

1 **Running headline:** Eco-evolutionary dynamics in ungulates

2 **Fluctuating effects of genetic and plastic changes in body mass on**
3 **population dynamics in a large herbivore**

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Abstract

Recent studies suggest that evolutionary changes can occur on a contemporary time scale. Hence, evolution can influence ecology and vice-versa. To understand the importance of eco-evolutionary dynamics in population dynamics, we must quantify the relative contribution of ecological and evolutionary changes to population growth and other ecological processes. To date, however, most eco-evolutionary dynamics studies have not partitioned the relative contribution of plastic and evolutionary changes in traits on population, community and ecosystem processes. Here, we quantify the effects of heritable and non-heritable changes in body mass distribution on survival, recruitment and population growth in wild bighorn sheep (*Ovis canadensis*) and compare their importance to the effects of changes in age structure, population density and weather. We applied a combination of a pedigree-based quantitative genetics model, statistical analyses on demography and a new statistical decomposition technique, the Geber method, to a long-term dataset of bighorn sheep on Ram Mountain (Canada), monitored individually from 1975 to 2012. We show three main results: (1) The relative importance of heritable change in mass, non-heritable change in mass, age structure, density and climate on population growth rate changed substantially over time. (2) An increase in body mass was accompanied by an increase in population growth through higher survival and recruitment rate. (3) Over the entire study period, changes in the body mass distribution of ewes, mostly through non-heritable changes, affected population growth to a similar extent as changes in age structure or in density. The importance of evolutionary changes was small compared to that of other drivers of changes in population growth but increased with time as evolutionary changes accumulated. Evolutionary changes became increasingly important for population growth as the length of the study period considered increased.

43 Our results highlight the complex ways in which ecological and evolutionary changes can affect
44 population dynamics and illustrate the large potential effect of trait changes on population processes.

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51 **Key words:** Eco-evolutionary, Population Dynamics, Ungulates, Biological Evolution, Breeding
52 values, Animal model
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Introduction

It has recently become apparent that evolutionary changes can occur on an ecological timescale (Thompson 1998). Rates of phenotypic changes are on average 1/4 (and up to 2/3) the rates of change in population size, suggesting that both processes occur on similar time scales (DeLong et al. 2016). Evolution on contemporary timescales has been documented in a wide range of organisms including plants (Maron et al. 2004), fish (Hendry et al. 2000), birds (Grant and Grant 2006) and humans (Milot et al. 2011). Contemporary evolutionary changes may affect ecological processes including population, community and ecosystem dynamics (Post and Palkovacs 2009) which could, in turn, produce a new selective landscape. Those reciprocal interactions between evolution and ecology are termed eco-evolutionary dynamics (Pelletier et al. 2009), and their quantification is important for a holistic understanding of factors driving population dynamics. For example, in predator-prey systems of rotifers (*Brachionus calyciflorus*) and green algae (*Chlorella vulgaris*), populations cycles varied according to whether or not prey populations were allowed to evolve (Yoshida et al. 2003). In another rotifer system, phosphorus limitation led to evolution of reduced investment in sex, which impacted population dynamics (Declerck et al. 2015). Most studies of eco-evolutionary dynamics to date, however, have been on short-lived species or in experimental systems (Hendry 2016a). Given the increasing reports of human-driven trait changes in nature (Darimont et al. 2009, Alberti et al. 2017), it is critical to assess the potential consequences of those changes in traits on population processes in wild and exploited species. For example, recent data-based models of the northeast Arctic cod stock suggest that both evolutionary and plastic changes in traits must be considered to accurately explain the observed trends in life-history traits and population dynamics (Eikeset et al. 2016).

77 Phenotypes can shape an individual's ability to survive, grow and reproduce. Consequently, the
78 distribution of phenotypes in a population can have a strong impact on its dynamics (Pelletier et al.
79 2007a). In five ungulate populations under long-term monitoring, the effect of a change in mean birth
80 weight on population growth was of similar magnitude to that of climatic drivers (Ezard et al. 2009).
81 This suggests that evolutionary changes of traits with a genetic basis can modify population dynamics.
82 Although a change in mean trait value in a population can be due to evolution, it can also be due to
83 other ecological factors including changes in age structure or plastic changes (Coulson and Tuljapurkar
84 2008). Most eco-evolutionary analyses to date, however, have not evaluated whether traits changes are
85 due to heritable or plastic changes (Ezard et al. 2009, Hendry 2016a).

86
87 Phenotypic plasticity, the ability of a single genotype to produce a range of phenotypes in
88 response to environmental variation (Forsman 2015), can have a major effect on population growth and
89 persistence (Reed et al. 2010) because it allows individuals to adjust their traits to variable
90 environments to maximize their survival and reproduction. For example, defense structures, such as the
91 spine in *Daphnia pulex*, can be induced by high predation risk (Spitze 1992) and the breeding
92 phenology of red deer (*Cervus elaphus*) can change to match ecological conditions (Moyes et al. 2011).
93 In the context of eco-evolutionary dynamics, it is important to know whether an effect of trait changes
94 on population processes is due to plasticity, genetic evolution, or both to identify whether the feedbacks
95 between traits and population processes are due to an eco-to-eco interaction, caused by a plastic change
96 in trait, or to an evo-to-eco interaction catalysed by genetic change. Eco-to-eco interactions are likely to
97 occur more frequently due to the rapidity of plastic trait change and may be of greater magnitude
98 (Hendry 2016b), but evo-to-eco effects may be more critical to the persistence of populations facing a
99 changing environment such as in the case of evolutionary rescue (Carlson et al. 2014). Partitioning the
100 relative contribution of plastic and evolutionary changes is also important because phenotypic plasticity

101 may compensate for evolutionary change, leading to cryptic eco-evolutionary dynamics (Kinnison et al.
102 2015). From an applied perspective, evolutionary changes in traits due to selective harvest are expected
103 to revert to their pre-selection state more slowly than plastic changes when harvesting stops (Swain et
104 al. 2007, Allendorf and Hard 2009, Pigeon et al. 2016). Thus, evo-to-eco interactions may have long-
105 term effects on population growth and productivity (Dunlop et al. 2015) and might be critical for
106 conservation and management. Thus, it is important to partition the effects of traits changes into those
107 due to heritable and non-heritable change to obtain realistic effect sizes of the relative importance of
108 evolutionary and ecological processes. Empirical examples of such distinctions remain scarce (Hendry
109 2016a).

