong directional selection towards longer or shorter
389 tarsus in the experimental populations in the years 2002-2005 (Fig. 1). There was no
390 direct artificial selection on the other phenotypic traits (all $P > 0.05$, Table S4). When
391 excluding artificial selection, there was significant directional natural selection on tarsus
392 length towards the pre-experimental phenotypic mean in males of the *low* population
393 (Fig. 1B). When separating natural selection into viability and fecundity selection, only
394 fecundity selection was significant (Fig. 1B). A similar non-significant trend of directional
395 natural selection towards pre-experimental means was also observed in females in the *low*
396 population and in both sexes in the *high* population (Fig. 1A and B). Hence, there was
397 a tendency for natural selection towards phenotypic pre-experimental means (Fig. 1A
398 and B). There was no significant directional natural selection on phenotypic traits in the
399 *control* population over the years 2002-2005 (all $P > 0.05$, Table S4).

400 During the seven years after the artificial selection ended (2006-2011), there was
401 significant viability selection towards pre-experimental mean tarsus length in females
402 in the *high* population, but the total directional selection was non-significant (Fig. 1C).
403 Instead, there was positive directional selection for longer tarsus in males of the *high*
404 population (Fig. 1D). This was the result of a combined effect of both fecundity and
405 viability selection, as neither component was significant when analysed separately (Fig.
406 1D). There was no further significant directional natural selection detected in either the
407 *high* or *low* population in the years 2006-2011 (all $P > 0.05$, Table S5).

408 The demographic variance (σ_d^2) was generally larger in both experimental populations
 409 during the period of artificial selection than in the subsequent period (*high*: $\Delta\sigma_d^2 = -$
 410 0.18 , *low*: $\Delta\sigma_d^2 = -0.45$, Table S6). On average across the populations, the variance in
 411 recruit production decreased by 34.0 % and the variance in survival decreased by 4.3
 412 % after completion of the period with artificial selection. Hence, removing individuals
 413 from the populations increased the demographic variation in recruit production during
 414 the manipulated breeding seasons.

415 Variation in individual fitness components

416 Selected and intermediate individuals produced significantly fewer recruits than unse-
 417 lected individuals in the *high* population ($\chi^2 = 9.65$, $df = 2$, $P = 0.008$, Table 2A). In the
 418 *low* population a similar pattern was evident among age 1 individuals (*selection status*
 419 \times *age*: $\chi^2 = 10.92$, $df = 2$, $P < 0.001$, Table 2B), where selected individuals produced
 420 fewer recruits than unselected individuals. There were no significant differences in sur-
 421 vival among individuals in different selection categories (*high*: $\chi^2 = 0.98$, $df = 2$, $P =$
 422 0.613 , *Low*: $\chi^2 = 2.58$, $df = 2$, $P = 0.275$). Hence, individuals with artificially selected
 423 parents appeared to have lower fitness than individuals with unselected parents.

424 Observed phenotypic change

425 In the period 2002-2006, the weighted mean tarsus length of both sexes significantly in-
 426 creased in the *high* population ($b_{year} = 0.126 \pm 0.021$, $P < 0.001$, Fig. 2A) and decreased in
 427 the *low* population ($b_{year} = -0.112 \pm 0.020$, $P < 0.001$, Fig. 2C). In the *control* population
 428 there was no significant change in weighted mean tarsus length during the same period
 429 ($b_{year} = -0.027 \pm 0.028$, $P = 0.367$, Fig. 2E). The weighted phenotypic mean of some of
 430 the other four traits also changed significantly from 2002 to 2006 in the experimental
 431 populations (Table 3).

432 Across the cohorts 2000-2005, arithmetic mean tarsus length of selected offspring
 433 increased significantly in the *high* population ($b_{cohort} = 0.167 \pm 0.040$, $P < 0.001$, Table
 434 4A) and decreased in the *low* population ($b_{cohort} = -0.091 \pm 0.041$, $P = 0.035$, Table 4B).

Such changes were not evident among unselected offspring (*high*: $b_{cohort} = 0.008 \pm 0.079$, $P = 0.898$, *low*: $b_{cohort} = 0.002 \pm 0.044$, $P = 0.964$, Table 4). When pooling all offspring, there was still a significant increase in tarsus length across cohorts 2000-2005 in the *high* population ($b_{cohort} = 0.099 \pm 0.031$, $P = 0.002$, Fig. S1 and Table S7), whereas there was no significant change in the *low* population ($b_{cohort} = -0.007 \pm 0.024$, $P = 0.786$, Fig. S1 and Table S7). In the *control* population there was no significant change in tarsus length across the same cohorts ($b_{cohort} = 0.000 \pm 0.038$, $P = 0.994$, Fig. S1 and Table S7).

