MOLECULAR ECOLOGY

Sensitivity Analysis of Effective Population Size to Demographic Parameters in House Sparrow Populations

Journal:	Molecular Ecology	
Manuscript ID	MEC-16-1120.R1	
Manuscript Type:	Original Article	
Date Submitted by the Author:	n/a	
Complete List of Authors:	Stubberud, Marlene; Norges Teknisk Naturvitenskapelige Universitet Institutt for biologi; Norwegian University of Science and Technology, Dept. of Biology Myhre, Ane; Norwegian University of Science and Technology, Dept. of Biology Holand, Håkon; Norwegian University of Science and Technology, Dept. of Biology Kvalnes, Thomas; Norwegian University of Science and Technology, Dept. of Biology Ringsby, Thor Harald; Norwegian University of Science and Technology, Department of Biology, Centre for Conservation Biology Sæther, Bernt-Erik; Norwegian University of Science and Technology, Centre for Biodiversity Dynamics, Department of Biology Jensen, Henrik; Norwegian University of Science and Technology, Dept. of Biology	
Keywords:	Conservation Biology, Conservation Genetics, Population Genetics - Empirical, Population Ecology	

SCHOLARONE™ Manuscripts

1	Title page:
2	
3	Sensitivity Analysis of Effective Population Size to Demographic Parameters
4	in House Sparrow Populations
5	
6	Marlene Wæge Stubberud ^{1,2} , Ane Marlene Myhre ¹ , Håkon Holand ¹ , Thomas Kvalnes ¹ ,
7	Thor Harald Ringsby ¹ , Bernt-Erik Sæther ¹ , and Henrik Jensen ^{1,*}
8	
9	¹ Centre for Biodiversity Dynamics, Department of Biology, Norwegian University of
10	Science and Technology, NO-7491 Trondheim, Norway.
11	² Current address: Centre for Ecological and Evolutionary Synthesis, Department of
12	Biosciences, University of Oslo, NO-0316 Oslo, Norway.
13	
14	Key-words: Sensitivity analysis, population size, effective population size, two-sex demographic
15	model, population management.
16	
17	*Author for correspondence: Henrik Jensen, Centre for Biodiversity Dynamics, Department
18	of Biology, Norwegian University of Science and Technology, N-7491 Trondheim, Norway.
19	E-mail: henrik.jensen@ntnu.no; fax: $+47$ 73596100; office: $+47$ 73596949; mobile: $+47$
20	91897064.
21	
22	Running title: Sensitivity of effective population size

23 Abstract

The ratio between the effective and the census population size, N_e/N , is an important 24 measure of the long-term viability and sustainability of a population. Understanding 25 which demographic processes that affect N_e/N most will improve our understanding of 26how genetic drift and the probability of fixation of alleles is affected by demography. This 27 knowledge may also be of vital importance in management of endangered populations 28 and species. Here, we use data from 13 natural populations of house sparrow (Passer 29 domsticus) in Norway to calculate the demographic parameters that determine N_e/N . By 30 using the global variance-based Sobol' method for the sensitivity analyses, we found that 31 N_e/N was most sensitive to demographic variance, especially among older individuals. 32 Furthermore, the individual reproductive values (that determine the demographic variance) 33 34 were most sensitive to variation in fecundity. Our results draw attention to the applicability of sensitivity analyses in population management and conservation. For population 35 36 management aiming to reduce the loss of genetic variation, a sensitivity analysis may indicate the demographic parameters towards which resources should be focused. The 37 result of such an analysis may depend on the life history and mating system of the 38 population or species under consideration, since the vital rates and sex-age classes that 39 N_e/N is most sensitive to may change accordingly.

41 Introduction

42 A growing number of natural populations face threats originating from human activities (Primack 2010), many of which lead to increased fragmentation and decreased population 43 size (Frankham et al. 2010). Small populations are more vulnerable to extinction due 44 to increased inbreeding, more rapid loss of genetic variation due to genetic drift, and a 45decreased resilience when faced with environmental change (Lande et al. 2003; Legendre 46 et al. 2008). Processes affecting population size (N) (e.g. survival, reproduction, and 47 migration, Tuljapurkar & Caswell 1997), depend on both genes and the environment 48 (Caswell 2001), and the effective population size (N_e) influence evolutionary processes in 49 the population through rate of loss of genetic variation. By maximising both N_e and N, 50 51 the resilience to change is expected to be maintained in the population, and the risk of 52 extinction will then decrease (Frankham et al. 2010; Allendorf et al. 2012). In population genetics, knowledge of N_e , which depends on both ecological and genetic 5354 factors, is crucial to make both short- and long-term predictions regarding loss of genetic variation due to drift as well as the probability of fixation of advantageous alleles due to 55 selection (Shaffer 1981; Nunney & Elam 1994; Palstra & Fraser 2012). As a consequence, 56 it is important to identify the parameters that influence N_e most, i.e. the parameters 57 to which N_e is most sensitive to (Caswell 2001; Saltelli et al. 2004). N_e is defined as 58 59 the size of a Wright-Fisher ideal population that experiences the same rate of genetic drift and loss of heterozygosity as the observed population (Wright 1931). Such an 60 ideal population is a conceptual panmictic population with constant size, consisting of 61 monoecious diploid individuals that have discrete generations, Poisson distributed family 62sizes, and no selection or mutations in autosomal loci (Nunney 1993; Caballero 1994; 63 Wang & Caballero 1999; Kalinowski & Waples 2002; Frankham et al. 2010). Natural 64 populations usually violate the ideal conditions in several ways (Harris & Allendorf 1989; 65 Wang & Caballero 1999; Frankham et al. 2010), and most populations therefore behave as 66 if they were a lot smaller than their census size (Halliburton 2003; Freeman & Herron 2007; 67

68 Frankham et al. 2010). Of all deviations from the ideal population, variation in N has been shown to often reduce N_e most (Kalinowski & Waples 2002; Engen et al. 2005b; Frankham 69 et al. 2010), followed by variation in family size, and skewed sex ratios (Caballero 1994; 70 Frankham 1995; Frankham et al. 2010). Other factors that affect N_e include mating 71 system, overlapping generations, generation time (Nunney 1993; Engen et al. 2005b), 72and population age-structure. Environmental stochasticity indirectly affect the effective 73 population size by amplifying fluctuations in population size (Lande et al. 2003). Random 7475 variation in survival and reproduction within years, termed demographic stochasticity, may also cause deviations from an ideal population and reduce N_e (Lebreton et al. 1992; 76 Ardren & Kapuscinski 2003; Frankham et al. 2010; Myhre et al. 2016). Demographic 77 stochasticity, measured by demographic variance (σ_d^2) , increases the rate of genetic drift, 78 and is especially important for small populations (Lande et al. 2003; Engen et al. 2005b; 79 Shpak 2007). 80 Following the approach of Pollak (2000) considering the dynamics of a subgroup of 81 82 individuals bearing a rare neutral allele, Engen et al. (2005b) derived a formula for the effective population size per generation based on the demographic variance, N_e 83 $N/(\sigma_{dg}^2T)$, where the subscript g indicates that the demographic variance also has a 84 genetic component due to Mendelian segregation, and T is the generation time. Based 85 on the realization that the dynamics of age-structured density-independent populations 86 could well be approximated by the dynamics of the total reproductive value (Engen et al. 87 2007a), Engen et al. (2010) derived formulas for the ratio of effective to actual size. 88 89 This provided an extension of previous models (Felsenstein 1971; Hill 1972, 1979) for genetic drift in age-structured populations without assuming a stable age-distribution or 90 no environmental fluctuations. 91 92 In this study, data from a long-term study on Norwegian populations of house sparrow (Passer domesticus) was used to estimate the variance N_e and the effective to census 93 population size ratio, N_e/N (Engen et al. 2010). The demographic parameters used to 94

estimate N_e include generation time and demographic variance, where the latter is given 95 by the sum of sex-age specific variances in individual reproductive values weighted by 96 the stable sex-age distribution. A sample-based Sobol' method was used for a global 97 variance-based sensitivity analysis, to determine the sensitivity of N_e/N to each of the 98 demographic parameters. The principle of global analyses is that all inputs are varied 99 simultaneously, the entire parameter space is explored, and any interactions between input 100 101 parameters are accounted for (Sobol' 2001). We believe management efforts to regulate populations should be focused on the demographic parameters that N_e/N is most sensitive 102 103 to, and that this study will provide insight on identification of these parameters.

104 Materials and methods

105 Study system

The data in this study were collected from 13 populations of house sparrow located along 106 107 the Norwegian coast (Fig. 1). Six of the northernmost populations were located in a 108 long-term study system consisting of an insular metapopulation of house sparrows at the 109 Helgeland coast (66°N, 13°E). The remaining seven study populations were located on 110 the mainland, and on islands to the south of this insular metapopulation. See Ringsby 111 et al. (1999), Ringsby et al. (2002) and Pärn et al. (2012) for more details about the study 112 area. The dataset included 4074 individuals, with an average of seven years of data per population 113 and with a range of two to twenty years. The data encompassed populations with highly 114 different histories. One of the populations, on the island Aldra (population 1 in Fig.), was 115 116 founded in 1998 by one female and three males, and suffers from substantial inbreeding (Billing et al. 2012), while the population at Ytre Kvarøy (population 13 in Fig. 1) went 117 extinct in 2000 (Ringsby et al. 2006). Two other populations, Leka and Vega (population 118 6 and 12 in Fig. 1), were part of an experiment in 2002-2005 (Kvalnes et al. in review) 119 where approximately 60 % of the individuals in each population were removed each year 120 121 following artificial selection on tarsus length. Individuals with short or long tarsi were removed on Leka and Vega, respectively (Kvalnes et al. in review). These four islands 122 123 were included to increase the range of variation, within natural limits, in the dataset.