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111 Several methods have been developed to decompose trait changes into their ecological and
112 evolutionary components (overview in van Benthem et al. 2017). When sufficient data are available,
113 the best approach is to use a pedigree-based quantitative genetics model (e.g., the animal model; Kruuk
114 2004) to evaluate the genetic basis of a trait, which can then be combined with the Geber approach
115 developed by Ellner et al. (2011) to partition the effects of plastic and evolutionary trait change on
116 population growth. This is an extension of the approach proposed by Hairston et al (2005), combining
117 it with the Price equation (Price 1970, 1972), to partition trait changes into the effects of heritable
118 change, non-heritable phenotypic change and environment. The Geber approach provides a general
119 framework to partition the contribution of change in breeding values, in trait and in environmental
120 factors on population-level parameter such as population growth rate. It does this by: 1) fitting models
121 to quantify the effects of the trait and ecological factors on population growth rate, 2) distinguishing
122 heritable trait change from non-heritable trait change (i.e. plastic change, also referred to as
123 environmental deviation) by comparing temporal phenotypic changes to changes in breeding values,
124 and 3) partitioning the variance in population growth rate by combining the observed changes in

125 average population breeding values, environmental deviation and ecological factors with their
126 estimated effects obtained from statistical modeling (Ellner et al. 2011).

127

128 Here we compare the relative importance of changes in phenotypic traits, climate, density and
129 age structure on population dynamics. We use the Geber approach to decompose the effects of heritable
130 and non-heritable change in mass, age structure, density and a large-scale climate index on survival and
131 reproduction of bighorn sheep (*Ovis canadensis*). Finally, we combine age-specific changes in survival
132 and recruitment with age structure to quantify the effects of a change in environment, focal trait or age
133 structure on population growth (Coulson et al. 2008). We do this for different time periods, when the
134 population growth rates increased and decreased. We expect that non-heritable changes in traits should
135 be more important than heritable changes in explaining the association between traits and population
136 growth on a short time scale. Further, we explore the impact of period length on our quantification of
137 the magnitude of ecological and evolutionary effects. We expected that the importance of heritable trait
138 changes on population processes would increase over longer time periods. We used detailed long-term
139 individual monitoring of female bighorn sheep from a wild population in Canada with a pedigree 8
140 generations deep (Coltman et al. 2005). By combining pedigree-based quantitative genetics models,
141 demographic statistical analyses and the extended Geber approach (Ellner et al. 2011), we bring novel
142 insights into the relative importance of density, age structure, heritable and non-heritable changes in
143 mass on changes in population size. Our results suggest that while the distribution of body mass can
144 have as much impact on population dynamics as density or age structure, this effect is mostly due to
145 plastic changes. Heritable changes in morphological traits of long-lived species tend to be slow and
146 may become larger than the non-heritable effect of traits changes only over a long temporal scale.

147

148 **Methods**

149 **Population and study area**

150 We studied bighorn sheep on Ram Mountain, Alberta, Canada (52°N, 115°W, elevation 1080 to
151 2170 m). The study area covers about 38 km² of alpine and subalpine habitat approximately 30 km east
152 of the Rocky Mountains. The population has been closely monitored each summer since 1975
153 (Jorgenson et al. 1993). Individuals are marked using ear tags or visual collars. Annual resighting
154 probability for ewes is over 99% (Jorgenson et al. 1997), so they can be considered dead when not seen
155 for a year. Since all females in the population are marked and an exact census is made yearly
156 (Jorgenson et al. 1997), we can precisely determine their annual survival rate. Sheep were captured
157 each year between late May and late September in a corral trap baited with salt (Jorgenson et al. 1997).
158 Ewes were typically recaptured every 4 to 5 weeks (mean= 3.10 captures/year, SD= 1.41).

160 **Phenotypic and environmental measurements**

161 Individuals were weighed at each capture. We adjusted mass to September 15 using linear mixed
162 models with restricted maximum likelihood where both the intercept and the slope were allowed to
163 vary for each individual (Martin and Pelletier 2011). Density was the number of adult females alive in
164 June each year, because females are the recruitment-limiting sex. Climate was represented by the
165 Pacific Decadal Oscillation (PDO) index, obtained from the Joint Institute for the Study of the
166 Atmosphere and Ocean website (<http://jisao.washington.edu>). We used winter PDO, the average of
167 monthly values from December to April, to evaluate the effect of winter harshness. Winter PDO affects
168 population growth (Ezard et al. 2009) and horn length (Douhard et al. 2016) in mountain sheep.

170 **Quantifying age structure**

171 Several age classes were needed to quantify age structure because survival and recruitment have
172 different age-specific patterns in large mammals. To choose age classes that best represent how survival
173 and recruitment differ through age, we compared the Akaike Information Criterion corrected for small
174 sample size (AICc) of all possible age class groupings and selected the classification that minimized
175 AICc for both survival and recruitment (Appendix: Fig S1). Animals were thus classified as lamb,
176 yearling, 2-3, 4-5, 6-7, 8-12 and older than 13 years. These classes broadly represent the ontogenetic
177 changes in mass, survival and recruitment identified previously using alternative techniques (Bérubé et
178 al. 1999, Loison et al. 1999).