In the period 2006-2012, there was a significant decrease in weighted mean tarsus length in the *high* population ($b_{year} = -0.088 \pm 0.013$, $P < 0.001$, Fig. 2A). The *low* population displayed a marginally non-significant increase in weighted mean tarsus length over the same period ($b_{year} = 0.027 \pm 0.013$, $P = 0.055$, Fig. 2C). However, fig. 2C shows that the *low* population reached its pre-experimental weighted mean tarsus length already in 2007. Hence, both populations returned towards their pre-experimental tarsus length following the end of artificial selection. The other four traits generally also returned towards pre-experimental weighted means (Table 3). In the *control* population there was a slight decrease in weighted mean tarsus length over the years 2006-2012 ($b_{year} = -0.035 \pm 0.016$, $P = 0.014$, Fig. 2E).

Across the cohorts in the same period (2005-2011) there was a significant decrease in arithmetic mean tarsus length in the *high* population ($b_{cohort} = -0.073 \pm 0.018$, $P < 0.001$, Fig. S1 and Table S7). However, there was also a significant decrease in arithmetic mean tarsus length in both the *low* ($b_{cohort} = -0.050 \pm 0.016$, $P = 0.002$, Fig. S1 and Table S7) and *control* ($b_{cohort} = -0.069 \pm 0.027$, $P = 0.003$, Fig. S1 and Table S7) population.

457 Observed genetic change

In all three populations there were significant heritability for tarsus length and the other four traits (Table 5). Furthermore, there were positive genetic correlations between tarsus length and several of the other traits in the *high* and *low* populations (Table 5A and B). A similar pattern was found in the *control* population, but credibility intervals were wide enough to include zero for all genetic correlations (Table 5C).

463 Over the years 2002-2006, the weighted mean estimated breeding values for tarsus
 464 length increased significantly in the *high* population ($b_{year} = 0.110$, CI = [0.072, 0.152],
 465 Fig. 2B and Table 6A) and decreased significantly in the *low* population ($b_{year} = -0.103$,
 466 CI = [-0.137, -0.059], Fig. 2D and Table 6B). These changes were of larger magnitude
 467 than expected by genetic drift alone (*high*: $P < 0.001$, *low*: $P = 0.002$, Fig. 2B and D).
 468 In the subsequent period (2006-2012), the weighted mean estimated breeding values for
 469 tarsus length returned towards their pre-experimental means (*high*: $b_{year} = -0.055$, CI =
 470 [-0.081, -0.026], *low*: $b_{year} = 0.044$, CI = [0.021, 0.067]). Again the rates of change were
 471 larger than expected by genetic drift alone (*high*: $P = 0.037$, *low*: $P = 0.013$, Fig. 2B
 472 and D). Correlational change in estimated breeding values for the other traits were not
 473 larger than expected from genetic drift alone (Table 6A and B). Similarly, in the *control*
 474 population there were no changes in estimated breeding values larger than expected by
 475 genetic drift alone (Fig. 2F and Table 6C). Similar results were obtained for the annual
 476 changes in cohort arithmetic mean estimated breeding values (Fig. S1 and Table S8).

477 **Observed and predicted response to selection**

478 The observed response to selection closely followed the predicted response during the
 479 years of artificial selection ($r_{2002-2005} = 0.96$, Fig. 3A), with a tendency for observed
 480 responses to be of larger magnitude than predicted. This observed response include both
 481 adults which had survived and offspring that had recruited. Hence, the overshoot of the
 482 predicted response was as expected. When limiting the observed response to offspring that
 483 recruited, the partial observed response also followed the predicted response, but with
 484 larger deviation from the 1:1 line ($r_{2002-2005} = 0.56$, Fig. 3B). In the seven consecutive
 485 years with no artificial selection, there was no clear relationship between predicted and
 486 observed responses (total: $r_{2006-2011} = 0.15$, only recruits: $r_{2006-2011} = 0.06$).

487 Discussion

488 Artificial selection on tarsus length resulted in strong directional selection in opposite
489 directions in two house sparrow populations (Fig. 1). However, individuals with at least
490 one artificially selected parent produced fewer recruits than unselected individuals (Table
491 2), such that there was a tendency for natural selection to counteract artificial selection
492 (Fig. 1). Still, artificial selection was much stronger than natural selection and resulted
493 in a significant response in tarsus length in both experimental populations (Fig. 2, Tables
494 3 and 4). The observed phenotypic response during artificial selection closely followed
495 the predicted response according to the multivariate breeder's equation (Fig. 3). Fur-
496 thermore, the response in breeding values was much larger than expected by genetic drift
497 alone (Fig. 2, Table 6). During the seven years period following the artificial selection,
498 the mean tarsus length and estimated breeding values in the populations gradually re-
499 turned towards their pre-experimental means (Fig. 2, Tables 3 and 6). Again, the rates
500 of change in breeding values were larger than expected by genetic drift alone (Table 6).