124 Study species

The house sparrow is a socially monogamous species, with some extra-pair mating (Jensen et al. 2008), where both parents contribute to brooding and feeding of nestlings (Ringsby et al. 2009). In the study area, the breeding season is from May-August, and each pair

produce 1-3 clutches with an average of five eggs per clutch during this time (Ringsby et al. 128 2002; Husby et al. 2006). Only 15-20 % of fledglings survive their first winter (i.e. still 129 alive after 1 February) to become recruits in the next years' breeding population (Ringsby 130 131 et al. 2002). The average generation time is 1.97 years in this system (Jensen et al. 2008) with an annual adult survival rate at approximately 50 % (Ringsby et al. 1999). After 132 the post-natal and post-breeding moult, adult and juvenile house sparrows are impossible 133 to discriminate (Anderson 2006). Hence, all full-grown individuals were assumed to be 134 hatched in the most recent completed breeding season upon first capture (see Table S1, 135 136 Supporting information). For islands with many years of data this assumption should be of little concern, but it will affect the estimated generation time in populations with 137 few years of data (see *Results* and *Discussion*). The natal dispersal rates among islands 138 139 within the metapopulation system are low; only around 10 % of the juveniles disperse, and adult dispersal is negligible (Altwegg et al. 2000; Pärn et al. 2009, 2012). Among the 140 dispersers, approximately 60 % of them travel distances shorter than 13 km (Tufto et al. 141 2005). 142

143 Data collection and handling

Most of the data was collected during the breeding season. Adults and fledged juveniles 144 145 were caught using mist nets, while nestlings were caught in the nests. All individuals were marked with a numbered metal ring, and a unique combination of three coloured plastic 146 147 rings for later identification in the field (Ringsby et al. 2002; Jensen et al. 2004). A blood 148 sample of 25 µl was drawn from the brachial vein underneath the wing, and provided 149 the DNA necessary to genotype individuals on 14 unlinked polymorphic microsatellite markers for genetic parentage analyses (Jensen et al. 2003). Genetic pedigrees for the 150 populations Handnesøy, Linesøya, Løkta, Rånes and Røvass, were established (see Table 151 152 S2 and Parentage analyses in Supporting information), while genetic pedigrees for the remaining populations were already available, see Jensen et al. (2003), Billing et al. (2012), 153

Jensen et al. (2013) and Holand et al. (2015). The pedigrees in tandem with individual capture and observation data, provided information on annual survival, fecundity and census population size estimated as the number of adult individuals captured and/or observed before or during breeding season in a given year. These estimates are assumed to be accurate as the annual capture rates in the study system are above 70% and usually close to 100% (Jensen et al. 2006, 2013).

160 The females and males in the dataset were split into two age classes each, (1) yearlings of age one, and (2) all individuals of age two years or older. In total there were four 161 groups, hereafter addressed as sex-age classes, denoted by $i = (f_1, f_2, m_1, m_2)$ for females 162 and males in age class 1 and 2. The pooling of individuals into a terminal age class was 163 164 done to exclude the effect of small sample sizes of older individuals. This approach has little effect on estimates of N_e , because the vital rates vary little with age (Engen et al. 165 166 2010). The sex-ratio at birth (q) given as the proportion of females, was assumed to be 0.5 (Husby et al. 2006), hence, the total number of recruits divided by two could be used 167 in the calculations (Engen et al. 2010). In accordance with Engen et al. (2007, 2009), we 168 assumed that the population dynamics were density-independent, and that there were no 169 170 temporal autocorrelations in vital rates. All data handling and statistical analyses were 171 done in the software R version 3.1.3 (R Core Team 2016).

172 Demographic parameters

For all populations, N_e/N was estimated following the method for age-structured populations with two sexes in Engen *et al.* (2010):

$$\frac{N_e}{N} = \frac{1}{\sigma_{dg}^2 T},\tag{1}$$

where σ_{dg}^2 is the population specific demographic variance, and T is the generation time. The subscript g in σ_{dg}^2 denotes that the demographic variance has a genetic component due to Mendelian segregation in diploid species, as explained in Engen *et al.* (2005b). This definition of variance N_e has both theoretically and by simulations been shown to predict the correct amount of genetic drift (Engen *et al.* 2005b, see also Myhre *et al.* 2016). To calculate σ_{dg}^2 , we used data on individual survival and fecundity to build the expected population projection matrix, \mathbf{l} , for each population:

$$\mathbf{l} = \begin{bmatrix} \frac{1}{2}q\bar{B}_{f_1} & \frac{1}{2}q\bar{B}_{f_2} & \frac{1}{2}q\bar{B}_{m_1} & \frac{1}{2}q\bar{B}_{m_2} \\ \bar{J}_{f_1} & \bar{J}_{f_2} & 0 & 0 \\ & & & & \\ \frac{1}{2}(1-q)\bar{B}_{f_1} & \frac{1}{2}(1-q)\bar{B}_{f_2} & \frac{1}{2}(1-q)\bar{B}_{m_1} & \frac{1}{2}(1-q)\bar{B}_{m_2} \\ & & & & \\ 0 & 0 & \bar{J}_{m_1} & \bar{J}_{m_2} \end{bmatrix}$$

$$(2)$$

184 The matrix included the mean number of total recruits produced by individuals in each of the sex-age classes (\bar{B}_i) and the mean specific survival probabilities of each sex-age class 185 (\bar{J}_i) , where i signifies the four different sex-age classes. The fecundities are multiplied by 186 187 0.5 to account for the Mendelian segregation as offspring receive half its genes from each 188 parent. As mentioned, q is the proportion of females, so 1-q is the proportion of males at birth, which in this study were assumed to be equal (q = 0.5). The asymptotic population 189 190 growth rate (λ) was calculated as the real dominant eigenvalue of I. The left and right eigenvectors of l, u and v, are the stable sex-age distribution and reproductive values of 191 each sex-age class i, provided that these are scaled so that $\sum u_i = 1$, and $\sum u_i v_i = 1$ 192 (Caswell 2001). Both **u** and **v** are conceptual values for a population at equilibrium, 193 such that when the population has obtained its stable sex-age distribution, the census 194 population size equals the total reproductive value of the population (Felsenstein 1971; 195 Engen et al. 2010). 196

197 The number of offspring from males and females must necessary be exactly the same.

198 Hence, there are constraints on the vital rates in the matrix I to ensure that the asymptotic growth rate of the male (λ_m) and female (λ_f) subpopulation are exactly the same (conditioned 199 200 on the sex ratio at birth, q). When this is true we have that $\lambda = \lambda_f = \lambda_m$, and this was achieved by scaling the recruit production by males $(\bar{B}_{m_1}, \bar{B}_{m_2})$ by a constant, c. We 201 introduced the constant c to the Euler-Lotka equation, $cq \sum \lambda^{-i} l_i m_i = 1$, and solved it 202 203 using the Newtons method for the growth rate of the female subpopulation (c=1). Then the male growth rate was set equal to λ_f and the equation solved for c with q replaced 204 by q-1. The R-package lmf, version 1.2 (Engen et al. 2012) was used to calculate λ , ${\bf u}$ 205 206 and \mathbf{v} .

To estimate σ_{dg}^2 , the individual contributions to the future population growth was needed. This can be calculated as the individual reproductive value (Engen *et al.* 2009) which for an individual j in sex-age class i, is defined as:

$$W_{ij} = J_{ij}v_{i+1} + \frac{1}{2}B_{ij}v_1 + \frac{1}{2}B_{ij}v_3.$$
 (3)

The individual reproductive value gives the annual individual contribution to the total 211reproductive value of the population the following year. Here, there are two age classes 212 per sex, and i = 1, 2 represent females of age 1 and 2, respectively, while i = 3, 4 represent 213 males of age 1 and 2, respectively. J is an individual's own survival (1 if it survives, 214 otherwise 0), B is the number of recruits produced, and v(i+1), v_1 and v_3 is the age-sex 215 216 specific reproductive values, with recruiting females and males in sex-age class 1 and 217 3. Since q = 0.5, half of the recruits were expected to be female and half to be male, 218 which gave B/2. The value v_{i+1} , could due to the number of age classes in this case, maximally take the value v_2 for females, and v_4 for males. Since W_{ij} was calculated from 219 220 observed data on viability and fecundity, the annual demographic variance for each sex-age class $(\sigma_{dgi}^2(t))$, will not only capture variance due to demographic stochasticity, but also 221 sampling error and variance governing selection pressures. $\sigma_{dqi}^2(t)$ can be estimated as the 222

223 sum of squares (Engen et al. 2009):

$$\sigma_{dgi}^{2}(t) = \frac{1}{m_{ti} - 1} \sum_{i=1}^{m_{ti}} \left(W_{ij} - \bar{W}_{ti} \right)^{2}, \tag{4}$$

where m_{ti} is the total number of individuals in sex-age class i in year t, and \bar{W}_{ti} is the mean value of the W_{ij} in year t. The sex-age class specific demographic variance, σ_{dgi}^2 , is the weighted mean of $\sigma_{dgi}^2(t)$ over all years with $m_{ti} - 1$ as weights:

$$\sigma_{dgi}^{2} = \frac{\sum \sigma_{dgi}^{2}(t)m_{ti} - 1}{\sum m_{ti} - 1}$$
 (5)

The total demographic variance of the population, σ_{dg}^2 , is the weighted mean of σ_{dgi}^2 with weights u_i ,

$$\sigma_{dg}^2 = \sum \sigma_{dgi}^2 u_i. \tag{6}$$

The population specific generation time, T, was calculated as the mean age of all adults with recruits in each population (Lande *et al.* 2003). The N_e/N ratios of all populations were calculated by setting the values of σ_{dg}^2 into equation (1), with both the population specific T, and the average T across multiple populations and years in the study system (T = 1.97, Jensen *et al.* 2008).

237 Sensitivity analysis

A global variance-based method was used in the sensitivity analysis as it provides quantitative measures, is model independent, and there are no assumptions about linearity, monotonicity or independence of inputs (Saltelli & Annoni 2010). This approach with few assumptions suits biological data well, as biological parameters often are dependent on each other (Caswell 2001). The variance-based method can be illustrated with a general model: Y = f(X), where Y is the output with an unconditional variance V(Y), f is the function, and V(Y) are the inputs. In general, global sensitivity analyses are performed as follows: (i) the model is defined, (ii) the input parameters X are assigned probability density functions, from which (iii) they are sampled randomly, and (iv) the relative influence of the input parameters on the output is assessed by the preferred method (Chan et al. 1997). In this study, we chose the Sobol' method as it has been proven robust (Tang et al. 2007; Yang 2011).

250 Sobol's method

Sobol's method (Sobol' 1990, 1993) use Sobol' indices, or sensitivity indices, to quantitatively measure the importance of inputs. It is based on variance decomposition of the total variance, var(Y) = V, into partial, or conditional variances, $var(Y|x_p) = V_p$. The smaller the conditional variance is, but the larger influence it has on the total variance, the more important the input is. Ranking of inputs is most easily done by estimating the first order indices (Sobol' 2001):

$$S_p = \frac{V_p}{V},\tag{7}$$

where the first order index S_p , is the main effect of input parameter p. However, these indices do not include possible interactions among inputs, and might lead to erroneous ranking. To account for the effect of possible interactions, it is preferable to use the total-order index S_{Tp} , which is calculated as:

$$S_{Tp} = 1 - \frac{V_{\sim p}}{V} = 1 - S_{\sim p},\tag{8}$$

where $\sim p$ is all input parameters, except p (Homma & Saltelli 1996; Chan et~al. 1997).

The interaction effect of a parameter, $\Delta S = S_{Tp} - S_p$ (Nossent & Bauwens 2012), was

considered significant if $\Delta S \geq 0.05$ (see Chu-Agor et~al. 2011). The input parameters for

the analysis were sampled from their respective distributions with a Sobol' quasi-random

sampling scheme (Saltelli et~al. 2010). Each parameter was sampled n = 5000 times, and

bootstrapped 10,000 times to ensure convergence of the indices (Saltelli 2002; Yang 2011).