179

180 **Model fitting**

181 The first step to partition variation of female population growth rate into the effects of five
182 factors of interest (non-heritable and heritable change in mass, density, PDO and age structure) was to
183 build models of individual annual survival and recruitment as a function of these factors. Population
184 growth rate, in a closed population like Ram Mountain, can be approximated by the combination of
185 survival and recruitment. Body mass was centered and scaled within each of the 7 previously
186 determined age class to facilitate model convergence and remove multicollinearity with age class. This
187 procedure also removed any effect of age from the change in mass, making it easier to partition the
188 effect of age structure from the effect of change in mass. Density was also centered and scaled to one
189 standard deviation to facilitate convergence (Bolker et al. 2013). We modeled survival from one year to
190 the next and recruitment (the probability of weaning a lamb the following year) using binomial
191 generalized mixed models, which controlled for non-independence due to repeated measurements by
192 including individual identity and year as random effects. We used model selection based on AICc to

193 identify parsimonious models with good predictive power. Candidate models included age class, mass,
194 density, winter PDO and years of high cougar (*Puma concolor*) predation (Festa-Bianchet et al. 2006)
195 as explanatory variables. Previous work detected positive density-dependence in a bighorn sheep
196 population at very low densities (Bourbeau-Lemieux et al. 2011). We therefore tested quadratic effects
197 of density to account for potential non-linear effects. We also tested the quadratic effects of mass and
198 PDO as well as the interactions between mass, density, PDO and age. Given that multiple candidate
199 models had similar support (difference in AICc < 4; Burnham and Anderson 2002), we used model
200 averaging (Mazerolle 2016) to obtain predicted responses and unconditional standard errors weighted
201 according to each model's AICc weight.

202

203 **Distinguishing heritable from non-heritable changes in mass**

204 To estimate breeding values, we constructed a pedigree of the population. The pedigree included
205 1066 marked sheep with 836 maternities and 508 paternities. Maternities are established from field
206 observations. From 1988 (except 1994 to 1996), tissue samples were collected for genetic analysis
207 (Poissant et al. 2012). Paternities were assigned using CERVUS (Marshall et al. 1998) at a confidence
208 level of >95% following Coltman et al. (2005).

209

210 Estimated breeding values (EBV) of mass, which represent an individuals' genetic value for this
211 trait, were obtained with a bivariate animal model with female and male mass as response variables
212 using a Bayesian framework with the "MCMCglmm" R library V2.21 (Hadfield 2010, Hadfield et al.
213 2010, Wilson et al. 2010). More details on this animal model are reported in Pigeon et al. (2016). In
214 dimorphic species with imperfect genetic correlation between male and female traits, a bivariate model

215 maximizes information while accounting for genetic correlations (Wolak et al. 2015). Estimation of
216 genetic parameters must account for the high genetic correlation between female and male mass
217 (posterior mean= 0.74, CI= 0.39 – 0.99) in order to obtain less biased estimates. We centered and
218 standardized mass in each age/sex (females aged 13 years and older were pooled due to low sample
219 size (N= 91) above that age) to have a mean of 0 and a variance 1 before analysis, because phenotypic
220 variances differed between sexes and increased with age. Centering and standardizing keeps EBV on
221 the same scale for further analysis. The initial model partitioned the phenotypic variance in male and
222 female body mass into its additive genetic (V_a), permanent environmental (V_{pe}), maternal (V_m),
223 cohort (V_{ce}) and yearly environmental (V_e) components and residual variance (V_r) (see Pigeon et al
224 (2016) for details on model parameterization). To obtain a comprehensive EBV accounting for the
225 above variance partitions, models with different random effects were compared using the deviance
226 information criterion (DIC), which balances model fit and complexity simultaneously (Wilson et al.
227 2010). The maternal variance component was dropped from the final model since it did not reduce
228 DIC. The model was first fitted with a multivariate inverse-Wishart prior (Wilson et al. 2010) and run
229 for 9000000 iterations with a burn-in period of 1500000 and a thinning of 7500. Sensitivity to the prior
230 was tested by re-running the model with more informative priors, leading to similar results (Pigeon et
231 al. 2016). We extracted the posterior distribution of breeding values from this final model, to reduce
232 bias in error estimation (Hadfield et al. 2010).

233

234 Finally, we distinguished the effect of heritable from non-heritable changes in population average
235 body mass. To do so, we fitted two linear regressions for each age-class, one of annual average
236 population mass as a function of time and a second of annual average EBV as a function of time. The
237 difference between the fitted values of these two regressions each year is the environmental deviation

for that year. The environmental deviation is, therefore, the change in trait mean expected if genotype frequencies were constant. Any deviation from this constant expectation is a non-heritable change in the trait and considered to be a plastic change in trait. This smoothing using regression has been suggested to reduce noise due to yearly environmental stochasticity (Ellner et al. 2011).

Variance partitioning

During our study, population size varied substantially through periods of increase and decline (Fig. 1b). To ensure parsimony while allowing time for heritable change in mass to occur, we separated the study into periods characterized by changes in population growth rate. To define these periods, we fitted a generalized additive model of mean annual growth rate as a function of year. We then compared this model to broken stick regressions using version 0.5-1.1 of the “segmented” R library (Muggeo 2003) with an increasing number of breakpoints. The most parsimonious model included three periods: period 1 (1975-1989), period 2 (1989-1997) and period 3 (1997-2012) (Appendix: Fig S2). We partitioned the variance in survival and recruitment independently for each period using an extension of the Geber method (Ellner et al. 2011). The method partitions the changes in each focal variable between the first and the last years of the period. To test the importance of period length, we also applied the variance partitioning approach (without error estimation, using posterior mode of EBV) on a range of periods from 37 one-year periods to a single 37-year period. We therefore produced an additional 115 periods lasting 1 to 37 years (mean = 6), which started in different years to cover systematically the entire study. We then tested for linear and quadratic effects of period length on the absolute effect size of heritable change in mass, non-heritable change in mass, change in density, age structure and climate on population growth rate using linear regressions.