501 Any finite population may undergo random phenotypic and genetic changes due to ge-
502 netic drift (and mutation in the long run) (Lande, 1976; Swallow et al., 2009). Replicated
503 selection lines in artificial selection experiments have obvious advantages for estimating
504 the average response and to separate between selection and genetic drift as causes of
505 phenotypic change (Henderson, 1989, 1997; Konarzewski et al., 2005; Swallow et al.,
506 2009). However, in artificial selection experiments in natural populations, adding repli-
507 cates involves synchronous experiments on additional suitable populations with similar
508 population dynamics and under the same environmental influences. Even if such popu-
509 lations were available, it would represent a considerable increase in logistic effort, which
510 was infeasible in the present system. Instead, we applied a bidirectional design to explore
511 selection for both increased and decreased trait values. The construction of genetic pedi-
512 grees allowed us to conduct simulations of change in breeding values under genetic drift.
513 Hence, the probability that the observed changes could have occurred by genetic drift
514 alone could be quantified following Hadfield et al. (2010) and Postma (2006). Although,
515 we were not able to estimate confidence intervals on the average expected responses under

516 replicated experiments, we were still able to exclude genetic drift as an explanation for
517 our results. This approach has previously been applied to observational studies in natural
518 populations. For instance, to distinguish the effects of genetic drift and trophy hunting
519 as causes of temporal change in horn length in bighorn sheep *Ovis canadensis* (Pigeon
520 et al., 2016).

521 Artificial selection experiments in the wild necessitate capture and tracking of a large
522 proportion of individuals in a population to perform selection and obtain unbiased es-
523 timates of responses. Here, a morphological trait was subject to selection by removing
524 individuals with phenotypic values more extreme than a given threshold value. Our effort
525 to capture and include all individuals in the experiment was considerable. Despite this,
526 sampling was still incomplete and approximately 20-25 % of the breeding populations
527 remained unselected each year. Most of the unselected individuals were located in un-
528 available subpopulations at mainly one farm on each study island. This resulted in a
529 mixture of selected, intermediate and unselected offspring which recruited to the popula-
530 tions. High quality genetic pedigrees allowed us to distinguish between these individuals.
531 Hence, offspring with unselected parents could be excluded to obtain unbiased estimates
532 of responses to artificial selection, and offspring which differed in selective background
533 could be contrasted to explore the variation in each component of individual fitness. A
534 similar use of contrasts was applied in an artificial selection experiment by Flux and Flux
535 (1982) and enabled robust conclusions about the evolutionary dynamics.

536 Immigrants into the experimental populations originate from distant populations or
537 from the unavailable subpopulations on the study islands. These were pooled together
538 with any few unselected residents as individuals in these two groups could not be dis-
539 tinguished. The focal populations are located distant to other known populations and
540 house sparrows are generally highly sedentary (Anderson, 2006). Previous studies have
541 found that only a small fraction of individuals disperse between populations separated
542 by more than a few kilometres (Altwegg et al., 2000; Tufto et al., 2005; Anderson, 2006;
543 Pärn et al., 2009, 2012). Immigrant house sparrows do not differ morphologically from
544 residents (Altwegg et al., 2000), but immigrant males produce fewer recruits than resident

545 males (Pärn et al., 2009, 2012). Hence, immigrants were likely to mostly originate from
546 the unselected subpopulations, and to have morphological trait values that were randomly
547 distributed around pre-experimental means (see Table S3; see also Holand et al., 2011).
548 Any immigrants from distant populations should not compromise the conclusions on vari-
549 ation in individual fitness, but rather make the analyses more conservative as they might
550 contribute to smooth out fitness differences between selected and unselected individuals.

551 Mean tarsus length responded to our artificial selection, with significant changes to-
552 wards more extreme phenotypic and genetic values in both experimental populations
553 (Fig. 2, Tables 3, 4 and 6). Individuals with one or both parents artificially selected
554 (i.e. with tarsus length shifted from the population mean) were shown to produce fewer
555 recruits than unselected individuals (Table 2). However, when combining recruit pro-
556 duction and survival into a measure of individual fitness, the natural selection towards
557 pre-experimental means was only significant for males in the *low* population (Fig. 1B). In
558 the seven years after artificial selection, there was no significant natural selection toward
559 pre-experimental means (Fig. 1C and D). This points to the fact that the detectability of
560 a given strength of selection generally is strongly dependent on the magnitude of demo-
561 graphic stochasticity (Hersch and Phillips, 2004; Engen et al., 2012; Engen and Sæther,
562 2014; Haller and Hendry, 2014). Here, the demographic variance was found to be large
563 during the years of artificial selection (Table S6), compared to previous estimates for
564 house sparrows (Engen et al., 2007; Stubberud et al., 2017) and other small passerines
565 (Sæther et al., 2004). This was probably an effect of translocating individuals, which re-
566 duced population size (N) and may have affected the social structure in the populations.
567 Another effect of reducing N was necessarily a reduction of the population density during
568 the breeding season in the two experimental populations. The demographic framework
569 for estimating selection used in this study rest on the simplifying assumption of density-
570 independent vital rates (Engen et al., 2012). A previous study, including the present
571 study populations and other populations from the same area, found no effect of N on
572 population growth (ΔN) during the present study period (Stubberud et al., 2017). Thus,
573 the reduction of N should not have affected our results above the increased random vari-