269 Parameter distributions

270 The R-package fitdistrplus (Delignette-Muller & Dutang 2015) was used to evaluate each parameters' density function which were used in the sensitivity analyses. Due to few 271272 data points for most parameters (because we had data from 13 populations), the best parameter distribution was not always clear. In tandem with considering the biological 273 process underlying the parameter, a suitable distribution was chosen (see Table S3, 274 Supporting information, for information on the distribution chosen for each parameter). 275 All parameters, except survival that is binomially distributed, were evaluated as continuous 276 277 due to the high number of individuals and the long timespan of the dataset.

278 Estimation of indices

To estimate the sensitivity indices, the *soboljansen* estimator in the R-package *sensitivity* 279 version 1.11 (Pujol & Janon 2015) was used. The soboljansen estimator has been shown 280 to be a computationally efficient estimator with low mean absolute error (Saltelli et al. 281 2010), and is suitable for large first-order indices, and large and small total indices (Pujol & 282 283 Janon 2015). All inputs for the estimation of the indices were sampled from the parameter 284 distributions (Table S3, Supporting information). We used a hierarchical approach in our sensitivity analysis to better understand what parameters N_e/N was most sensitive to. 285 First we examined the sensitivity of N_e/N to T and σ_{dg}^2 (eqn 1). Then generation time 286 287 was fixed at the average generation time in the study system (T = 1.97), and we examined the sensitivity of N_e/N to σ_{dg}^2 's components σ_{dgi}^2 and u_i (eqn 6). Since a stochastic annual 288 289 function was too complex for the chosen sensitivity estimator, the annual input parameter $\sigma_{dqi}^2(t)$ (eqn 4) was not included in the sensitivity analysis. Instead, our sensitivity analysis 290 continued at the individual level (eqn 3), where the sensitivity of W_{ij} to J_{ij} and B_{ij} was 291 292 evaluated, with the v_i 's fixed at their average values.

293 Results

294 Demographic parameters

The demographic parameters used to estimate N_e/N , are summarised in Tables 1, 2 and 295 S4 (Supporting information). The mean population size over all years of data varied from 296 297 19 to 170 individuals (Table 3), and annual population sizes ranged between 4 and 336 (see Table S1, Supporting information). 298 The population specific generation time, T, ranged from 1.20 to 2.39 years, with a mean of 299 1.72 years (Table 1). T was highest (> 2.1 years) for Aldra, Gjerøy and Nesøy, and lowest 300 (< 1.5 years) for Handnesøy, Linesøya, Løkta, Rånes and Røvass. The relatively large 301 302 range in T among these populations, are probably partly due to the length of data series; the four populations with low T all had n < 4 years of data (Table 3). The asymptotic 303 population growth rate was positive $(\lambda > 1)$ in four populations (Aldra, Leka, Rånes and 304 305 Vega), while for all other populations the estimated growth rate was negative ($\lambda < 1$). The demographic variance, σ_{dg}^2 , ranged from 0.62 to 2.98 in different populations, with a 306 mean of 1.45 (Table 1). Four populations had a higher σ_{dg}^2 than the overall mean, and 307 three of them were populations with special demographic histories: the inbred population 308 309 of Aldra, and the two artificially selected populations Leka and Vega. Rånes also had a high σ_{dg}^2 ($\sigma_{dg}^2 = 2.64$), but this may be due to chance, as there were only 2 years of 310 311 data for this population (Table 3). For the sex-age class specific demographic variance, σ_{dgi}^2 , both the mean and the variance were highest in the two oldest sex-age classes (Table 312 2). The highest value of σ_{dgi}^2 was found for older males on Rånes (Table S4, Supporting 313 information). 314 315 Both the stable sex-age distribution, u_i , and the sex-age class specific reproductive values, 316 v_i , had estimate means that were highest in age class 2 for both sexes (Table 2). Moreover, the variance of the estimates were very low (< 0.1) for all classes. In general (12 out of 13 317 populations) one or both sexes in age class 2 had larger **u** values than age class 1 (Tables 318

- 319 2 and S4, Supporting information). For v, there was a tendency for age class 2 to have
- 320 higher values than age class 1 for one or both sexes (11 out of 13 populations; Tables 2
- 321 and S4, Supporting information).
- 322 For survival, J, the mean was approximately the same (≈ 0.50) for all sex-age classes, and
- 323 the difference between the highest and lowest mean value was small ($\Delta J = 0.03$, Table 2).
- 324 Furthermore, the variance was the same for all age classes (Table 2). For the fecundity,
- 325 B, the highest mean value was found in older females ($f_2 = 0.94$), while it was lowest for
- 326 young males $(m_1 = 0.69)$. Age class 2 had the highest mean values for both sexes, but
- 327 the difference between age classes was smaller for females ($\Delta B = 0.11$), than for males
- 328 ($\Delta B = 0.19$). Finally, the highest variance in B was found in older females, whereas the
- 329 lowest was found in young males (Tables 2 and S4, Supporting information).

$330 N_e/N$

- 331 In general, N_e/N calculated with T=1.97 tended to be lower than when calculated
- 332 with the population specific T (range of N_e/N (T = 1.97): 0.17-0.82, range of N_e/N
- 333 (population specific T): 0.22-1.35), but the opposite was true for three populations with
- 334 high population specific T. Hestmannøy and Indre Kvarøy had population specific T close
- 335 to T = 1.97, hence the ratio was approximately the same using either estimate (Table 1).

336 Sensitivity analysis

- 337 In the first sensitivity analysis of N_e/N , with input parameters being the population
- 338 specific estimates of σ_{dg}^2 and T, the demographic variance was found to be the most
- 339 important parameter (Fig. 2, Table S5, Supporting information). The total order indices
- 340 showed that N_e/N was significantly more sensitive to σ_{dg}^2 than to T (σ_{dg}^2 : 95% CI
- 341 [0.671, 0.795], T: 95% CI [0.259, 0.316]) (Fig. 2). Because $\Delta S < 0.05$, interactions
- 342 between the input parameters were not likely to be important (Table S5, Supporting

343 information).

344 In the second sensitivity analysis of N_e/N with input parameters σ_{dqi}^2 and u_i (eqn 6), the demographic variance of age class 2 was most important (Fig. 3, Table S6, Supporting 345 information). In particular, N_e/N was most sensitive to the demographic variance of 346 males in age class 2, followed by females in the same age class. The total order indices 347 of these two input parameters, σ_{dgm2}^2 and σ_{dgf2}^2 , were significantly different (m_2 : 95% CI 348 $[0.513, 0.576], f_2: 95\%$ CI [0.421, 0.477]; Fig. 3). The total order index for males in age 349 class 1 was not significantly different from females in the same age class (m_1 : 95% CI 350 $[0.035, 0.044], f_1: 95\%$ CI [0.037, 0.046]; Fig. 3). Both σ_{dgf2}^2 and σ_{dgm2}^2 had significant 351 interaction effects ($\Delta S \geq 0.05$) between parameters (Table S6, Supporting information), 352 which might be related to the importance of having individuals of both sexes in age class 353 2, due to their higher reproductive success and the socially monogamous mating system 354 of the house sparrow. 355 In the third level of sensitivity analysis, the response variable was the individual reproductive 356 value W_{ij} for a given sex-age class, and J_{ij} and B_{ij} were input parameters. In all four 357 358 analyses, fecundity, B, was significantly more important than the other input parameters 359 as none of the 95% CI's overlapped (Fig. 4). Interactions between the input parameters were only found to be significant ($\Delta S \geq 0.05$) for age class 1, which is possibly explained 360 by the importance of surviving to age class 2, which have higher reproductive success. 361 362 To investigate whether the highest sex-age specific demographic variance of older males on Rånes (Table S4, Supporting information) affected the results, we re-analysed the 363 364 data when this population was removed. We also re-analysed the data with subsets that 365 excluded populations with less than 4 years of data, the population that went extinct, and the two populations that had undergone selection. These analyses did not change 366 the conclusions from the main sensitivity analyses (results not shown). Nevertheless, 367 including as many populations as possible is important to make realistic assumptions 368 369 regarding the parameter distributions. We also believe the variation observed in the 13

- 370 populations reflects true variation as natural populations are founded, go extinct, and
- 371 undergo strong selection events.



372 Discussion

We found that N_e/N was most sensitive to the demographic variance, σ_{dg}^2 (Fig. 2), 373 indicating that variation in individual fitness within years is important for the process of 374 genetic drift in this system. More specifically, N_e/N was most sensitive to the demographic 375 variance of individuals in the terminal age class, especially the male's (Fig. 3). When we 376 decomposed σ_{dgi}^2 into survival and fecundity, it was found that the individual reproductive 377 values, W_{ij} , and implicitly N_e/N (Engen et al. 2010), was most sensitive to fecundity in 378all four sex-age classes (Fig. 4). As the contribution to the total demographic variance 379 380 of the population show large age-specific variation (Sæther et al. 2013), this indicates that age-dependence in demographic traits strongly affect the genetic drift in natural 381 382 populations.

$383 N_e/N$ and sensitivity analysis

The mean values of N_e/N (Table 1) were similar to the mean values estimated in six 384 house sparrow populations in the same study system ($N_e/N = 0.37$, and $N_e/N = 0.50$ 385 with $\lambda = 1$, Engen et al. 2007b). Furthermore, Engen et al. (2010) used the same approach 386 and estimated a $N_e/N = 0.69$ in a growing population of Finnish Siberian jays (*Perisoreus* 387 infaustus). N_e/N of different passerines has been found to range from 0.09-0.80, with an 388 average of 0.49 (see O'Connor et al. 2006, Table 2). In a review by Frankham (1995), it 389 was found that the average N_e/N in birds and mammals was 0.37 and 0.47, respectively. 390 Hence, our N_e/N ratios (calculated with T=1.97) are well within the range of N_e/N 391 392values found in mammals and other birds. There are many methods to estimate N_e/N , and as shown by Schmeller & Merilä (2007) the estimates can vary a lot between methods. 393 For example, in a previous study we used genetic data and methods to estimate N_e in a 394 395 set of insular house sparrow populations (some of which were also included in the present study), and found that genetic estimates of N_e in general were similar to, or even larger 396

than N (Baalsrud et al. 2014). This pattern was likely due to the genetic consequences