260

261 Partitioning the variance in survival and recruitment was done in several steps. First, assuming a
 262 constant reaction norm, we added the mean annual EBV predicted by the linear regression with time to
 263 the environmental deviation to obtain a prediction of the expected trait for a given breeding value and
 264 environmental deviation. Second, we used the previously parameterized models (see model fitting
 265 section) to predict the expected value of survival or recruitment for each age class, given all
 266 combinations of annual mean EBV, annual environmental deviation, density and PDO. Cougar
 267 predation was fixed as absent, as it only occurred in 5 of 38 years. Age-specific predicted survival and
 268 recruitment were obtained using model averaging (Mazerolle 2016). The age-specific predicted
 269 survival and recruitment rates were then combined to produce a weighted population average according
 270 to their respective proportions in the population at each given time (Coulson et al. 2008) to account
 271 explicitly for changes in age structure (Appendix: Fig S3). Finally, we estimated the average effect of
 272 the observed change in a given variable on population mean survival and recruitment when all other
 273 factors were kept constant. For example, the following equation was used to estimate the effect of
 274 heritable change on survival and recruitment (Ellner et al. 2011).

$$EVO = \frac{\sum_{e=1}^2 \sum_{d=1}^2 \sum_{a=1}^2 \sum_{w=1}^2 (X_{2edaw} - X_{1edaw})}{16}$$

275

276 Where X_{2edaw} is the predicted survival or recruitment of a population with mean breeding value at the
 277 end of the period, the environmental deviation e , density d , age structure a and PDO w . X_{1edaw} is the
 278 same but with the mean breeding value at the beginning of the period. Values of 1 or 2 for e , d , a and w

279 refer to their given value at beginning and end of the period respectively. The resulting value of EVO is
280 the expected change in survival or recruitment due to the observed change in breeding value during the
281 period. Equivalent formulas were used for all other factors of interest. The effects of the 5 partitioned
282 factors (ECO, EVO, AGE.STR, DENS and PDO) sum to the change in survival and recruitment
283 predicted by the models for a given period. To better understand the impact of these effects on
284 population dynamics, we also repeated this last step with the predicted population growth rate, obtained
285 by dividing the predicted recruitment by 2 (to account for the production of male lambs assuming an
286 even birth sex-ratio) and adding the predicted survival.

287 **Measuring uncertainty**

288 We measured uncertainty at multiple levels. We used a Bayesian framework to estimate
289 breeding values and properly quantify EBV errors (Hadfield et al. 2010). For each sample of the
290 posterior distribution in breeding value, we re-calculated the temporal trend in breeding value and the
291 yearly environmental deviation. We then used model averaging as previously described to estimate
292 expected survival and recruitment for each age class and unconditional errors (eq. 6.12 of Burnham and
293 Anderson (2002)) around these values. To account for errors in these estimates, we randomly drew
294 values from a normal distribution centered on the expected value with standard deviation equal to the
295 unconditional errors of the expected value. Hence, we obtained distributions of predicted survival and
296 recruitment for each age-class that reflected uncertainty in both the estimation of environmental
297 deviance and in the estimation of the model parameters. We then applied the variance decomposition
298 approach described above on each realization of the Markov chain. The effects were considered
299 significant when the 95% highest posterior density interval did not overlap 0. All analyses were done in
300 R (Version 3.3.3; R Core Team 2016).

301

302 **Results**

303 **Model fitting**

304 Model selection for survival revealed that age class, density, their interaction and body mass
305 best explained female survival, as these variables were present in all models with difference in AICc <
306 4 (Appendix: Table S1 and Fig S4). Density had a strong negative effect on lamb survival but no effect
307 on survival for prime age classes (Fig S4). Body mass had a positive effect on survival of all age
308 classes. Predation decreased survival, while PDO had a very small quadratic effect. Predictions of
309 mean survival by year fitted observed values with a correlation of 0.36 ($P < 0.001$). Model selection for
310 recruitment revealed that age class, body mass, their interaction, density and predation best explained
311 female recruitment, as these variables were present in all models within <4 AICc units of the best
312 model (Appendix : Table S2 and Fig S5). Density had quadratic effects on recruitment, with maximum
313 recruitment at intermediate densities, suggesting an Allee effect. Mass increased the probability of
314 recruitment for females of all ages, although it was least important in prime-aged females, which had a
315 high probability of reproducing even when very light. PDO (associated with warm and dry winters) had
316 negligible effects on recruitment. Predation decreased recruitment. Predictions of average recruitment
317 by year fitted observed values with a correlation of 0.71 ($P < 0.001$).

318

319 **Variance partitioning**

320 Female mass varied substantially over time (Fig 1a) with periods of increase and decrease
321 (Appendix: Table S3). A significant portion of this variation in phenotype could be explained by
322 additive genetic variance; female mass was heritable ($h^2 = 0.22$, Appendix: Table S4). Temporal
323 variation in EBV of female mass, however, was much smaller than the observed variation in mass
324 (Appendix: Table S5 and Fig S6).

325

326 Partitioning the variance in survival into the effects of heritable change in mass, non-heritable
327 change in mass, change in density, age structure and climate revealed substantial variation in their
328 relative contribution over time (Fig 2a), except for PDO whose contribution was negligible over the
329 entire study. During the first period, a change in age structure associated with aging of the population
330 had the strongest effect (posterior mean=0.031, 95% CI = 0.022 – 0.040). Changes in heritable and
331 non-heritable mass made smaller and non-significant contributions (posterior means of 0.002; CI=-
332 0.013 – 0.015 and -0.008; CI=-0.008 – 0.019, respectively) resulting in an overall increase in survival
333 of 0.044. Changes in both density and PDO made negligible contributions to change in survival during
334 this period. During the second period, the observed non-heritable decline in mass and the observed
335 increase in density contributed most to the decline in survival (posterior means = -0.042; CI=-0.054 – -
336 0.031 and -0.020; CI=-0.030 – -0.011 respectively), resulting in an overall decrease in survival of
337 0.071. A non-significant negative effect (posterior mean of -0.002; CI=-0.013 – 0.010) of the heritable
338 change in mass was also detected. The third period was characterized by a strong positive effect of non-
339 heritable change in mass (posterior mean of 0.059; CI=0.039 – 0.077), which was opposed by the effect
340 of the change in density (posterior mean of -0.031; CI=-0.044 – -0.016) to yield a very small increase
341 in survival (0.011). Overall, non-heritable change in mass had the greatest effect on survival, followed
342 by density, age structure and heritable changes in mass (average absolute effect size of 0.036, 0.018,
343 0.013 and 0.006 respectively). Winter PDO had a negligible effect (average absolute effect size of
344 0.001) on changes in survival observed over 4 decades.