574 ation in individual fitness among individuals (i.e. increased demographic variance).

575 When population mean phenotypes are stable over longer time periods, stabilizing se-
576 lection is a likely explanation (Charlesworth et al., 1982; Estes and Arnold, 2007; Uyeda
577 et al., 2011; Chevin and Haller, 2014; Haller and Hendry, 2014). Stabilizing selection
578 maintains the mean phenotype of fitness related traits at intermediate values of high
579 fitness (Lande, 1976, 1979; Arnold et al., 2001; Kinnison and Hendry, 2001; Sæther and
580 Engen, 2015). In this study, individual fitness was reduced in both directions from the
581 pre-experimental mean tarsus lengths, which suggests that tarsus length was moved away
582 from an adaptive fitness peak (Table 2). However, an alternative explanation may be that
583 tarsus length is constrained by genetic correlations with an unmeasured trait (Lande and
584 Arnold, 1983; Hansen and Houle, 2004, 2008; Morrissey et al., 2010). Then, both traits
585 could be kept from reaching their optimum in a balance of opposing directional selection.
586 This explanation would require that the genetic correlation was so strong that the applied
587 artificial selection also had a large effect on the unmeasured trait. While it is not possible
588 to conclusively exclude an effect of such an unmeasured trait, at least none of the other
589 traits in this study displayed significant changes in breeding values (see Table 6). Still,
590 the expected ubiquitous effect of stabilizing selection is rarely detected in empirical stud-
591 ies of contemporary populations (Kingsolver et al., 2001, 2012). One reason is the low
592 power to detect stabilizing selection in most studies with limited sample size (Kingsolver
593 et al., 2001; Haller and Hendry, 2014), an issue that increases with increasing demo-
594 graphic stochasticity (Engen et al., 2012; Engen and Sæther, 2014; Haller and Hendry,
595 2014). Stabilizing selection might also be hard to detect due to low phenotypic variance
596 around the peak, as less fit individuals continuously are removed, and the interference
597 of ecological mechanisms, such as competition for resources (Rueffler et al., 2006; Haller
598 and Hendry, 2014). Competition may lead to negative frequency-dependent selection,
599 where intermediate phenotypes experience the largest reduction in fitness (Rueffler et al.,
600 2006; Bolnick and Lau, 2008). Such mechanisms could lead to a flattening of the fitness
601 peak which reduces the possibility for detecting stabilizing selection, or in extreme cases
602 could cause disruptive selection (Haller and Hendry, 2014; Hendry, 2017). Frequency-

603 dependent selection may often arise under parasitism, predation, sexual selection, sexual
604 conflicts or asymmetric resource competition within species (Lande, 1980; Goldberg and
605 Lande, 2006; Hendry, 2017). However, there were no indications of such mechanisms in
606 the present study, where there was a clear reduction in fitness for selected individuals
607 throughout the study period (see Table 2).

608 A fluctuating environment might constantly induce small random changes in the phe-
609 notypic fitness optimum, such that in any year or period of years selection might be
610 directional (Arnold et al., 2001; Lande, 2007). During the period after artificial selection,
611 the directional selection in males of the *high* population was in the same direction as dur-
612 ing the artificial selection (Fig. 1D). Intuitively, one might think that artificially enlarged
613 males were at an advantage relative to unselected smaller males. However, *post hoc* tests
614 showed that the reduced recruit production of selected males relative to unselected males
615 was not significantly different between the two periods in the *high* population (2003-2006
616 vs 2007-2012, *selection status* \times *period*: $\chi^2 = 2.31$, $df = 2$, $P = 0.32$). In addition, the
617 estimated environmental variance was quite large (see Table S6 and previous estimates
618 in Sæther et al., 2004) and after maintaining long tarsus for 2-3 years, phenotypic and
619 genetic values returned towards the pre-experimental means (Fig. 2). Hence, the unex-
620 pected positive selection on tarsus length in males was probably due to environmental
621 fluctuations resulting in a brief period with selection for large body size.