397

that immigration had in the local populations (see also Gilbert & Whitlock 2015). Here 398 399 we used annual demographic estimates to ensure correct link between N_e and N (Waples 2005; Palstra & Fraser 2012), and used only the individuals assumed to be part of the 400 401 active breeding population, since these are the individuals that ultimately determine N_e (Palstra & Fraser 2012). Importantly, the current study and other studies that use similar 402 demographic methods to estimate N_e/N show that N_e/N can vary between populations 403 of the same species in a restricted geographic area (Table 1; Kaeuffer et al. 2004; Cutrera 404 405 et al. 2006; Engen et al. 2007b; Schmeller & Merilä 2007). This is also true for studies that have used genetic estimates of N_e and N_e/N (e.g. Palstra & Fraser 2012; Prado-Martinez 406 et al. 2013; Baalsrud et al. 2014). 407 We found that N_e/N was significantly more sensitive to σ_{dg}^2 than T (Fig. 2). 408 strongly suggests that variation among individuals in their contribution to future breeding 409 populations, both directly through survival, and indirectly through production of recruits, 410 affects the effective size of populations. It is also caused by the large age-dependent 411 variation in the contribution to the demographic variance found in most populations 412413 (Sæther et al. 2013). Our result is consistent with other studies, both theoretical (e.g. Nomura 2002; Hedrick 2005) and empirical (e.g. Ardren & Kapuscinski 2003; Kaeuffer 414et al. 2004; Araki et al. 2007), where demographic variance and variation in reproductive 415 success was found to influence N_e/N most. Lee et al. (2011) showed theoretically that for 416 populations with short generation time (i.e. T < 3, as studied here), changes in mating 417 418 system realized through changes in male reproductive success and hence demographic variance, markedly changed the N_e/N ratio, irrespective of whether or not there were 419 persistent differences in male quality. When demographic variance increased, N_e/N 420 421 decreased, which is in accordance with our results (Table 1). We used a demographic model to estimate N_e/N , but this ratio has also been estimated by genetic methods 422 423 in (partly) the same study system (Baalsrud et al. 2014). In that study, Baalsrud et al. (2014) found that demographic characteristics of the house sparrow populations were able 424

425to explain 31-71% of the observed variance in N_e/N based on different genetic estimators. Importantly, sex ratio, which is closely linked to variation in reproductive success and 426 demographic variance, was found to affect N_e/N . Comparing our estimates of σ_{dg}^2 and 427 T with the harmonic mean of the preferred genetic estimator (the LDNE-estimator) by 428 Baalsrud et al. (2014) shows a high correlation for σ_{dg}^2 (r = -0.846), and a modest 429correlation for T (r=0.185). The mean value of σ_{dg}^2 found here (1.45, Table 1) is 430 similar to σ_{dq}^2 's found in other bird species with similar life history to the house sparrow, 431e.g. blue tit (*Cyanistes caeruleus*) ($\sigma_{dg}^2 = 1.70$, Sæther *et al.* 2004) and great tit (*Parus* 432 major) ($\sigma_{dg}^2 = 1.14$, Engen et al. 2003). It is also worth noting that in the current 433study, demographic variance was positively correlated with the mean contribution to the 434 next generation, i.e. the asymptotic population growth rate, λ (Table 1). Although the 435 436 sensitivity of N_e/N to λ could not be evaluated directly as it is not a parameter in the equations used to calculate N_e/N here, the relationship with demographic variance may 437 438 suggest that the mean contribution of individuals to the next generation has an important role on the effective size of a population. 439The mean generation time found in our study (Table 1), was lower than the average 440 441 generation time (T = 1.97) found by Jensen et al. (2008) in the same study system. 442 Since the age of individual house sparrows can only be determined if they have been marked as nestling or juvenile, and given the 50% annual adult mortality, the true age 443 of most ($\approx 94\%$) individuals is known after 4 years. For the 5 populations with less 444 than 4 years of data, the mean was 1.32 years, while for the 8 populations with more 445 446 than 4 years of data, the mean was 1.97 years, which equals the value found by Jensen et al. (2008) in the same study system. Underestimating T may lead to an upward 447bias in the estimate of N_e/N , which argues that using the average T of a species may 448 be more appropriate when few years of data is available. Although N_e/N was more 449 sensitive to the demographic variance than to generation time, we found that the latter 450 was far from unimportant. Indeed, generation time is considered one of the important 451 452factors affecting the between-life-histories variation in effective population size (Nunney 453

1991, 1993; Caballero 1994). Theoretically, the contemporary ratio N_e/N decreases (or

sometimes increases) with increasing generation time, before it approaches a value of 0.5 454 (Nunney 1993; Caballero 1994; Lee et al. 2011, but see Waite & Parker 1996). This 455relationship may be caused by the reduction in variance in reproductive success, and/or 456 change in other life history traits often accompanied by an increase in generation time 457(Waples et al. 2013). Since most of these findings are based on different life histories 458 aligned along the fast-slow continuum, they encompass a much greater span of generation 459 time than experienced within most species. Actually, a change in T within the limits of 460 our values is not expected to create great changes (> 0.1) in N_e/N (Nunney 1993; Lee 461 et al. 2011), unless there are permanent differences in the mating success of individuals 462(Lee et al. 2011). Taking this into consideration, N_e/N was almost surprisingly sensitive 463 464 to variation in generation time. This may be explained by the use of population specific estimates of T to define the distribution of which samples were taken in the first sensitivity 465 466 analysis ($T \sim N(1.718, 0.399)$; Table S3, Supporting information). Due to the assumption that all adults were 1 year old in their first year of capture, this distribution most likely 467 468 possessed a larger variance than the true variance in house sparrows, which could explain the relatively high sensitivity of N_e/N to variation in T (Fig. 2). 469 In the second analysis of N_e/N , the demographic variances of the terminal age class were 470 found to be of higher importance for the ratio N_e/N than the demographic variances of 471 the first age classes and the stable sex-age distributions (Fig. 3). Particularly, N_e/N 472 was sensitive to the demographic variance of older males. The high estimated variance 473 of σ_{dgi}^2 in males (Table 2) was due to a few values, especially that of older males on 474Rånes (Table S4, Supporting information). Removal of this value reduced the estimate 475variance of older males $(m_2: 0.89)$, but it did not affect the outcome of the sensitivity 476 477 analysis. This shows that the analysis was not strictly dependent on the level of variance in the data, and that N_e/N was significantly more sensitive to older males than females, 478 even when their estimated variances were almost equal. Older males was found to 479 be the age class that affects N_e/N most, and this is consistent with other studies on 480

species with similar life history and mating system (e.g. Green 2001; Geslin et al. 2004). 481 Given that mating system (Nunney 1993; Nomura 2002; Lee et al. 2011) and life history 482 parameters (Orive 1993; Waples et al. 2011, 2013) are important determinants of effective 483 population size, the sex-age class specific demographic parameters with largest effect, 484 should probably vary accordingly. For instance, in many species senescence negatively 485affects reproductive output (e.g. the common lizard (Lacerta vivipara), Richard et al. 486 2005; and European badgers (Meles meles), Dugdale et al. 2011), suggesting that N_e/N 487 would be more sensitive to young or middle age-classes. 488 489 The age-class of highest importance for N_e/N will depend on the sampling scheme, that is, whether a population is sampled just before or after reproduction. For instance, using 490 post-reproduction census in populations with high fecundity and type III survival curves 491 492 (high mortality in early life stages), we could expect N_e/N to be most sensitive to the 493 survival to adulthood by younger age classes (Gaggiotti & Vetter 1999). However, using pre-reproduction census, such life histories often imply very large variance in reproductive 494 success among adults (in terms of production of recruits) which is one of the main causes of 495 low N_e/N ratios in such species (Hedgecock 1994; Hedgecock & Pudovkin 2011). Hence, if 496497 sampling is performed right before reproduction (as in this study), survival to recruitment 498 will be included in the variance in reproductive success by the adult age-class(es), in which case we could expect the result to be similar as in the current study. 499 500 In the sensitivity analyses of the individual reproductive values, all four sex-age classes were most sensitive to fecundity, implying that fecundity was important for N_e . Accordingly, 501 fecundity has also earlier been found to be one of the main factors that affect N_e (Caballero 502 1994; Frankham et al. 2010), and variation in fecundity should generally decrease the 503 504 effective size of a population (Nunney 1996). Effects of fecundity and variance in reproductive 505 success on effective population size and genetic drift have been found in a diverse set of species (e.g. many marine organisms, Hauser et al. 2002). Here, because we considered 506 507 production of recruits, the fecundity measures includes a survival component by offspring.

508 Although survival of adults were of less importance than adult fecundity, the survival rates of juveniles to recruitment may be essential for this species' genetic diversity. Populations 509 at different points in the fast-slow continuum may exhibit different trade-offs between 510 survival and reproduction, in which current reproduction may come at the cost of either 511 future reproduction (slow species) or survival (fast species) (Ricklefs 2000; Bleu et al. 512 2016). This may suggest that the ratio N_e/N could be more sensitive to adult survival 513 relative to fecundity or variance in reproductive success in populations with slower life-histories. 514 Moreover, a negative effect on N_e by increased variance in reproductive success may be 515 compensated for by delayed maturity (Broquet et al. 2009), but this may again depend 516 on the juvenile survival rates (Lee et al. 2011). 517

518 Demographic parameters affecting $N_{\rm e}/N$

519 Our dataset included both a recently founded population, a population that went extinct, and two artificially selected populations (Table 3), thus the range for each of the parameters 520 was expected to be large, but biologically reasonable. By including these four populations, 521 our results may reflect the relative sensitivity of N_e/N to different demographic histories 522 523 which may occur both in isolated populations, and in dynamic metapopulations. The habitat quality varied among the 13 populations, and they differed in quality of nesting 524 525 sites, food availability and shelter (Jensen et al. 2013). It is likely that this resulted in differences among the populations in demography and life-history, e.g. population growth 526 rates, mean population sizes, and age structure (Table 3). 527 Our result shows that the reproductive success of breeders increase with age for both sexes, 528 and this is consistent with several other studies of mammals (e.g. European badgers, 529530 Dugdale et al. 2011), several bird species (e.g. Rockwell et al. 1993; Forslund & Pärt 1995; McCleery et al. 2008), and fish species (e.g. Hixon et al. 2014; Waples 2016). Many 531 studies on life history parameters only consider female reproductive success, but there are 532 some exceptions. In birds, examples include the Australian brown thornbills (Acanthiza 533

534 pusilla), where only males improved their reproductive success with age (Green 2001), in French bluethroats (Luscinia svecica) both sexes improved, but males improved most 535 536 (Geslin et al. 2004), and in North-American house sparrows, males and females improved their reproductive success equally with age (Hatch & Westneat 2007). 537 538 Increased feeding of chicks by older males, especially in the first days after hatching, is one of the reasons males have the highest increase in reproductive value with age in brown 539 540 thornbills and bluethroats (Green 2001; Geslin et al. 2004). In house sparrows there is a positive relationship between feeding rates and visible badge size in males (Ringsby 541 542 et al. 2009), and both total and visible badge size has been shown to increase from age class 1 to age class 2 (Jensen et al. 2006). Furthermore, there is a positive relationship 543 between badge size, mating success and recruit production (Jensen et al. 2004, 2008). 544 The relationship between age and reproductive success in males, probably explain why 545the demographic variance in older males was the parameter N_e/N was most sensitive to 546 (Fig. 3, Table S6, Supporting information). 547 Survival was also identified as an important factor for W_{ij} , and thus N_e/N (Fig. 4). The 548 interaction effect of the two parameters was only significant ($\Delta S \geq 0.5$) in age class 1 549 550 (Table S7, Supporting information), and it might be related to the importance of surviving to age class 2, where on average, individuals have higher reproductive success (Table 2). 551 Accordingly, we found that age class 2 of both sexes had the highest mean age-specific 552reproductive values, v_i (Table 2), which reflect the expected future contribution from 553 an individual in a specific sex-age class to the long-term population growth (Keyfitz & 554 Caswell 2005). Three of the estimated parameters were from the population projection 555 556 matrix, 1 (eqn 2): the asymptotic population growth rate, λ , the reproductive values of the sex-age classes, v_i , and the stable sex-age distribution, u_i . The eigenvector v_i 557 was fixed at the average values for the dataset (Table 2) and thus not evaluated in the 558 sensitivity analysis. The other eigenvector u_i , is part of the definition of N_e/N (equations 559 1 and 6) and was included as a parameter in the sensitivity analysis, but was not found 560

