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1 **Assessing seasonal demographic covariation to understand environmental-**
2 **change impacts on a hibernating mammal**

3 For consideration as *Letter in Ecology Letters*

4 Running title: Seasonal mechanisms of population dynamics

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30 conducted the analyses with substantial help from J.M. and D.C.; M.P. wrote the manuscript; all
31 authors discussed the results and commented on the manuscript.

32 Data accessibility statement: The datasets generated and analyzed during the current study as

33 well as the R scripts to run all analyses are available in the *Dryad* repository, [DOI].

34

35 **ABSTRACT**

36 Natural populations are exposed to seasonal variation in environmental factors that
37 simultaneously affect several demographic rates (survival, development, reproduction). The
38 resulting covariation in these rates determines population dynamics, but accounting for its
39 numerous biotic and abiotic drivers is a significant challenge. Here, we use a factor-analytic
40 approach to capture partially unobserved drivers of seasonal population dynamics. We use 40
41 years of individual-based demography from yellow-bellied marmots (*Marmota flaviventer*) to fit
42 and project population models that account for seasonal demographic covariation using a latent
43 variable. We show that this latent variable, by producing positive covariation among winter
44 demographic rates, depicts a measure of environmental quality. Simultaneous, negative
45 responses of winter survival and reproductive-status change to declining environmental quality
46 result in a higher risk of population quasi-extinction, regardless of summer demography where
47 recruitment takes place. We demonstrate how complex environmental processes can be
48 summarized to understand population persistence in seasonal environments.

49 INTRODUCTION

50 Effects of environmental change on survival, growth, and reproduction are typically investigated
51 based on annual transitions among life-history stages in structured population models (Salguero-
52 Gómez et al., 2016; Paniw et al., 2018). However, all natural ecosystems show some level of
53 seasonal fluctuations in environmental conditions, and numerous species have evolved life cycles
54 that are cued to such seasonality (Ruf et al., 2012; Varpe, 2017). For example, most temperate-
55 and many arid-environment species show strong differences in survival and growth among
56 seasons, with reproduction being confined mostly to one season (Childs et al., 2011; Rushing et
57 al., 2017; Woodroffe et al., 2017). Species with highly adapted, seasonal life cycles are likely to
58 be particularly vulnerable to environmental change, even if they are relatively long-lived
59 (Jenouvrier et al., 2012; Campos et al., 2017; Paniw et al., 2019). This is because adverse
60 environmental conditions in the non-reproductive season may carry-over and negate positive
61 environmental effects in the reproductive season in which key life-history events occur (Marra et
62 al., 2015). For instance, in species where individual traits such as body mass determine
63 demographic rates, environment-driven changes in the trait distribution in one season can affect
64 trait-dependent demographic rates in the next season (Bassar et al., 2016; Paniw et al., 2019).
65 Investigating annual dynamics, averaged over multiple seasons, may, therefore, obscure the
66 mechanisms that allow populations to persist under environmental change.

67 Despite the potential to gain a more mechanistic view of population dynamics, modeling
68 the effects of seasonal environmental change is an analytically complex and data-hungry
69 endeavor (Benton et al., 2006; Bassar et al., 2016). This is in part because multiple
70 environmental factors that change throughout the year can interact with each other and
71 individual-level (e.g., body mass) or population-level factors (e.g., density dependence) to

72 influence season-specific demographic rates (Benton et al., 2006; Lawson et al., 2015; Ozgul et
73 al., 2007; Paniw et al., 2019; Töpper et al., 2018). One major analytical challenge for ecologists
74 is that typically only a small subset of the numerous biotic and abiotic drivers of important life-
75 history processes are known and measured continuously (Teller et al., 2016); and this challenge
76 is amplified in seasonal models where more detail on such drivers may be required while
77 biological processes such as hibernation are cryptic to researchers (van de Pol et al., 2016).
78 Assessing whether the available information provides meaningful measures of biological
79 processes is another challenge. Nonlinear interactions among the myriad of biotic and abiotic
80 factors are common in nature, and teasing apart their effects on natural populations requires
81 detailed and long-term data (Benton et al., 2006; Paniw et al., 2019), which is not available for
82 most systems (Salguero-Gómez et al., 2015; 2016).