622 The mating of unselected and artificially selected parents produced intermediate indi-
623 viduals with increased mean phenotypic values in the *high* population, but no change in
624 the *low* population (see Table 4). This gene flow between the unselected and selected seg-
625 ment of each population decreased the overall response to artificial selection. Gene flow
626 between wild populations under different selective regimes has repeatedly been suggested
627 as a possible constraint on the phenotypic response in heritable traits (e.g. Slatkin, 1973;
628 Storfer and Sih, 1998; Hendry et al., 2001; Postma and van Noordwijk, 2005; Postma
629 et al., 2007; Rice and Papadopoulos, 2009; Siepielski et al., 2013; Hendry, 2017). Hence,
630 the identification of spatially varying patterns of selection and evolutionary responses
631 in wild unmanipulated populations depends on our ability to distinguish individuals of

632 different origins. Failing to do so could be an important cause of mismatch between
633 expected and observed phenotypic response to selection.

634 The experimental populations gradually returned towards their pre-experimental mean
635 tarsus length and breeding values after the period of artificial selection ended (Fig. 2).
636 The return in breeding values was faster than expected by genetic drift alone (Table
637 6). Three interacting mechanisms are believed to be involved in this process: natural
638 selection counteracting the artificial selection, immigration, and recombination between
639 selected and unselected individuals. Provided the recorded strength of natural selection,
640 it would have taken a long time for natural selection alone to restore phenotypes in the
641 populations. Hence, immigration and recombination between selected and unselected
642 individuals were active drivers of changes in phenotypes and breeding values during the
643 period after artificial selection. The expected proportion of the genome in a randomly
644 chosen individual which was inherited from artificially selected ancestors decreased from
645 0.6-0.7 at the end of artificial selection to c. 0.25 at the end of the study period (Figure
646 S2). Thus, the proportion of individuals that were unselected in each cohort increased
647 towards the end of the study period and there were no selected individuals in the 2011
648 cohort (see Tables S9 and S10). The change in phenotype may be separated into a se-
649 lection differential and a transmission term using the Price equation (Price, 1970, 1972;
650 Frank, 2012; Engen et al., 2014; Queller, 2017). In these terms, the transmission term
651 was a large component in the return towards pre-experimental means. Still, the effect
652 of counteracting natural selection was important. Selected individuals produced 35-45
653 % less recruits than unselected individuals, thus the change in phenotype from the se-
654 lection differential during artificial selection was reduced (see Table 2). These effects on
655 the phenotypic change might be concealed in age-structured populations, where the final
656 evolutionary response to selection is delayed until the individuals under selection have
657 realized their lifetime reproduction (Hill, 1974; Engen et al., 2014).

658 Manipulating selection in the wild can yield novel insights into several aspects of
659 evolutionary dynamics in populations under natural conditions. We have demonstrated
660 that strong directional selection on heritable traits produce evolutionary responses in

661 accordance with well-known quantitative genetic models. However, we also illustrate the
662 potential for gene flow to impact the phenotypic trajectory of natural populations under
663 temporal or spatial variation in selection. Perturbing the phenotype away from their
664 natural mean had profound negative fitness consequences. Overall, the results provided
665 indications of a phenotype maintained by selection for an intermediate value subject to
666 environmental variation.

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947 **Tables**

Table 1: Individual selection category based on whether both, one or none of the parents had been artificially selected in two house sparrow populations in Norway. The populations were subject to artificial selection for long or short tarsus.

Selection category	Description
Selected	Both parents artificially selected
Unselected	No parent artificially selected
Intermediate	One parent artificially selected
Other	All other individuals

Table 2: Parameter estimates and 95% confidence intervals for models explaining the production of recruits over the years 2003-2012 in two house sparrow populations in Norway. The populations were subjected to artificial selection for long (*high*) or short (*low*) tarsus. The selection categories were unselected, intermediate and selected (see Table 1). Estimates are given relative to unselected females of age 1 (Intercept). Generalized mixed effects models were fitted with a Poisson error structure and a log link function. Models were fitted with a random intercept for individual identity.

	Estimate	Confidence interval	
		Lower	Upper
<hr/>			
(A) <i>High</i>			
Intercept	-0.23	-0.46	0.00
Selection category			
Selected	-0.43	-0.69	-0.16
Intermediate	-0.16	-0.43	0.10
Age 2	0.28	0.09	0.47
Male	-0.10	-0.32	0.12
<hr/>			
(B) <i>Low</i>			
Intercept	-0.05	-0.31	0.20
Selection category			
Selected	-0.58	-0.96	-0.21
Intermediate	-0.35	-0.70	-0.01
Age 2	-0.00	-0.28	0.27
Male	0.06	-0.17	0.30
Sel.status × age			
Selected × age 2	0.77	0.30	1.24
Intermediate × age 2	0.10	-0.40	0.59
<hr/>			

Table 3: Annual phenotypic change (slope \pm SE) in weighted means in three house sparrow populations in Norway. Two of the populations were subjected to artificial selection for longer (*high*) or shorter (*low*) tarsus in the years 2002-2005. In the period 2006-2012 the populations were monitored with no further manipulations. Permutation tests with 10 000 iterations were used to assess the significance of the estimated annual changes. Annual changes were estimated using linear regression, accounting for mean differences between sexes in phenotypes. * $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$.