561

to be of significant importance (Figs. 3 and Table S6, Supporting information). This

might be because it is truly non-important for variation in N_e/N , or simply because the 562 563 importance is not observable for a population at equilibrium. It would be possible to do a perturbation analysis of the parameter to see if an increase in variance would affect the 564 outcome of the sensitivity analysis, but as we used the global variance-based method, an 565 increase in conditional parameter variance would most likely not change the result (see 566 section Sensitivity analysis in Methods). 567 The model used to estimate N_e/N in the house sparrow populations rest on a few 568 569 simplifying assumptions, particularly the assumptions of density independent vital rates and isolated populations. We do not believe that the former should be of great concern 570 in the current study as there is little evidence for a relationship between ΔN and N in 571 any of the viable populations with more than 4 years of data (although at Gjerøy, there 572was a slight relationship, (linear regression: $\beta = -0.731 \pm 0.309, p = 0.050$)). Also, in 573 the study system the environment fluctuates considerably between years, which seems 574 to be an important factor affecting population size, particularly through its effects on 575 juvenile survival probabilities (Ringsby et al. 1999). This effect is accounted for in the 576 577 demographic variance. The second assumption (no gene flow) is violated for most of the populations and will affect the effective population size and rate of genetic drift if migrants 578 579 successfully produce recruits. Migration has previously been shown to affect genetic estimators in the same study system, causing them to give higher N_e estimates than 580 the demographic estimator (Baalsrud et al. 2014). This may be attributed to the effect 581 582successful migrants can have on the genetic stochasticity compared to the demographic stochasticity. The demographic effects of an immigrant will be captured by our method 583 through the fecundity and survival measures, probably causing a reduction in the estimate 584 585 of demographic N_e since immigrant males seems to have lower fitness in the study system (Pärn et al. 2009). However, it will not capture the direct (positive) effect a reproducing 586 immigrant may have in slowing the rate of genetic drift. On the other hand, the genetic 587 estimators will most probably be biased upwards by immigration in this study system 588

(see Baalsrud *et al.* 2014 for further discussion), which makes it likely that our estimates of N_e/N in the different populations are somewhat conservative, but not considerably biased due to this assumption.

592 Management

593 In population management, it is of great importance to determine factors that affect N_e . The demographic model we used to estimate N_e/N is based on annual demographic 594 data, which is mainly affected by demographic stochasticity, but also by selection, causing 595 variation in survival and reproductive success between individuals within a year. Nevertheless, 596 597 it is important to acknowledge that N_e/N will be affected by environmental stochasticity in the long-run (Engen et al. 2010). As a consequence, it is important to keep in mind that 598 N_e/N is affected not only by the species specific life history (Lee et al. 2011) and mating 599 600 system (Nunney 1991, 1993; Nomura 2002; Engen et al. 2003; Kaeuffer et al. 2004), but also external factors such as human activities (Therkildsen et al. 2010). 601 602 Depending on the management goals for the species or population in question, a manager might want to improve N_e or N_e/N . Given a specific population size, N_e should be 603 maximised to be as close to N as possible to minimise the genetic stochasticity in the 604 605 population. There are many ways to achieve this, e.g. manipulation of the sex ratio (Caballero 1994; Frankham et al. 2010), limit fluctuations in N (Kalinowski & Waples 606 2002; Engen et al. 2005b), and reduce variance in the mean family size (Manning et al. 607 608 2000; Gayet et al. 2016), as little to no variance will give a N_e that is larger than N (Wright 1984; Frankham et al. 2010). Under random mating, the number of inbreeding 609 610 events is independent of N, and the increase in N_e due to inbreeding avoidance diminishes 611 as N increases (Caballero & Hill 1992). In other words, for managers of small populations, inbreeding avoidance is important to increase N_e , but for larger populations, the efforts to maximise N_e should be focused elsewhere. However, in most management situations 613 it is N that is observed and manipulated. Due to problems related to small population

615sizes, such as high influence of demographic stochasticity and drift that increase the probability of extinction (Engen et al. 2005a; Legendre et al. 2008), it is therefore desirable 616 617to maximise N as well as N_e . The relative value of N_e/N can be used to assess population persistence based on influence of demographic, genetic and ecological factors (Kalinowski 618 619 & Waples 2002; Palstra & Fraser 2012). Of course, N_e/N can be numerically increased by reducing N (Kuparinen et al. 2016), but as the two are strongly correlated, a reduction in 620 621 N will also reduce N_e , leading an unchanged, or even reduced ratio. As small populations can have a high ratio, and enormous populations can have tiny ratios (see Hauser et al. 622 623 2002), its value is only informative in combination with information on N. However, 624 the ratio is expected to be more predictable in species with low variance in reproductive 625success (Frankham 1995; Frankham et al. 2014), but the relationship between N_e and N626 is still not well enough known to make inferences based on N_e or N alone (Luikart et al. 627 2010). For managers to extrapolate from an estimated N_e to N, they have to consider 628 the life history of the species in question. For instance, Waples et al. (2013) found that up to half the variation in N_e/N can be explained by age at maturity and adult lifespan. 629630 Combined with data on cross-generational fluctuations in N, it can provide more precise extrapolations and informative N_e/N estimates (Frankham et al. 2014). 631 632 Our sensitivity analysis suggest that for species with life histories resembling the house sparrow, management and conservation actions should focus on the demographic stochasticity 633 634 of older individuals, especially males. It is important to keep in mind that different 635 parameters can be important for a population depending on the time-scale (e.g. age structure is most important in the short-term, Waples 2010). To increase N_e , σ_{dgm2}^2 must 636 be reduced through manipulation of fecundity, but also survival (Fig. 4d, Table S7, 637 Supporting information). Specific management actions in this study system would be to 638 639 maintain suitable habitat, which often are dairy farms. It could also be possible to decrease variance in reproductive success by increasing chick survival to recruits, e.g. by increasing 640 the number of nest boxes and subject them to flea removal, have supplemental feeding 641 stations and remove predators such as feral cats. This study shows, as others before (e.g. 642

Taylor et al. 2012), that sensitivity analysis provides a useful, and perhaps vital tool to 643 assure the desired outcome in management and conservation. Implementing management 644 action without such information could result in ineffective (Carter et al. 2007; Cook et al. 645 2010; Walsh et al. 2012), or even devastating effects (e.g. supplementary feeding of female 646 kakapos (Strigops habroptilus), Robertson et al. 2006) of management actions. 647 We believe incorporation of empirical data that reflects the expected rate of loss of 648 genetic variation through genetic drift, such as N_e and N_e/N , together with traditional 649 knowledge-based management (Cook et al. 2010), will be important for conservation 650 651 and management in the future. This provides an area where sensitivity analyses can 652 be especially useful. For a manager, sensitivity analyses indicates the critical parameters 653 in a model (Tuljapurkar & Caswell 1997), and this insight can contribute to more effective and better decisions (Cook et al. 2010). In absence of data, or lack of resources to collect 654 655annual and individual data, it is possible to use samples to estimate the demographic parameters (Engen et al. 2010). The importance of number of generations and years of 656 657 data was not evaluated in this study, but to be able to maximize the output of management efforts, it should be further studied. 658

659 Conclusion

660 A population's resilience and vulnerability to extinction is determined by N and N_e , and our results provide knowledge on which demographic parameters that are important for 661 662 the rate of genetic drift in natural populations. Our study also shows the value of applying sensitivity analyses in population management, as they might identify where efforts 663 and resources should be focused. Due to the relatively extensive range of demographic 664 665 characteristics of the populations in this study, the results may be relevant to other isolated, fragmented, and perhaps threatened populations and species with similar life 666 histories and demography. 667

671

681

668 Acknowledgements

We would like to thank the inhabitants in our study area whose hospitality and friendliness 669 made this study possible. We are also grateful to our many field workers for assistance in 670 collecting the data, Margit Dagsdatter Haugsnes, Randi Røsbak and other technicians for help with the microsatellite genotyping, Steinar Engen and Jarle Tufto for guidance and 672 673 advice on the model and analysis, and Øystein Flagstad, editor Karen Chambers and three 674 anonymous referees for helpful comments on previous versions of the paper. This work was supported by grants from the European Research Council (ERC-2010-AdG 268562), 675 676 the Research Council of Norway (FRIBIO 204303 and 221956), and NTNU. This work was also partly supported by the Research Council of Norway through its Centres of Excellence 677 funding scheme (223257). The research was carried out in accordance with permits from 678 the Norwegian Environment Agency 2012/1201 ART-VI-ORD), the Norwegian Animal 679 Research Authority (FOTS ID 6094), and the Norwegian Bird Ringing Centre at Stavanger 680 Museum, Norway.