345

346 The effects of heritable change in mass, non-heritable change in mass, change in density, age
347 structure and climate on recruitment also varied considerably over the study periods (Fig. 2b). The first
348 period was dominated by a positive effect of the change in age structure associated with aging of the

349 population (posterior mean=0.174, 95% CI = 0.161 – 0.188). Ageing reduced the proportion of young
 350 females, especially lambs and yearlings, whose survival and recruitment are low (Appendix: Figure
 351 S7). Changes in density, heritable and non-heritable mass made smaller contributions (posterior means
 352 of 0.010 [CI=-0.004 – 0.022], -0.005 [CI=-0.030 – 0.017] and -0.002 [CI=0.027 – 0.024] respectively)
 353 resulting in an overall increase in recruitment of 0.177. During the second period, the non-heritable
 354 decline in mass and the strong change in density explained most of the 0.176 decline in probability to
 355 wean a lamb (posterior means of -0.059 [CI=-0.085 – -0.031] and -0.159 [CI=-0.180 – -0.140]
 356 respectively). Heritable change in mass had no effect (posterior mean of -0.002; CI=-0.0288 –0.024).
 357 These effects were counteracted in part by the positive effect on recruitment of the change in age
 358 structure, with more females in the more productive age classes (posterior mean of 0.043; CI=0.023 –
 359 0.061). The third period was characterized by a strong positive effect of non-heritable change in mass
 360 (posterior mean of 0.097; CI=0.063 – 0.134). This was opposed by the combined negative effect of the
 361 change in density and an increasingly younger age structure (posterior mean of -0.079 [CI=-0.104 – -
 362 0.053] and -0.078[CI=-0.102 – -0.051] respectively) resulting in a net decrease in recruitment of 0.024.
 363 Overall, change in age structure had the greatest effect on recruitment, followed by density, plastic
 364 change in mass and heritable changes in mass (mean absolute effect size of 0.098, 0.083, 0.052 and
 365 0.011 respectively). Winter PDO had a negligible effect (mean absolute effect size of 0.003) on changes
 366 in recruitment observed in the 4 decades monitored.

367

368 When survival and recruitment were combined into population growth rate, the relative
 369 importance of heritable change in mass, non-heritable change in mass, change in density, age structure
 370 and climate were similar to those presented above (Fig. 3). The first period was dominated by a
 371 positive effect (posterior mean=0.118, 95% CI = 0.108 – 0.131) of the change in age structure
 372 associated with aging of the population. Changes in density, heritable and non-heritable mass made

373 smaller contributions (posterior means of 0.007 [CI=-0.004 – 0.018], 0.004 [CI=-0.017 – 0.028] and -
 374 0.0003 [CI=0.023 – 0.020] respectively). During the second period, the non-heritable decline in mass
 375 and the strong change in density explained most of the decline in population growth (posterior means
 376 of -0.071 [CI=-0.091 – -0.050] and -0.099 [CI=-0.113 – -0.085] respectively). No effect (posterior
 377 mean of -0.004; CI=-0.023 – 0.017) of the heritable change in mass was detected. The change in age
 378 structure had a slight positive effect on population growth (posterior mean of 0.014; CI=0.0004 –
 379 0.026). The third period was characterized by a strong positive effect of non-heritable change in mass
 380 (posterior mean of 0.107; CI=0.075 – 0.136). This was opposed by the combined negative effects of the
 381 change in density and a younger age structure (posterior mean of -0.071 [CI=-0.089 – -0.050] and -
 382 0.042 [CI=-0.059 – -0.023] respectively). Heritable change in mass had a small non-significant effect
 383 (0.026 [CI=-0.005– 0.056]). In this population, non-heritable change in mass had the greatest effect on
 384 population growth, followed by density, age structure and heritable changes in mass (mean absolute
 385 effect size of 0.061, 0.059, 0.058 and 0.010 respectively). Winter PDO had a negligible effect (mean
 386 absolute effect size of 0.002) on changes in population growth observed in the 4 decades monitored.

387

388 By partitioned the relative importance of heritable and non-heritable changes in mass, change in
 389 density, age structure and climate on not only population growth rate, but also on survival and
 390 recruitment, we obtained a more mechanistic approach to population dynamics. To determine whether
 391 the effects on population growth manifested themselves mostly through effects on survival or on
 392 recruitment, we compared the absolute effect size of survival to the absolute effect size of half the
 393 recruitment, because population growth rate is equal to the mean survival added to half the recruitment
 394 (Coulson et al. 2008). Generally, effects through survival and recruitment did not differ significantly,
 395 but the relative importance of survival and recruitment upon how the variables we examined affected
 396 population growth varied among periods (Appendix: Table S6). Changes in age structure more strongly

397 affected population growth in all 3 periods via their effects on recruitment rather than on survival ($P =$
398 0, 0.004 and 0.001 respectively). Changes in density had greater effects on population growth rate by
399 affecting recruitment rather than survival, but this difference was only significant ($P < 0.001$) during
400 the second period. In all other period-variable combinations, recruitment and survival did not differ
401 significantly in their contribution to change in population growth rate.