83 Overcoming the challenges in parameterizing seasonal population models is important
84 because a robust projections of such models require assessing the simultaneous effects of biotic
85 and abiotic factors on several demographic rates, causing the latter to covary within and among
86 seasons (Maldonado-Chaparro et al., 2018; Paniw et al., 2019). Positive environment-driven
87 covariation in demographic rates can amplify the population-level effects of environmental
88 change. For instance, Jongejans et al. (2010) demonstrated that positive covariation in survival
89 and reproduction in several plant populations magnified the effect of environmental variability
90 on population dynamics and increased extinction risk. On the other hand, antagonistic
91 demographic responses, either due to intrinsic tradeoffs or opposing effects of biotic/abiotic
92 factors, can buffer populations from environmental change (Knops et al., 2007; Van de Pol et al.,
93 2010); for instance, when population-level effects of decreased reproduction are offset by
94 increases in survival or growth (Connell & Ghedini, 2015; Reed et al., 2013; Vilellas et al.,

95 2015). Thus, explicit consideration of patterns in demographic covariation can allow for a fuller
96 picture of population persistence in a changing world. Such a consideration remains scarce
97 (Ehrlén & Morris, 2015; Ehrlén et al., 2016; but see Bassar et al., 2016; Compagnoni et al.,
98 2016).

99 Here, we investigated the population-level effects of seasonal covariation among trait-
100 mediated demographic rates (*i.e.*, collectively referred to as demographic processes), capitalizing
101 on 40 years (1976-2016) of individual-based data from a population of yellow-bellied marmots
102 (*Marmota flaviventris*). Our main aims were to (i) efficiently model demographic covariation in
103 the absence of knowledge on its underlying drivers and (ii) characterize the seasonal mechanisms
104 through which this covariation affects population viability. Yellow-bellied marmots have
105 adapted to a highly seasonal environment; individuals spend approximately eight months in
106 hibernation during the cold winter (September/October-April/May), and use the short summer
107 season (April/May-September/October) to reproduce and replenish fat reserves (Fig. 1). One
108 challenge that the marmot study shares with numerous other natural systems is the identification
109 of key proximal biotic and abiotic factors driving population dynamics. In marmots such factors
110 are numerous and affect population dynamics through complex, interactive pathways
111 (Maldonado-Chaparro et al., 2017; Oli & Armitage, 2004), which include interactions with
112 phenotypic-trait structure (Ozgul et al., 2010). As a result, measures of environmental covariates
113 (e.g., temperature or resource availability) have previously shown little effect on the covariation
114 of marmot demographic processes (Maldonado-Chaparro et al., 2018). To address this challenge,
115 we used a novel method, a hierarchical factor analysis (Hindle et al., 2018), to model the
116 covariation of demographic processes as a function of a shared latent variable, quantified in a
117 Bayesian modeling framework. We then built seasonal stage-, mass-, and environment-specific

118 integral projection models (IPMs; Ellner et al., 2016) for the marmot population, which allowed
119 us to simultaneously project trait distributions and population dynamics across seasons. We used
120 prospective stochastic perturbation analyses and population projections to assess how the
121 observed demographic covariation mediated population viability.

122 **METHODS**

123 *Study species*

124 Yellow-bellied marmots are an ideal study system to assess the effects of seasonal covariation in
125 demographic rates on population viability. These large, diurnal, burrow-dwelling rodents
126 experience strong seasonal fluctuations in environmental conditions, and their seasonal
127 demography has been studied for > 40 years (Armitage, 2014). Our study was conducted in the
128 Upper East River Valley near the Rocky Mountain Biological Laboratory, Gothic, Colorado (38°
129 57' N, 106° 59' W). Climatic conditions in both winter and summer have been shown to
130 influence reproduction and survival in the subsequent season (Lenihan & Van Vuren, 1996; Van
131 Vuren & Armitage, 1991). In addition, predation is major cause of death in the active summer
132 season (Van Vuren, 2001; Maldonado-Chaparro et al., 2017) and may be particularly severe
133 shortly before (Bryant & Page, 2005) or after hibernation (Armitage, 2014), especially in year
134 with heavy snow (Blumstein, pers. obs.). The effects of these factors on the demography of
135 yellow-bellied marmots are mediated through body mass, with heavier individuals more likely to
136 survive hibernation, reproduce in summer, and escape predation (Armitage et al., 1976; Ozgul et
137 al., 2010). Population dynamics of marmots are therefore likely to be susceptible to changes in
138 seasonal patterns of biotic and abiotic drivers. However, numerous interacting climatic factors,
139 such as temperature extremes and length of snow cover, determine both winter and summer
140 environmental conditions. The effects on marmot demography of these climatic factors, and of

141 interactions between climate and predation (the latter mostly a cryptic process) have been shown
142 to be difficult to disentangle (Schwartz & Armitage, 2002; Schwartz & Armitage, 2005).