682 References

- 683 Allendorf F W, Luikart G H, Aitken S N (2012) Conservation and the genetics of
- 684 populations. John Wiley & Sons.
- 685 Altwegg R, Ringsby T H, Sæther B-E (2000) Phenotypic correlates and consequences of
- dispersal in a metapopulation of house sparrows Passer domesticus. Journal of Animal
- 687 Ecology, **69**, 762–770.
- 688 Anderson T R (2006) Biology of the ubiquitous house sparrow: from genes to populations.
- 689 Oxford University Press, New York.
- 690 Araki H, Waples R S, Blouin M S (2007) A potential bias in the temporal method for
- estimating N_e in admixed populations under natural selection. Molecular Ecology, 16,
- 692 2261–2271.
- 693 Ardren W R, Kapuscinski A R (2003) Demographic and genetic estimates of effective
- 694 population size (N_e) reveals genetic compensation in steelhead trout. Molecular Ecology,
- 695 **12**, 35–49.
- 696 Baalsrud HT, Sæther B-E, Hagen IJ et al. (2014) Effects of population characteristics and
- structure on estimates of effective population size in a house sparrow metapopulation.
- 698 Molecular Ecology, **23**, 2653–2668.
- 699 Billing A M, Lee A M, Skjelseth S et al. (2012) Evidence of inbreeding depression but not
- inbreeding avoidance in a natural house sparrow population. Molecular Ecology, 21,
- 701 1487–1499.
- 702 Bleu J, Gamelon M, Sæther B-E (2016) Reproductive costs in terrestrial male vertebrates:
- 703 insights from bird studies. Proceedings of the Royal Society Biological Sciences, 283,
- 704 20152600.
- 705 Broquet T, Jaquiéry J, Perrin N (2009) Opportunity for sexual selection and effective
- 706 population size in the lek-breeding European treefrog (Hyla arborea). Evolution, 63,
- 707 674–683.
- 708 Caballero A (1994) Developments in the prediction of effective population size. Heredity,
- 709 **73**, 657–679.
- 710 Caballero A, Hill W G (1992) Effective size of nonrandom mating populations. Genetics,
- 711 **130**, 909–916.
- 712 Carter S P, Delahay R J, Smith G C et al. (2007) Culling-induced social perturbation in
- 713 Eurasian badgers *Meles meles* and the management of TB in cattle: an analysis of a
- critical problem in applied ecology. Proceedings of the Royal Society Biological Sciences,
- 715 **274**, 2769–2777.
- 716 Caswell H (2001) Matrix population models: construction, analysis, and interpretation.
- 717 Sinauer Associates Incorporated.
- 718 Chan K, Saltelli A, Tarantola S (1997) Sensitivity analysis of model output:
- variance-based methods make the difference. In Proceedings of the 1997 Winter
- 720 Simulation Conference (edited by S Andradóttir, KJ Healy, DH Withers, BL Nelson),
- 721 261–268.

- 722 Chu-Agor M L, Muñoz-Carpena R, Kiker G, Emanuelsson A, Linkov I (2011) Exploring
- vulnerability of coastal habitats to sea level rise through global sensitivity and
- uncertainty analyses. Environmental Modelling & Software, 26, 593–604.
- 725 Cook C N, Hockings M, Carter R B (2010) Conservation in the dark? The information
- used to support management decisions. Frontiers in Ecology and the Environment, 8,
- 727 181–186.
- 728 Cutrera A P, Lacey E A, Busch C (2006) Intraspecific variation in effective population
- 729 size in talar tuco-tucos (Ctenomys talarum): the role of demography. Journal of
- 730 *Mammalogy*, **87**, 108–116.
- 731 Delignette-Muller M L, Dutang C (2015) fitdistrplus: an R package for fitting
- distributions. Journal of Statistical Software, **64**, 1–34.
- 733 Dugdale H L, Pope L C, Newman C, Macdonald D W, Burke T (2011) Age-specific
- breeding success in a wild mammalian population: selection, constraint, restraint and
- senescence. Molecular Ecology, 20, 3261–3274.
- 736 Engen S, Lande R, aether B-E, Weimerskirch H (2005a) Extinction in relation to
- 737 demographic and environmental stochasticity in age-structured models. *Mathematical*
- 738 biosciences, **195**, 210–227.
- 739 Engen S, Lande R, Sæther B-E (2003) Demographic stochasticity and allee effects in
- populations with two sexes. *Ecology*, **84**, 2378–2386.
- 741 Engen S, Lande R, Sæther B-E (2005b) Effective size of a fluctuating age-structured
- 742 population. *Genetics*, **170**, 941–954.
- 743 Engen S, Lande R, Sæther B-E, Dobson F S (2009) Reproductive value and the stochastic
- demography of age-structured populations. *The American Naturalist*, **174**, 795–804.
- 745 Engen S, Lande R, Sæther B-E, Festa-Bianchet M (2007a) Using reproductive value to
- estimate key parameters in density-independent age-structured populations. Journal of
- 747 theoretical biology, **244**, 308–317.
- 748 Engen S, Lande R, Sæther B-E, Gienapp P (2010) Estimating the ratio of effective to
- actual size of an age-structured population from individual demographic data. Journal
- 750 of Evolutionary Biology, **23**, 1148–1158.
- 751 Engen S, Ringsby T H, Sæther B-E et al. (2007b) Effective size of fluctuating populations
- with two sexes and overlapping generations. *Evolution*, **61**, 1873–1885.
- 753 Engen S, Sæther B-E, Kvalnes T, Jensen H (2012) Estimating fluctuating selection in
- age-structured populations. Journal of Evolutionary Biology, 25, 1487–1499.
- 755 Felsenstein J (1971) Inbreeding and variance effective numbers in populations with
- overlapping generations. Genetics, **68**, 581–597.
- 757 Forslund P, Pärt T (1995) Age and reproduction in birds hypotheses and tests. Trends
- 758 in Ecology & Evolution, **10**, 374–378.
- 759 Frankham R (1995) Effective population size/adult population size ratios in wildlife: a
- review. Genetical Research, **66**, 95–107.
- 761 Frankham R, Ballou J D, Briscoe D A (2010) Introduction to conservation genetics.

- 762 Cambridge University Press, Cambridge, United Kingdom.
- 763 Frankham R, Bradshaw C J A, Brook B W (2014) Genetics in conservation management:
- Revised recommendations for the 50/500 rules, Red List criteria and population
- viability analyses. Biological Conservation, 170, 56–63.
- 766 Freeman S, Herron J C (2007) Evolutionary analysis, 4th edn. Pearson Prentice Hall.
- 767 Gaggiotti O E, Vetter R D (1999) Effect of life history strategy, environmental variability,
- and overexploitation on the genetic diversity of pelagic fish populations. Canadian
- Journal of Fisheries and Aquatic Sciences, **56**, 1376–1388.
- 770 Gayet T, Devillard S, Gamelon M, Brandt S, Say L, Baubet E (2016) On the evolutionary
- consequences of increasing litter size with multiple paternity in wild boar (Sus scrofa
- 772 scrofa). Evolution, **70**, 1386–1397.
- 773 Geslin T, Questiau S, Eybert M-C (2004) Age-related improvement of reproductive success
- in Bluethroats Luscinia svecica. Bird Study, **51**, 178–184.
- 775 Gilbert K J, Whitlock M C (2015) Evaluating methods for estimating local effective
- population size with and without migration. Evolution, **69**, 2154–2166.
- 777 Green D J (2001) The influence of age on reproductive performance in the Brown
- 778 Thornbill. Journal of Avian Biology, **32**, 6–14.
- 779 Halliburton R (2003) Introduction to population genetics, 1st edn. Pearson Prentice Hall,
- 780 Upper Saddle River, NJ.
- 781 Harris R B, Allendorf F W (1989) Genetically effective population size of large mammals:
- an assessment of estimators. Conservation Biology, 3, 181–191.
- 783 Hatch M I, Westneat D F (2007) Age-related patterns of reproductive success in house
- sparrows Passer domesticus. Journal of Avian Biology, 38, 603–611.
- 785 Hauser L, Adcock G J, Smith P J, Ramírez J H B, Carvalho G R (2002) Loss of
- microsatellite diversity and low effective population size in an overexploited population
- of New Zealand snapper (*Pagrus auratus*). PNAS, **99**, 11742–11747.
- 788 Hedgecock D (1994) Does variance in reproductive success limit effective population sizes
- of marine organisms? In Genetics and evolution of aquatic organisms (edited by
- 790 A Beaumont), 122–134. Chapman & Hall, London.
- 791 Hedgecock D, Pudovkin A I (2011) Sweepstakes reproductive success in highly fecund
- marine fish and shellfish: a review and commentary. Bulletin of Marine Science, 87,
- 793 971–1002.
- 794 Hedrick P (2005) Large variance in reproductive success and the N_e/N ratio. Evolution,
- 795 **59**, 1596–1599.
- 796 Hill W G (1972) Effective size of populations with overlapping generations. Theoretical
- 797 population biology, **3**, 278–289.
- 798 Hill W G (1979) A note on effective population size with overlapping generations.
- 799 Genetics, **92**, 317–322.
- 800 Hixon M A, Johnson D W, Sogard S M (2014) BOFFFFs: on the importance of conserving
- old-growth age structure in fishery populations. ICES Journal of Marine Science, 71,

- 802 2171-2185.
- 803 Holand H, Jensen H, Tufto J, Pärn H, Sæther B-E, Ringsby T H (2015) Endoparasite
- 804 infection has both short- and long-term negative effects on reproductive success of
- female house sparrows, as revealed by faecal parasitic egg counts. *PloS ONE*, 10,
- 806 e0125773.
- 807 Homma T, Saltelli A (1996) Importance measures in global sensitivity analysis of nonlinear
- models. Reliability Engineering & System Safety, **52**, 1–17.
- 809 Husby A, Sæther B-E, Jensen H, Ringsby T H (2006) Causes and consequences of
- 810 adaptive seasonal sex ratio variation in house sparrows. Journal of Animal Ecology,
- 811 **75**, 1128–1139.
- 812 Jensen H, Moe R, Hagen I J et al. (2013) Genetic variation and structure of house sparrow
- populations: is there an island effect? Molecular Ecology, 22, 1792–1805.
- 814 Jensen H, Sæther B-E, Ringsby T H, Tufto J, Griffith S C, Ellegren H (2003) Sexual
- variation in heritability and genetic correlations of morphological traits in house sparrow
- 816 (Passer domesticus). Journal of Evolutionary Biology, **16**, 1296–1307.
- 817 Jensen H, Sæther B-E, Ringsby T H, Tufto J, Griffith S C, Ellegren H (2004) Lifetime
- 818 reproductive success in relation to morphology in the house sparrow *Passer domesticus*.
- 819 Journal of Animal Ecology, **73**, 599–611.
- 820 Jensen H, Steinsland I, Ringsby T H, Sæther B-E (2008) Evolutionary dynamics of a
- sexual ornament in the house sparrow (*Passer domesticus*): the role of indirect selection
- within and between sexes. Evolution, 62, 1275–1293.
- 823 Jensen H, Svorkmo-Lundberg T, Ringsby T H, Sæther B-E (2006) Environmental
- influence and cohort effects in a sexual ornament in the house sparrow, Passer
- 825 domesticus. Oikos, **114**, 212–224.
- 826 Kaeuffer R, Pontier D, Devillard S, Perrin N (2004) Effective size of two feral domestic
- cat populations (Felis catus L.): effect of the mating system. Molecular Ecology, 13,
- 828 483–490.
- 829 Kalinowski S T, Taper M L, Marshall T C (2007) Revising how the computer program
- 830 CERVUS accommodates genotyping error increases success in paternity assignment.
- 831 *Molecular Ecology*, **16**, 1099–1106.
- 832 Kalinowski S T, Waples R S (2002) Relationship of effective to census size in fluctuating
- populations. Conservation Biology, 16, 129–136.
- 834 Keyfitz N, Caswell H (2005) Applied mathematical demography. Springer Science &
- 835 Business Media, New York.
- 836 Kuparinen A, Hutchings J A, Waples R S (2016) Harvest-induced evolution and effective
- population size. Evolutionary applications, 9, 658–672.
- 838 Kvalnes T, Ringsby T H, Jensen H et al. (in review) Artificial selection on body size in a
- 839 wild passerine bird: presence of an optimal phenotype.
- 840 Lande R, Engen S, Sæther B-E (2003) Stochastic population dynamics in ecology and
- 841 conservation. Oxford University Press, New York.