402

403 The importance of heritable and non-heritable changes in mass, as well as changes in density,
404 age structure and climate on population growth rate varied according to the temporal scale at which
405 these processes were measured (Fig. 4). At smaller temporal scale, effects of change in age-structure
406 and non-heritable change in mass were most important. However, the direction and magnitude of those
407 effects were highly variable, ranging from -0.164 to 0.111 and -0.061 to 0.070 respectively. The
408 importance of heritable change was minimal over short periods but increased with period length,
409 producing a trend best fitted by a quadratic function ($B = 6.89\text{e-}4$, $P < 0.001$; $B.\text{quad} = -1.5\text{e-}5$, $P = 0.050$).
410 The absolute effects of both density ($B = 0.006$, $P < 0.001$; $B.\text{quad} = -0.0002$, $P < 0.001$) and non-heritable
411 change in mass ($B = 0.004$, $P < 0.001$; $B.\text{quad} = -0.0001$, $P < 0.001$) had strong quadratic trends, being
412 highest at intermediate lengths. We did not detect any significant trend between period length and the
413 importance of age structure. While a significant quadratic trend was found for PDO, its effect on
414 population growth rate remained very weak regardless of the length of the period considered.

415

416 Discussion

417

418 By combining a pedigree-based quantitative genetics model, demographic statistical analyses
419 and the recently developed Geber approach (Ellner et al. 2011) to exceptionally detailed data from a

420 wild ungulate population, we found that although ecological variables such as age structure and density
421 are major drivers of population dynamics, ecological and evolutionary changes in trait distribution also
422 have a significant effect that varies between periods of different population trends. When decomposed
423 by period, the most important cause of change in population dynamics was the non-heritable change in
424 mass, closely followed by density and age structure, although the magnitude of these effects varied
425 depending on the period considered. The effect of heritable changes in mass on survival, recruitment
426 and population growth on a yearly basis was not significant, but it increased with the length of the
427 period considered. We found only a weak signal of evolutionary change in bighorn ewe mass
428 (Appendix: Table S5). This result is not surprising given that breeding values for female mass did not
429 show major changes over the study period. Therefore, our retrospective analysis shows that female
430 plastic changes in mass play a dominant role over short time scales. Altogether, our study serves as an
431 example of how ecological and evolutionary variables can interact in rather complex ways within
432 populations, varying temporally and affecting different fitness components (survival or recruitment) to
433 ultimately drive population growth (Figs. 2, 3).

434

435 It has been suggested that intense selective hunting of males could have undesirable
436 consequences on population dynamics through indirect evolutionary impact on other segments of the
437 population (Conover and Munch 2002, Allendorf and Hard 2009). In bighorn sheep, intense selection
438 for smaller horns is exerted on males by trophy hunting. This anthropogenic selection led to a
439 significant decline in the EBV of male horn length (Pigeon et al. 2016), a trait genetically correlated to
440 female mass (genetic correlation = 0.43, (Poissant et al. 2012)). It was therefore suggested that
441 anthropogenic pressures, such as selective hunting, could lead indirectly to maladaptive changes in
442 female mass and potentially negatively affect population dynamics (Kuparinen and Festa-Bianchet

2017). Our study shows, however, that heritable change in mass played no significant role in the change in population growth rate of this population, likely because evolutionary changes in mass were minimal despite the evolutionary response of horn length to selective harvest (Pigeon et al. 2016). While female mass has a strong genetic correlation with male horn length, it is also under strong natural selection, which would have opposed any detrimental evolutionary change. In a stable environment, traits with a high impact on fitness are likely to be near evolutionary optimum and are therefore less likely to vary, reducing their influence on population growth. In a changing environment, however, species may find themselves with sub-optimal phenotypes, and evolutionary change could be of crucial importance.

The importance of phenotypic trait distribution for population growth has been recognized (Pelletier et al. 2007a), leading to increased interest in eco-evolutionary dynamics (Pelletier et al. 2009). In line with previous studies, we found important effects of changes in mass, a trait often used as an integrator of condition, for both survival (Festa-Bianchet et al. 1997, Côté and Festa-Bianchet 2001) and recruitment (Jorgenson et al. 1993, Martin and Festa-Bianchet 2011). We also show that this result translates into an important effect of mass on population growth, as previously suggested by a simpler analysis of five ungulate populations (Ezard et al. 2009). The impact of change in body mass on population dynamics, however, was mostly due to non-heritable changes, suggesting that our system is mostly driven by an eco-to-eco interaction. Mass is a highly plastic trait in bighorn sheep (Pelletier et al. 2007b). Temporal trends in the mean age-adjusted mass (Fig 1a) of the population are likely to reflect changes in body condition due to changes in resource availability or environmental conditions. High population density likely played an important role in reducing mass (Festa-Bianchet et al. 1998), through competition for resources. However, on a yearly basis, change in density explained less than

1% of the effects of plastic change in mass on population growth rate (Table S7). This may be an underestimate, however, since density may have delayed effect on mass through, for example, maternal effects (Monteith et al. 2009). While maternal effects are present in bighorn sheep, they decline with age and mostly disappear by age 2 (Wilson et al. 2005). Hence, on the temporal scale of the 3 observed periods, the effect of density on mass through maternal effect is unlikely to be the main driver of population growth. Further, the effects of density and environmental changes are not always positively correlated: plastic changes in mass and density had opposite effects during the third period of the study, suggesting that drivers of mass change other than density are also present.

Important effects of age structure on population dynamics have been reported for many ungulates including bighorn sheep (Festa-Bianchet et al. 2003), Soay sheep (Coulson et al. 2001) and red deer (Clutton-Brock et al. 1997, Clutton-Brock and Coulson 2002). Survival and recruitment vary substantially according to age (Gaillard et al. 2000). Hence, population age structure can drastically affect population growth. Lambs have much lower survival than adults and do not reproduce, so an increase in the proportion of lambs can reduce population growth rate. The impact of age structure may be reduced over longer time periods if it was to stabilize, which has yet to happen in this population. While density had a strong negative effect on population growth during the second period, it had no significant effect during the first period. The negative effect of change in density during the third period was unexpected. A decrease in density is usually associated with increased population growth. The opposite effect we detect is due to the quadratic relationship between density and both survival and recruitment, which appeared in all but one of the top recruitment models. Although a quadratic relationship may oversimplify how density affects this population (Sugeno and Munch 2013), the shape of this curve suggests positive density-dependence at low density (Courchamp et al. 1999). This might

489 reflect Allee effects which have also been observed in another population of bighorn sheep, where
490 probability to wean a lamb increased with density up to a threshold of around 90 sheep (Bourbeau-
491 Lemieux et al. 2011).