	Years	
	2002-2006	2006-2012
<i>(A) High</i>		
Tarsus length	0.126 \pm 0.021***	-0.088 \pm 0.013***
Wing length	0.211 \pm 0.044***	-0.003 \pm 0.023
Body condition	0.009 \pm 0.001***	-0.003 \pm 0.001***
Bill length	0.042 \pm 0.015**	-0.039 \pm 0.009***
Bill depth	0.015 \pm 0.007*	-0.007 \pm 0.004
<i>(B) Low</i>		
Tarsus length	-0.112 \pm 0.020***	0.027 \pm 0.013
Wing length	-0.027 \pm 0.048	0.129 \pm 0.028***
Body condition	0.016 \pm 0.002***	-0.003 \pm 0.001***
Bill length	0.017 \pm 0.014	-0.009 \pm 0.008
Bill depth	-0.025 \pm 0.006***	0.008 \pm 0.005
<i>(C) Control</i>		
Tarsus length	-0.027 \pm 0.028	-0.035 \pm 0.016*
Wing length	-0.085 \pm 0.063	0.029 \pm 0.031
Body condition	0.005 \pm 0.002	-0.002 \pm 0.001
Bill length	0.016 \pm 0.022	-0.004 \pm 0.010
Bill depth	-0.010 \pm 0.010	-0.004 \pm 0.005

Table 4: Cohort phenotypic change (slope \pm SE) in arithmetic mean over the cohorts 2000-2005, for each of three selection categories in two house sparrow populations in Norway. Artificial selection on tarsus length was performed for longer (*high*) or shorter (*low*) tarsus on the pre-breeding populations in the years 2002-2005. Any sexual dimorphism in the traits was accounted for by including sex in the models. * $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$.

	Selection category		
	Selected	Intermediate	Unselected
<i>(A) High</i>			
Tarsus length	0.167 \pm 0.040***	0.127 \pm 0.047*	0.008 \pm 0.079
Wing length	0.178 \pm 0.074*	0.170 \pm 0.070	0.271 \pm 0.128
Body condition	0.012 \pm 0.002***	0.004 \pm 0.003	0.008 \pm 0.003
Bill length	-0.008 \pm 0.028	0.063 \pm 0.039*	0.061 \pm 0.034
Bill depth	0.020 \pm 0.012	0.002 \pm 0.014	0.029 \pm 0.018
<i>(B) Low</i>			
Tarsus length	-0.091 \pm 0.041*	0.057 \pm 0.042	0.002 \pm 0.044
Wing length	-0.068 \pm 0.070	-0.049 \pm 0.075	0.093 \pm 0.107
Body condition	0.018 \pm 0.004***	0.016 \pm 0.004***	0.014 \pm 0.003***
Bill length	0.056 \pm 0.024*	0.048 \pm 0.019	0.055 \pm 0.027
Bill depth	-0.023 \pm 0.015	-0.010 \pm 0.013	-0.024 \pm 0.017

Table 5: The \mathbf{G}_σ -matrix for three house sparrow populations (*high*, *low* and *control*) in the years 2002-2012 in Norway. The final cohort included in the analyses was 2011. Two of the populations were subjected to artificial selection for longer (*high*) or shorter (*low*) tarsus in the years 2002-2005. Posterior modes with 95% credibility intervals are given. All traits were SD-standardised prior to analyses, such that the matrices have heritabilities on the diagonal and genetic correlations in the off-diagonal elements.