- 842 Lebreton J D, Burnham K P, Clobert J, Anderson D R (1992) Modeling survival
- and testing biological hypotheses using marked animals: a unified approach with
- case-studies. Ecological Monographs, **62**, 67–118.
- 845 Lee A M, Engen S, Sæther B-E (2011) The influence of persistent individual differences
- and age at maturity on effective population size. Proceedings of the Royal Society
- 847 Biological Sciences, **278**, 3303–3312.
- 848 Legendre S, Schoener T W, Clobert J, Spiller D A (2008) How is extinction risk related
- to population-size variability over time? A family of models for species with repeated
- extinction and immigration. The American Naturalist, 172, 282–298.
- 851 Luikart G, Ryman N, Tallmon D A, Schwartz M K, Allendorf F W (2010) Estimation
- of census and effective population sizes: the increasing usefulness of DNA-based
- approaches. Conservation Genetics, 11, 355–373.
- 854 Manning H, Margan S H, Briscoe D A (2000) Does equalization of family sizes reduce
- genetic adaptation to captivity? Animal Conservation, 3, 357–363.
- 856 McCleery R H, Perrins C M, Sheldon B C, Charmantier A (2008) Age-specific
- reproduction in a long-lived species: the combined effects of senescence and individual
- quality. Proceedings of the Royal Society Biological Sciences, 275, 963–970.
- 859 Myhre A M, Engen S, Sæther B-E (2016) Effective size of density-dependent populations
- in fluctuating environments. *Evolution*, **70**, 2431–2446.
- 861 Nomura T (2002) Effective size of populations with unequal sex ratio and variation in
- mating success. Journal of Animal Breeding and Genetics, 118, 297–310.
- 863 Nossent J, Bauwens W (2012) Optimising the convergence of a Sobol' sensitivity analysis
- for an environmental model: application of an appropriate estimate for the square of the
- expectation value and the total variance. In *Proceedings of the sixth biannial meeting*
- of the International Environmental Modelling and Software Society, Leipzig, Germany,
- 867 July 1-5, 2012 (edited by R Seppelt, AA Voinov, S Lange, D Bankamp).
- 868 Nunney L (1991) The influence of age structure and fecundity on effective population size.
- Proceedings of the Royal Society Biological Sciences, 246, 71–76.
- 870 Nunney L (1993) The influence of mating system and overlapping generations on effective
- 871 population size. *Evolution*, **47**, 1342–1359.
- 872 Nunney L (1996) The influence of variation in female fecundity on effective population
- size. Biological Journal of the Linnean Society, **59**, 411–425.
- 874 Nunney L, Elam D R (1994) Estimating the effective population size of conserved
- populations. Conservation Biology, 8, 175–184.
- 876 O'Connor K D, Marr A B, Keller L F, Jeffery K J, Bruford M W (2006) Extra-pair
- 877 fertilization and effective population size in the song sparrow Melospiza melodia.
- 878 Journal of Avian Biology, **37**, 572–578.
- 879 Orive M E (1993) Effective population size in organisms with complex life-histories.
- 880 Theoretical population biology, 44, 316–340.
- 881 Palstra F P, Fraser D J (2012) Effective/census population size ratio estimation: a
- compendium and appraisal. Ecology and Evolution, 2, 2357–2365.

- 883 Pärn H, Jensen H, Ringsby T H, Sæther B-E (2009) Sex-specific fitness correlates
- 884 of dispersal in a house sparrow metapopulation. Journal of Animal Ecology, 78,
- 885 1216–1225.
- 886 Pärn H, Ringsby T H, Jensen H, Sæther B-E (2012) Spatial heterogeneity in the effects
- of climate and density-dependence on dispersal in a house sparrow metapopulation.
- Proceedings of the Royal Society Biological Sciences, 279, 144–152.
- 889 Pew J, Muir P H, Wang J, Frasier T R (2015) related: an R package for analysing
- pairwise relatedness from codominant molecular markers. Molecular Ecology Resources,
- 891 **15**, 557–561.
- 892 Pollak E (2000) The effective population size of some age-structured populations.
- 893 Mathematical biosciences, **168**, 39–56.
- 894 Prado-Martinez J, Sudmant P H, Kidd J M et al. (2013) Great ape genetic diversity and
- 895 population history. *Nature*, **499**, 471–475.
- 896 Primack R B (2010) Essentials of conservation biology, 5th edn. Sinauer Associates,
- 897 Sunderland, MA.
- 898 Pujol G, Janon A (2015) sensitivity: Functions for sensitivity analysis.
- 899 R Core Team (2016) R: A language and environment for statistical computing. R
- 900 Foundation for Statistical Computing, Vienna, Austria. https://www.R-project.org/.
- 901 Richard M, Lecomte J, de Fraipont M, Clobert J (2005) Age-specific mating strategies
- and reproductive senescence. *Molecular Ecology*, **14**, 3147–3155.
- 903 Ricklefs R E (2000) Density dependence, evolutionary optimization, and the diversification
- 904 of avian life histories. The Condor, 102, 9–22.
- 905 Ringsby T H, Berge T, Sæther B-E, Jensen H (2009) Reproductive success and individual
- 906 variation in feeding frequency of House Sparrows (Passer domesticus). Journal of
- 907 Ornithology, **150**, 469–481.
- 908 Ringsby T H, Sæther B-E, Altwegg R, Solberg E J (1999) Temporal and spatial variation
- 909 in survival rates of a house sparrow, Passer domesticus, metapopulation. Oikos, 85,
- 910 419–425.
- 911 Ringsby T H, Sæther B-E, Jensen H, Engen S (2006) Demographic characteristics of
- 912 extinction in a small, insular population of house sparrows in Northern Norway.
- 913 Conservation Biology, **20**, 1761–1767.
- 914 Ringsby T H, Sæther B-E, Tufto J, Jensen H, Solberg E J (2002) Asynchronous
- 915 spatiotemporal demography of a house sparrow metapopulation in a correlated
- 916 environment. *Ecology*, **83**, 561–569.
- 917 Robertson B C, Elliott G P, Eason D K, Clout M N, Gemmell N J (2006) Sex allocation
- 918 theory aids species conservation. Biology Letters, 2, 229–231.
- 919 Rockwell R F, Cooch E G, Thompson C B, Cooke F (1993) Age and reproductive success
- 920 in female lesser snow geese: experience, senescence and the cost of philopatry. The
- 921 Journal of Animal Ecology, **62**, 323.
- 922 Sæther B-E, Coulson T, Grøtan V et al. (2013) How life history influences population

- 923 dynamics in fluctuating environments. The American Naturalist, 182, 743–759.
- 924 Sæther B-E, Engen S, Møller A P et al. (2004) Life-history variation predicts the effects
- of demographic stochasticity on avian population dynamics. American Naturalist, 164,
- 926 793–802.
- 927 Saltelli A (2002) Making best use of model evaluations to compute sensitivity indices.
- 928 **145**, 280–297.
- 929 Saltelli A, Annoni P (2010) How to avoid a perfunctory sensitivity analysis. *Environmental*
- 930 *Modelling & Software*, **25**, 1508–1517.
- 931 Saltelli A, Annoni P, Azzini I, Campolongo F, Ratto M, Tarantola S (2010) Variance based
- 932 sensitivity analysis of model output. Design and estimator for the total sensitivity index.
- 933 Computer Physics Communications, 181, 259–270.
- 934 Saltelli A, Tarantola S, Campolongo F, Ratto M (2004) Sensitivity analysis in practice:
- 935 a guide to assessing scientific models. John Wiley & Sons, New York.
- 936 Schmeller D S, Merilä J (2007) Demographic and genetic estimates of effective population
- and breeding size in the amphibian Rana temporaria. Conservation Biology, 21,
- 938 142–151.
- 939 Shaffer M L (1981) Minimum population sizes for species conservation. Bioscience, 31,
- 940 131–134.
- 941 Shpak M (2007) Selection against demographic stochasticity in age-structured
- 942 populations. *Genetics*, **177**, 2181–2194.
- 943 Sobol' I M (1990) Sensitivity estimates for nonlinear mathematical models.
- 944 Matematicheskoe Modelirovanie, 2, 112–118.
- 945 Sobol' I M (1993) Sensitivity estimates for nonlinear mathematical models [English
- 946 Translation]. Mathematical Modeling and Computational Experiment, 1, 407–414.
- 947 Sobol' I M (2001) Global sensitivity indices for nonlinear mathematical models and their
- 948 Monte Carlo estimates. Mathematics and Computers in Simulation, 55, 271–280.
- 949 Speed D, Balding D J (2014) Relatedness in the post-genomic era: is it still useful? Nature
- 950 Reviews: Genetics, **16**, 33–44.
- 951 Tang Y, Reed P, Wagener T, Van Werkhoven K (2007) Comparing sensitivity analysis
- 952 methods to advance lumped watershed model identification and evaluation. Hydrology
- 953 and Earth System Sciences Discussions, 11, 793–817.
- 954 Taylor R L, Walker B L, Naugle D E, Mills L S (2012) Managing multiple vital rates to
- 955 maximize greater sage-grouse population growth. The Journal of Wildlife Management,
- 956 **76**, 336–347.
- 957 Therkildsen NO, Nielsen EE, Swain DP, Pedersen JS (2010) Large effective population
- 958 size and temporal genetic stability in Atlantic cod (Gadus morhua) in the southern Gulf
- 959 of St. Lawrence. Canadian Journal of Fisheries and Aquatic Sciences, 67, 1585–1595.
- 960 Tufto J, Ringsby T H, Dhondt A A, Adriaensen F, Matthysen E (2005) A parametric
- 961 model for estimation of dispersal patterns applied to five passerine spatially structured
- populations. The American Naturalist, 165, E13–E26.