143

144 *Seasonal demographic rates and trait transitions*

145 For this study, we focused on the population dynamics of eight major colonies continuously
146 monitored since 1976 (Armitage, 2014; Supporting Material S1). Each year, marmots were live-
147 trapped throughout the growing season in summer (and ear-tagged upon first capture), and their
148 sex, age, mass, and reproductive status were recorded (Armitage & Downhower, 1974; Schwartz
149 et al., 1998). All young males disperse from their natal colonies, and female immigration into
150 existing colonies is extremely rare; as such, local demography can be accurately represented by
151 the female segment of the population (Armitage, 2010). Thus, we focused on seasonal
152 demographic processes of females only. We classified female marmots by age and reproductive
153 status: juveniles (< 1 year old), yearlings (1 year old), and non-reproductive (≥ 2 years old; not
154 observed pregnant or with offspring) and reproductive adults (≥ 2 years old; observed pregnant
155 or with offspring) (Armitage & Downhower, 1974).

174 Female marmots give birth to one litter from mid-May to mid-June. In our population
175 model, females ≥ 2 year of age that survived the winter were considered reproductive adults at
176 the beginning of summer if they were observed to be pregnant or with pups, or non-reproductive
177 adults otherwise (Fig. 1). Upon successful reproduction, weaned offspring emerge from burrows
178 ca. 35 days after birth (Armitage et al., 1976); we therefore defined recruitment as the number of
179 female juveniles weaned by reproductive females that survive the summer (Fig. 1). The sex ratio
180 of female:male recruits was assumed to be 1:1 (Armitage & Downhower, 1974). Observations
181 and pedigree analyses allowed us to determine the mother of each new juvenile recruited into the
182 population (Ozgul et al., 2010).

183 To assess changes in body mass from one season to the next, we estimated body mass of
184 every female at the beginning of each season: June 1 (beginning of the summer season when
185 marmots begin foraging) and August 15 (beginning of the winter season in our models when
186 foraging activity decreases). Mid-August is the latest that body mass for the vast majority of
187 individuals can be measured and has been shown to be a good estimate of pre-hibernation mass
188 (Maldonado-Chaparro et al., 2017). Body-mass estimates on the two specific dates were
189 estimated using linear mixed effect models. These models were fitted for each age class and
190 included the fixed effect of day-of-year on body mass, and the random effects of year, site and
191 individual identity on the intercept and on the day-of-year slope (for details see Ozgul *et al.*,
192 2010; Maldonado-Chaparro et al., 2017). Body mass of juvenile females was estimated for
193 August 15.

194

195 *Modelling covariation in demographic processes – latent-variable approach*

196 We jointly modeled all seasonal demographic and mass change rates (*i.e.*, demographic
197 processes) as a function of stage and body mass - or mother's mass in the case of juvenile mass -
198 at the beginning of a season, using a Bayesian modeling framework (Table 1; Supporting
199 Material S1). All mass estimates were cube-root transformed to stabilize the variance and
200 improve the normality of the residuals in the Gaussian submodels (Maldonado-Chaparro et al.,
201 2017). We fitted all demographic-process submodels as generalized linear mixed effects models
202 (GLMMs). We assumed a binomial error distribution (logit link function) for the probability of
203 winter (θ_w) and summer (θ_s) survival and of probability of reproducing (*i.e.*, being in the
204 reproductive adult stage at the beginning of summer; ϕ_0); a Poisson error distribution (log link
205 function) for the number of recruits (ϕ_1); and a Gaussian error distribution (identity link) for the
206 masses (z^*) at the end of each season (Table 1). Mass-change (*i.e.*, mass gain or loss) rates (γ)
207 were then defined as functions of current (z) and next (z^*) mass using a normal probability
208 density function. For the juvenile mass distribution (ϕ_2), the density function depended on the
209 mother's mass (z_M) (see below; Supporting Material S2).