492

493 Ecological and evolutionary variables may affect population growth by acting through survival
494 and/or recruitment. Our analyses show that while both pathways were significant, the effects through
495 recruitment tended to be larger, especially for age structure and density. This result is novel but not
496 surprising given that adult female survival is high and varies little due to environmental canalization
497 (Gaillard and Yoccoz 2003). It must be noted, however, that the correlation between fitted and observed
498 values was only 0.36 for the survival model compared to 0.71 for the recruitment model, which may
499 account for the lower importance of survival compared to recruitment. The direction of the effects of
500 change in age structure, density, climate and mass on survival was similar to how these variables
501 affected mean population growth. In large herbivores, density dependence usually first affects juvenile
502 mortality, then age at first reproduction, followed by reproduction of adult females, and finally adult
503 mortality (Bonenfant et al. 2009). There are two likely reasons why density affected population growth
504 less through survival than recruitment. First, density usually affects survival of juveniles to a much
505 greater extent than survival of adults (Eberhardt 2002). Lambs represent a small proportion of the
506 population (average = 18%, SD= 0.07), therefore changes in juvenile survival cannot have strong
507 immediate impacts on population growth if survival of other age classes remains unchanged (Gaillard
508 et al. 2000). Second, we measured recruitment as weaning success. Hence, changes in pre-weaning
509 lamb survival will drive changes in recruitment while only survival from weaning to 1 year will be
510 included in survival. High variability in recruitment for this species (Gaillard et al. 2000) is consistent
511 with the large effect sizes of the observed variables on population growth rate through recruitment.

512

513 Changes in non-heritable and heritable components of mass, density, age structure and PDO
514 varied substantially over 40 years, making the selection of periods used for analysis challenging. With
515 long periods, the choice of start and end-points can influence the results. Choosing the period to
516 maximize change in one of the factors influencing mean population growth will maximize its relative
517 effects on population growth, biasing the analysis. By defining periods based on changes in population
518 growth (the response variable), we sought to minimize these biases while still maximising the change
519 in population growth rate available to partition. However, given the retrospective nature of this
520 analysis, the choice of study period will always have a consequence, as shown by the variance in effect
521 sizes (Figure 4). Analyses using an annual approach, however, over-emphasize factors with the
522 potential for quick change (Gingerich 1983, Gingerich 2001). While evolutionary changes can occur
523 over an ecological time scale (Thompson 1998, DeLong et al. 2016), significant genetic change
524 requires a turnover of individuals and cannot occur over a single year in a ed species like bighorn sheep
525 with a generation time of 6.6 years. At Ram Mountain, the yearly average turnover rate was only
526 20.1%. Thus, on a very short time-scale, at least for long-lived species, demography and other
527 ecological factors will always prevail, while in this system the relative importance of evolutionary
528 variables increased as longer time periods were considered. For species with higher turnover rate,
529 however, evolutionary variables might drive ecological processes on shorter time scale. This suggests
530 that there may be a partial mismatch between ecological time scales and evolutionary timescales as
531 suggested by DeLong et al. (2016). This mismatch may be larger in species with slow life-history such
532 as bighorn. To compare the relative importance of evolution and ecology, one must consider a time
533 span when both processes are possible but also the life history speed of the organism under study.
534 Interspecific comparisons of the relative importance of evolutionary changes should standardize the

535 time span of observations relative to the potential change in genotypes, which will vary with generation
536 time (Haldane 1949). For example, evolutionary changes over a year are likely to be much more
537 important for species with a short generation time such as *Daphnia* than for bighorn sheep, where
538 accurate knowledge of population age structure is much more important to predict near-term population
539 dynamics. Standardizing time span over generation time would allow for comparison of the relative
540 importance of evolution not only between systems of the same species, but also across taxa.

541

542 In conclusion, distinguishing plastic and genetic changes in traits is crucial to quantify the
543 importance of eco-evolutionary dynamics. Ignoring this distinction would have largely inflated the
544 importance of evolutionary changes in driving population dynamics in bighorn sheep. Instead, we
545 concluded that changes in female mass, although a very important driver of the population dynamics in
546 our study, were not driven by genetic changes. Therefore, at least for body mass, the effect of trait
547 changes on population dynamics are driven by eco-to-eco feedbacks and less by an evo-to-eco one. Yet,
548 it is important to note that plasticity has been shown to be heritable too (Pelletier et al. 2007b) and
549 therefore our classification of all plasticity as an ecological process is very conservative. We may
550 underestimate the contribution of evolution by using such a narrow definition. Indeed, evolution of
551 plasticity can play an important role in variation in population growth (Stoks et al. 2015). Only three
552 generations exposed to indirect selection through trophy hunting were included in this study, which
553 may be insufficient to detect a strong signal of eco-evolutionary dynamics. More empirical studies,
554 with different traits and species, are necessary to disentangle the effects of plastic and genetic trait
555 changes before broad conclusions can be made about the importance of evolutionary changes in the
556 wild and furthermore about the temporal scale at which these changes substantially affect population
557 dynamics. Our study highlights both the potential of evolutionary changes to modify population growth

rate and the dangers of considering all trait changes as signs of evolution. The distinction between plastic and genetic trait change must be an integral part of empirical studies investigating the importance of eco-evolutionary dynamics.

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766 **Figure Captions**

767 **Figure 1:** Temporal change in a) mean age-adjusted female mass in mid-September (error bars
768 represent SD); b) number of adult females and mean age of all females; c) PDO from 1975 to 2012 for
769 bighorn sheep at Ram Mountain, Alberta, Canada. Splines were fitted using locally weighted scatterplot
770 smoothing (loess).