- Tuljapurkar S, Caswell H (1997) Structured-population models in marine, terrestrial, and freshwater systems. Springer Science & Business Media, New York.
- 965 Waite T A, Parker P G (1996) Dimensionless life histories and effective population size.
 966 Conservation Biology, 10, 1456–1462.
- 967 Walsh J C, Wilson K A, Benshemesh J, Possingham H P (2012) Unexpected outcomes
- 968 of invasive predator control: the importance of evaluating conservation management
- actions. Animal Conservation, 15, 319–328.
- 970 Wang J (2007) Triadic IBD coefficients and applications to estimating pairwise 971 relatedness. *Genetical Research*, **89**, 135–153.
- 972 Wang J (2011) COANCESTRY: a program for simulating, estimating and analysing 973 relatedness and inbreeding coefficients. *Molecular Ecology Resources*, **11**, 141–145.
- Wang J, Caballero A (1999) Developments in predicting the effective size of subdivided populations. *Heredity*, **82**, 212–226.
- Waples R S (2005) Genetic estimates of contemporary effective population size: to what time periods do the estimates apply? *Molecular Ecology*, **14**, 3335–3352.
- 978 Waples R S (2010) Spatial-temporal stratifications in natural populations and how they
- 979 affect understanding and estimation of effective population size. *Molecular Ecology*
- 980 Resources, **10**, 785–796.
- 981 Waples R S (2016) Tiny estimates of the Ne /N ratio in marine fishes: Are they real? 982 Journal of fish biology, 89, 2479–2504.
- 983 Waples R S, Do C, Chopelet J (2011) Calculating Ne and Ne/N in age-structured populations: a hybrid Felsenstein-Hill approach. *Ecology*, **92**, 1513–1522.
- 985 Waples R S, Luikart G, Faulkner J R, Tallmon D A (2013) Simple life-history traits
- explain key effective population size ratios across diverse taxa. Proceedings of the Royal
- 987 Society Biological Sciences, 280, 20131339.
- 988 Wright S (1931) Evolution in mendelian populations. Genetics, 16, 97–159.
- 989 Wright S (1984) Evolution and the Genetics of Populations, Volume 2: Theory of Gene 990 Frequencies. University of Chicago Press, Chicago.
- 991 Yang J (2011) Convergence and uncertainty analyses in Monte-Carlo based sensitivity 992 analysis. *Environmental Modelling & Software*, **26**, 444–457.

993 Data accessibility

994 Data will be submitted to dryad when this manuscript is accepted for publication.



995 Author contributions

996 M.W.S. wrote analysis scripts, performed parentage and sensitivity analyses, and wrote 997 the paper with A.M.M. and H.J. Both H.H. and T.K. contributed with advice and inputs 998 to the statistical analyses and the parentage analysis. B.E.S., H.J. and T.H.R conceived 999 the research question and acquired funding. All authors contributed to the fieldwork and 1000 comments on the manuscript.



1001 Tables

Table 1: Population level parameters for each of the populations.

Population	N_e/N (fix. T) ^a	N_e/N (pop. T) ^b	T	σ_{dg}^2	λ
Aldra	0.36	0.30	2.35	1.42	1.05
Gjerøy	0.40	0.36	2.19	1.28	0.91
Handnesøy	0.56	0.74	1.49	0.90	0.79
Hestmannøy	0.42	0.42	1.95	1.22	0.79
Indre Kvarøy	0.47	0.48	1.94	1.09	0.79
Leka	0.20	0.22	1.75	2.55	1.07
Linesøya	0.82	1.35	1.20	0.62	0.74
Løkta	0.69	1.11	1.23	0.74	0.60
Nesøy	0.54	0.44	2.39	0.95	0.88
Rånes	0.19	0.30	1.25	2.64	1.03
Røvass	0.60	0.83	1.44	0.85	0.77
Vega	0.17	0.22	1.55	2.98	1.17
Ytre Kvarøy	0.31	0.38	1.63	1.62	0.72
Mean values	0.44	0.55	1.72	1.45	0.87

T and σ_{dg}^2 , population specific generation time and demographic variance, respectively λ , asymptotic population growth rate from the projection matrix l (eqn 2) a N_e/N calculated with T=1.97 (Jensen *et al.* 2008)

b N_e/N calculated with the population specific T

Table 2: Sex-age class specific parameters averaged across all populations.

		Values				
Parameter	i^{a}	Min	Mean	Max	Var	
	f_1	0.62	1.35	2.73	0.38	
- 2	f_2	0.70	1.67	3.19	0.87	
σ_{dgi}^2	m_1	0.52	1.27	3.00	0.42	
	m_2	0.45	1.54	3.75	1.26	
	f_1	0.05	0.20	0.31	4.62×10^{-3}	
24	f_2	0.05	0.26	0.46	9.25×10^{-3}	
u_i	m_1	0.05	0.20	0.31	4.62×10^{-3}	
	m_2	0.18	0.34	0.85	2.66×10^{-2}	
	f_1	0.59	0.90	1.13	2.46×10^{-2}	
21	f_2	0.58	1.06	1.44	3.97×10^{-2}	
v_i	m_1	0.60	0.90	1.13	2.44×10^{-2}	
	m_2	0.85	1.04	1.19	1.20×10^{-2}	
	f_1	0.00	0.47	1.00	0.25	
Ī	f_2	0.00	0.49	1.00	0.25	
J_{ij}	m_1	0.00	0.49	1.00	0.25	
	m_2	0.00	0.50	1.00	0.25	
	f_1	0.00	0.83	14.00	1.74	
$R_{\cdot \cdot \cdot}$	f_2	0.00	0.94	8.00	1.99	
B_{ij}	m_1	0.00	0.69	11.00	1.44	
	m_2	0.00	0.88	9.00	1.81	

Mean, variance, minimum and maximum values are given

 $[\]sigma_{dgi}^2$, sex-age specific demographic variance u_i and v_i , stable sex-age distribution and reproductive values, respectively J_{ij} and B_{ij} , individual survival and fecundity, respectively

^a sex-age classes

Table 3: Overview of study populations and their mean population sizes, N.

Number ^a	Population	Years of data	$n^{ m b}$	Mean N
1	Aldra ^c	1998-2007	10	29
2	Gjerøy	1993-2002	10	45
3	Handnesøy	2011-2013	3	66
4	Hestmannøy	1993-2012	20	128
5	Indre Kvarøy	1993-2002	10	38
6	$Leka^d$	2002-2009	8	127
7	Linesøya	2012-2013	2	73
8	Løkta	2011-2013	3	34
9	Nesøy	1993-2002	10	19
10	Rånes	2012-2013	2	46
11	Røvass	2011-2013	3	20
12	$Vega^d$	2002-2009	8	170
13	Ytre Kvarøy ^e	1993-1998	6	25

^a Numbers on map in Fig. 1

^b Total number of years of data for each population ^c Founded in 1998, by four individuals (Billing *et al.* 2012)

^d Was part of a selection experiment (2002-2005) (Kvalnes *et al.* in review)

^e Went extinct in 2000 (Ringsby *et al.* 2006)

1002 Figures

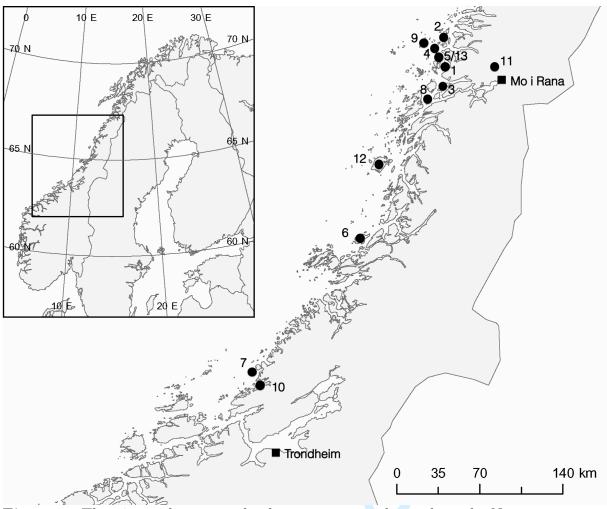


Figure 1: The 13 populations in the dataset are spread out along the Norwegian coast, from Trondheim (63°N, 10°E) in the south, to near Mo i Rana (66°N, 13°E) in the north (also see Table 3).

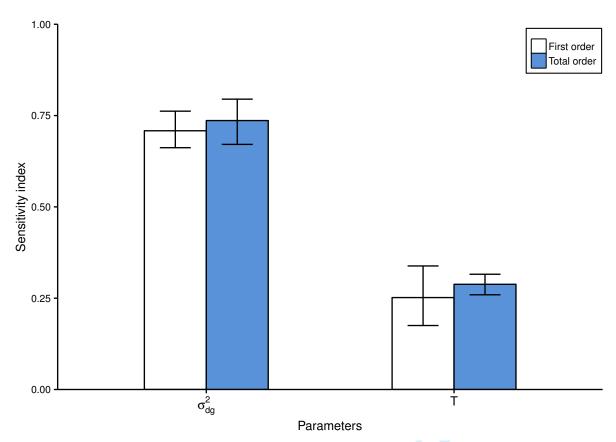


Figure 2: Results from the variance-based sensitivity analysis of N_e/N , with demographic variance, σ_{dg}^2 , and generation time, T, as input parameters, shown by first, S_p , and total order, S_{Tp} , sensitivity indices. The error bars show the 95% confidence intervals of the indicies.

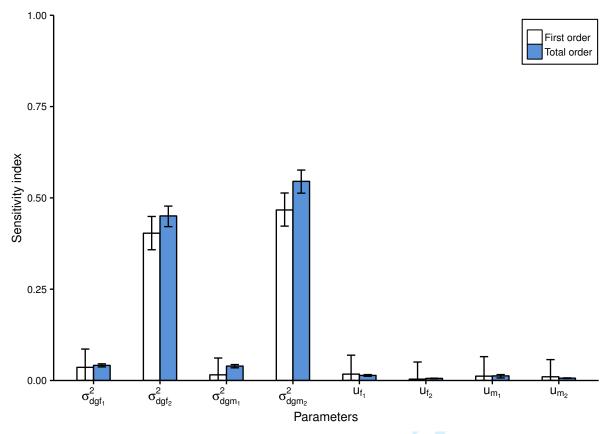


Figure 3: Results from the variance-based sensitivity analysis of N_e/N , where T=1.97 and with sex-age class specific demographic variance, σ_{dgi}^2 , and stable sex-age distribution, u_i , as input parameters, shown by first, S_p , and total order, S_{Tp} , sensitivity indices. The error bars show the 95% confidence intervals of the indicies

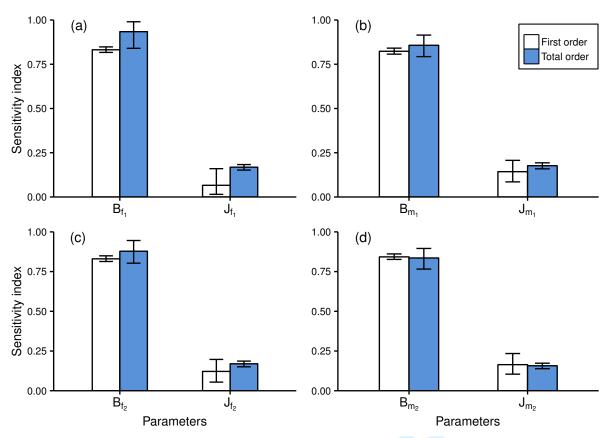


Figure 4: Results four variance-based sensitivity analyses of individual reproductive value, W_{ij} , with fecundity, B, and survival, J, as input parameters. The panels (a) and (b) show the results from females and males in age class 1, respectively, while panels (c) and (d) show the results from females and males in age class 2, respectively. Both the first, S_p , and total order, S_{Tp} , sensitivity indices are given. The error bars are the 95% confidence intervals of the indices.