210 To model temporal covariation in seasonal demography in the absence of explicit
211 knowledge on key biotic or abiotic drivers of this covariation, we used a factor-analytic
212 approach. This approach has recently been proposed by Hindle and coauthors (2018) as a
213 structured alternative to fit and project unstructured covariances among demographic processes
214 when factors explaining these covariances are not modeled. We implemented this novel
215 approach parameterizing a model-wide latent variable (Q_y) which affected all demographic
216 processes in a given year (y) (for details see Supporting Material S1 and Hindle *et al.*, 2018). Q_y
217 was incorporated as a covariate in all seven demographic-process submodels (Table 1). Year-
218 specific values of Q_y were drawn from a normal distribution with mean = 0 and SD = 1. The

219 associated β_q slope parameters then determine the magnitude and sign of the effect of Q_y on a
220 given, season-specific demographic process (Table 1). To make the Bayesian model identifiable,
221 we constrained the standard deviation of Q_y to equal 1 and arbitrarily set the β_q for summer
222 survival (θ_s) to be positive. The β_q of the remaining submodels can, therefore, be interpreted as
223 correlations of demographic processes with θ_s .

224 Aside from the latent variable Q_y simultaneously affecting all demographic processes, we
225 included a random year effect ($\varepsilon_{Ysubmodel}$) as a covariate in each submodel. While Q_y captured
226 demographic covariation, the year effect accounted for additional temporal variation of each
227 demographic process not captured by Q_y . We also tested for the effect of population density
228 (measured as total abundance, abundance of adults, or abundance of yearling and adults) in all
229 submodels. However, like previous studies, we could not detect any clear density effects
230 (Armitage, 1984; Maldonado-Chaparro et al., 2018).

231 The prior distributions of the Bayesian model and posterior parameter samples obtained
232 are detailed in Supporting Material S1. For each demographic-process submodel, we chose the
233 most parsimonious model structure by fitting a full model that included all covariates (mass,
234 stage, and Q_y) and two-way interactions between mass and stage and stage and Q_y , and retaining
235 only those parameters for which the posterior distribution ($\pm 95\%$ C.I.) did not overlap 0 (Table
236 1; Table S1.1).

237 **Table 1:** Parameterization of the most parsimonious models describing winter (W) and summer (S)
238 demographic processes in marmots. The distributions B, N, and P correspond to the Bernoulli, normal,
239 and Poisson distributions, respectively. *Stage* – life cycle stage. *Q* – latent environmental variable. *z* –
240 season-specific mass. *z_M* – mass of the mother.

Demographic process	Function	Likelihood distribution
Winter (W):		
Survival (θ_w)	$\text{logit}(\theta_w) = \alpha_{0\theta w} + \alpha_{a\theta w}[\text{stage}] + \beta_{z\theta w} \times z + \beta_{q\theta w} \times Q_y[\text{year}] + \varepsilon_{y\theta w}[\text{year}]$	$\mathcal{B}(\theta_w)$
Mass next (z_w^*)	$z_w^* = \alpha_{0z^*w} + \alpha_{az^*w}[\text{stage}] + (\beta_{zz^*w} + \beta_{zaz^*w}[\text{stage}]) \times z + \beta_{qz^*w} \times Q_y[\text{year}] + \varepsilon_{yz^*w}[\text{year}]$	$\mathcal{N}(z_w^*, \tau_{z^*w})$
Reproduction (φ_0)	$\text{logit}(\varphi_0) = \alpha_{0\varphi_0} + \alpha_{a\varphi_0}[\text{stage}] + \beta_{z\varphi_0} \times z + \beta_{q\varphi_0} \times Q_y[\text{year}] + \varepsilon_{y\varphi_0}[\text{year}]$	$\mathcal{B}(\varphi_0)$
Summer (S):		
Survival (θ_s)	$\text{logit}(\theta_s) = \alpha_{0\theta s} + \alpha_{a\theta s}[\text{stage}] + \beta_{z\theta s} \times z + \beta_{q\theta s} \times Q_y[\text{year}] + \varepsilon_{y\theta s}[\text{year}]$	$\mathcal{B}(\theta_s)$
Mass next (z_s^*)	$z_s^* = \alpha_{0z^*s} + \alpha_{az^*s}[\text{stage}] + (\beta_{zz^*s} + \beta_{zaz^*s}[\text{stage}]) \times z + \beta_{qz^*s} \times Q_y[\text{year}] + \varepsilon_{yz^*s}[\text{year}]$	$\mathcal{N}(z_s^*, \tau_{z^*s})$
Number of recruits (φ_1)	$\log(\varphi_1) = \alpha_{0\varphi_1} + \beta_{z\varphi_1} \times z + \beta_{q\varphi_1} \times Q_y[\text{year}] + \varepsilon_{y\varphi_1}[\text{year}]$	$\mathcal{P}(\varphi_1)$
Juvenile mass (z_j^*)	$z_j^* = \alpha_{0z^*j} + \beta_{zz^*j} \times z_M + \beta_{qz^*j} \times Q_y[\text{year}] + \varepsilon_{yz^*j}[\text{year}]$	$\mathcal{N}(z_j^*, \tau_{z^*j})$