	Tarsus length	Wing length	Body condition	Bill length	Bill depth
(A) <i>High</i>					
Tarsus length	0.396 (0.281,0.542)	0.144 (0.068,0.248)	0.064 (-0.035,0.146)	0.097 (-0.002,0.188)	0.156 (0.072,0.277)
Wing length		0.315 (0.225,0.447)	0.097 (0.031,0.190)	0.062 (-0.010,0.160)	0.106 (-0.008,0.189)
Body condition			0.408 (0.286,0.544)	0.125 (0.017,0.204)	0.115 (0.018,0.227)
Bill length				0.625 (0.469,0.734)	0.215 (0.112,0.321)
Bill depth					0.442 (0.314,0.626)
(B) <i>Low</i>					
Tarsus length	0.313 (0.229,0.436)	0.137 (0.080,0.219)	0.013 (-0.057,0.103)	0.120 (0.059,0.230)	0.113 (0.027,0.192)
Wing length		0.333 (0.243,0.412)	0.061 (-0.006,0.148)	0.075 (0.020,0.163)	0.078 (0.004,0.154)
Body condition			0.402 (0.272,0.524)	0.057 (-0.021,0.155)	0.170 (0.071,0.260)
Bill length				0.391 (0.246,0.506)	0.145 (0.049,0.233)
Bill depth					0.418 (0.310,0.578)
(C) <i>Control</i>					
Tarsus length	0.416 (0.260,0.625)	0.076 (-0.025,0.181)	-0.041 (-0.142,0.082)	0.094 (-0.034,0.234)	-0.059 (-0.163,0.089)
Wing length		0.289 (0.191,0.409)	0.033 (-0.039,0.141)	0.060 (-0.033,0.180)	0.037 (-0.047,0.150)
Body condition			0.154 (0.012,0.326)	-0.003 (-0.152,0.095)	0.086 (-0.015,0.215)
Bill length				0.458 (0.251,0.674)	0.097 (-0.054,0.227)
Bill depth					0.409 (0.225,0.626)

Table 6: Annual change in the weighted mean estimated breeding values for three house sparrow populations in Norway. Two of the populations were subjected to artificial selection for longer (*high*) or shorter (*low*) tarsus in the years 2002-2005. Stars indicates if the estimated changes are larger than expected by genetic drift alone. Posterior modes with 95% credibility intervals are given. * $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$.

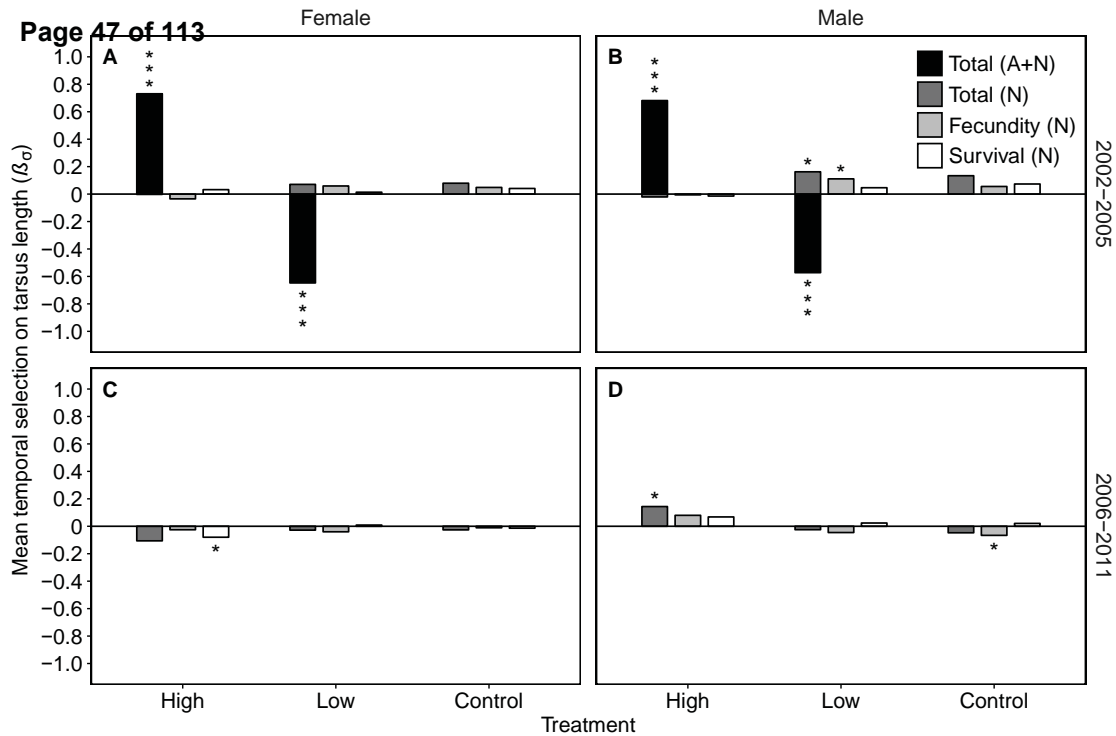
	Years	
	2002-2006	2006-2012
<i>(A) High</i>		
Tarsus length	0.110 (0.072, 0.152)***	-0.055 (-0.081, -0.026)*
Wing length	0.027 (-0.005, 0.064)	0.000 (-0.021, 0.026)
Body condition	0.050 (0.003, 0.083)	-0.008 (-0.037, 0.018)
Bill length	0.028 (-0.012, 0.070)	-0.013 (-0.043, 0.017)
Bill depth	0.035 (-0.002, 0.078)	-0.014 (-0.041, 0.012)
<i>(B) Low</i>		
Tarsus length	-0.103 (-0.137, -0.059)***	0.044 (0.021, 0.067)*
Wing length	-0.025 (-0.059, 0.005)	0.034 (0.016, 0.057)
Body condition	0.038 (0.000, 0.080)	-0.024 (-0.048, 0.001)
Bill length	-0.019 (-0.059, 0.019)	0.011 (-0.015, 0.034)
Bill depth	-0.019 (-0.063, 0.017)	0.003 (-0.023, 0.027)
<i>(C) Control</i>		
Tarsus length	0.003 (-0.051, 0.036)	-0.019 (-0.043, 0.019)
Wing length	-0.002 (-0.039, 0.033)	0.021 (0.001, 0.044)
Body condition	0.005 (-0.022, 0.044)	0.001 (-0.022, 0.019)
Bill length	0.011 (-0.030, 0.055)	0.006 (-0.020, 0.035)
Bill depth	-0.001 (-0.041, 0.045)	0.016 (-0.014, 0.044)