771

772 **Figure 2:** Partitioning the change in a) probability to survive and b) to wean a lamb into five
773 components: non-heritable change in mass (dark blue; “eco”), heritable change in mass (light blue;
774 “evo”), change in age structure (green; “age.str”), change in population density (beige; “dens”) and
775 change in PDO (pink; “PDO”). Analyses of a long-term study of bighorn sheep in Canada were
776 conducted separately for 3 periods: 1975-1984, 1984-1997 and 1997-2012. Bars show the effect on
777 survival and recruitment of the observed change in a given variable. The values of the bar and the
778 associated errors represent the mean and 95% CI (see Measuring uncertainty section).

779

780 **Figure 3:** Partitioning the change in population growth into five components: non-heritable change in
781 mass (dark blue; “eco”), heritable change in mass (light blue; “evo”), change in age structure (green;
782 “age.str”), change in population density (beige; “dens”) and change in PDO (pink; “PDO”). Analyses
783 of a long-term study of bighorn sheep in Canada were conducted separately for 3 periods: 1975-1984,
784 1984-1997 and 1997-2012. Bars show the effect on population growth of the observed change in a
785 given variable. The values of the bar and the associated errors represent the mean and 95% CI (see
786 Measuring uncertainty section).

787

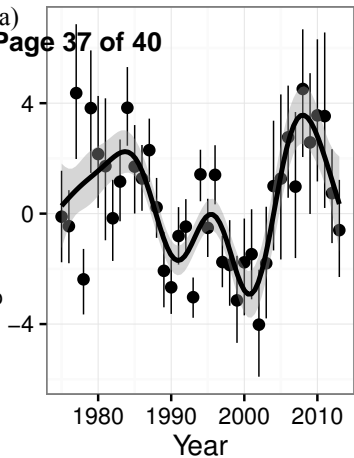
788 **Figure 4:** Absolute effect of non-heritable change in mass (dark blue, short dash), heritable change in
789 mass (light blue, dotted), change in age structure (green, long dash), change in population density
790 (beige, solid) and change in PDO (pink, dot-dash) on population growth rate according to the length of
791 the observation period (in years). Points represent the average absolute effect size (along with 95%
792 confidence limits) for each estimated period length. Lines represent linear or quadratic (depending on
793 which had lowest AICc) relationship between absolute effects size and the length of the period.

794

(a)

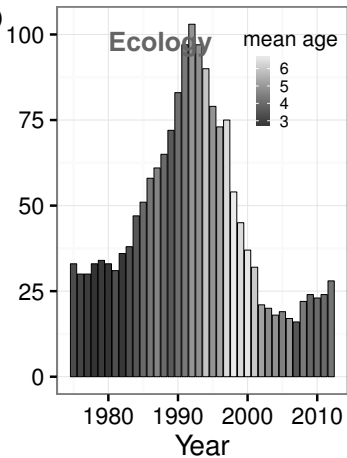
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Age corrected mass



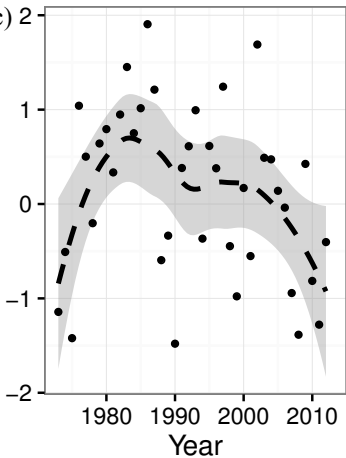
(b)

Number of adult females

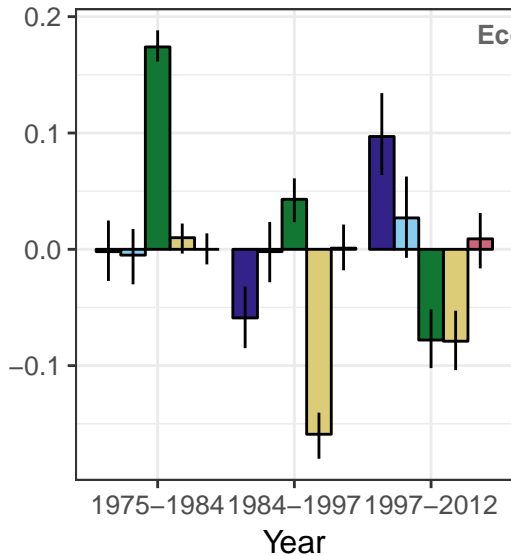


(c)

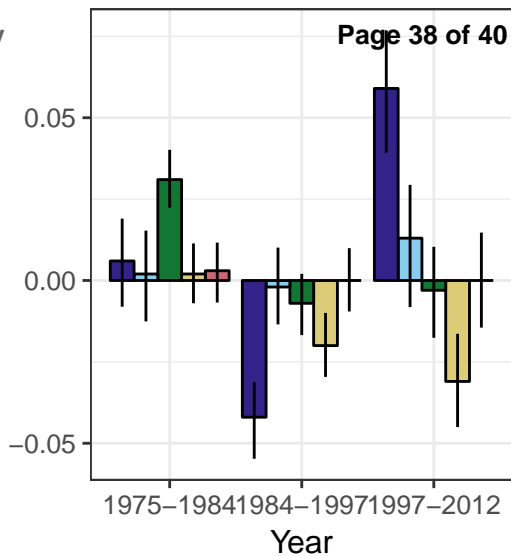
Winter PDO

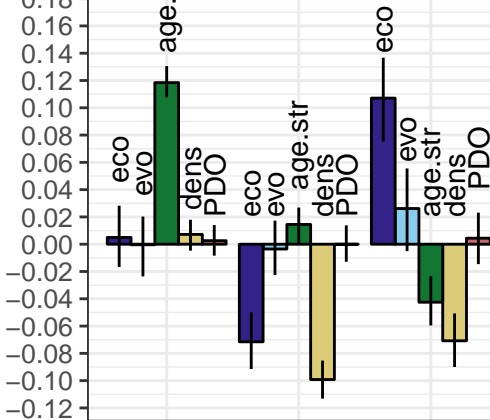


Effect on recruitment



Effect on survival



Effect on \bar{w} 

1975-1984 1984-1997 1997-2012

Year

Ecology

Effect on population growth rate

Period length