948 Figure legends

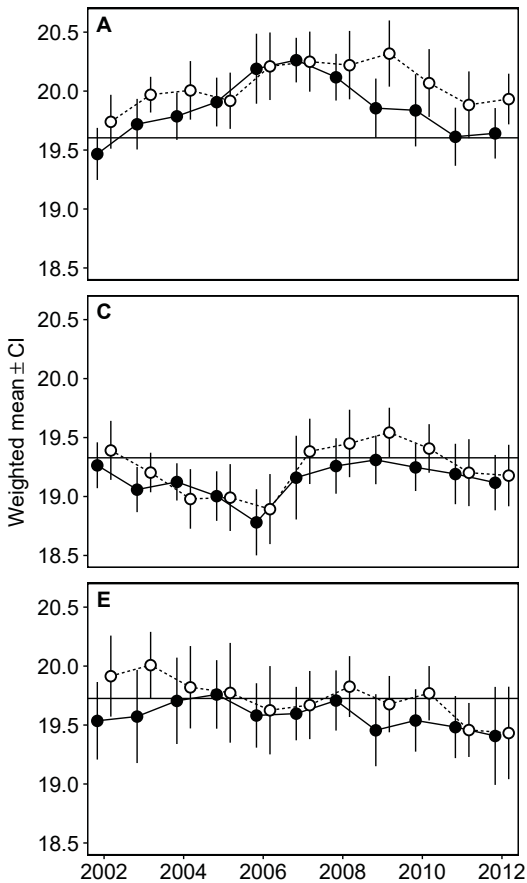
949 **Figure 1:** Temporal mean SD-scaled directional selection gradients (β_σ) over the periods
 950 2002-2005 and 2006-2011 for female and male house sparrows in each of three populations
 951 (*high*, *low* and *control*) in Norway. In the first period (**A**, **B**), two of the populations were
 952 subjected to artificial selection for long (*high*) or short (*low*) tarsus. In the subsequent
 953 period (**C**, **D**), all three populations were monitored with no artificial manipulations of
 954 the distribution of phenotypes. Selection was estimated including both artificial (A) and
 955 natural (N) selection, and natural selection was further decomposed into viability and
 956 fecundity selection. * $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$.

957 **Figure 2:** Annual estimates (weighted mean \pm 95% CI) of phenotypic (**A**, **C**, **E**) and
 958 genetic (**B**, **D**, **F**, estimated breeding value [EBV]) tarsus length (mm) in three house
 959 sparrow populations (*high* [**A**, **B**], *low* [**C**, **D**] and *control* [**E**, **F**]) in Norway. The *high*
 960 and *low* populations were subjected to artificial selection for longer (*high*) or shorter (*low*)
 961 tarsus before the breeding seasons in the years 2002-2005. Males (open circles, dashed
 962 lines) and females (solid circles and lines) were analysed together in the animal models,
 963 including sex as a fixed effect. EBV is are shown with solid circles and lines, while the
 964 stars and shaded areas are the expected EBV with 95% credibility intervals simulated
 965 under genetic drift alone. The horizontal lines in the left panels (**A**, **C**, **E**) are the mean
 966 tarsus length for each population across sexes in 2002.

967 **Figure 3:** Predicted and observed response to selection in two house sparrow populations
 968 in Norway. The populations were subjected to artificial selection for long (*high*) and
 969 short (*low*) tarsus in the years 2002-2005. The annual responses are averaged across
 970 sexes as sample sizes did not allow sex-specific **G**-matrices. During the period 2006-
 971 2011, populations were monitored without additional manipulations. (**A**) The complete
 972 annual response, which includes both survival of adults and recruitment of new individuals
 973 from known parents. (**B**) The partial annual response, includes only recruitment of new
 974 individuals from known parents. Unselected and intermediate individuals were excluded
 975 to estimate the observed responses.



Tarsus length (mm)



EBV for tarsus length