241

242

243 *Interpreting demographic covariation: latent variable as a measure of environmental quality*

244 The latent variable, Q_y , effectively captured the covariation among the demographic processes

245 (Supporting Material S1); therefore, using one latent variable across both seasons was sufficient.

246 Our GLMMs showed a strong effect of Q_y on winter but not summer demographic processes.

247 This effect was positive for all winter demographic processes, as evidenced by the positive β_q

248 (Table S1.1). The β_q for demographic processes in the summer, however, were comparatively

249 small and were not significantly different from 0 (95 % posterior C.I.s overlapped 0). The

250 positive β_q indicate that Q_y effectively estimates the overall annual environmental quality or

251 suitability, capturing both biotic and abiotic processes. A positive value of Q_y then depicts an
252 environmental condition at a given time point that increases winter survival and probability of
253 reproducing and decreases mass loss (Hindle et al., 2018). The variation in Q_y was in part
254 explained by environmental variables measured at the study site, but was unrelated to population
255 density (Supporting Material S1). Negative values of Q_y were associated with longer and more
256 severe winters and a higher snow cover, while positive Q_y indicated warmer winters and springs.
257 However, as the environmental variables explained $< 50\%$ of the variation in Q_y , the latent
258 variable captures multivariate, partly unobserved biotic and abiotic processes into a simple,
259 univariate measure of how bad ($Q_y < 0$) or good ($Q_y > 0$) environmental conditions are likely to
260 affect marmot demography.

261 Aside from the effects of environmental quality, our models are consistent with previous
262 findings on the importance of body mass and stage on yellow-bellied marmot demography
263 (Maldonado-Chaparro et al., 2017; Ozgul et al., 2010). The most parsimonious GLMMs (Table
264 S1.1) showed a positive effect of mass on all demographic processes, with the weakest effect of
265 mass on summer survival (θ_s) of reproductive adults. Survival, in particular θ_s , was highest for
266 reproductive adults; reproduction was also highest for adults that reproduced before (Fig. S1.5).

267

268 *Seasonal Integral Projection Models*

269 We used the most parsimonious models of demographic processes (Table 1) to parameterize
270 density-independent, stage-mass-structured, seasonal and environment-specific Integral
271 Projection Models (IPMs) (Easterling et al., 2000; Ellner et al., 2016). For each stage a , the IPMs
272 track the number of individuals (n_a) in the mass range $[z, z+dz]$ at time t . The fate of these
273 individuals at time $t+1$ is described by a set of coupled integral equations, which differ for each

274 season and are a function of the latent environmental variable Q_y . In the winter season,
275 individuals can survive (θ_w) and change mass (γ_w) according to their stage, mass, and
276 environment. Conditional on survival, juveniles (J) transition to yearlings (Y), while all other
277 stages are distributed to either the reproductive (R) or non-reproductive (N) adult stage at the
278 beginning of summer, depending on the stage-specific probability of reproducing (ϕ_0). During
279 the summer season, individuals in stages Y, N, and R survive (θ_s) and change mass (γ_s)
280 according to their stage and mass at the beginning of summer and according to the environment;
281 but, in summer, transitions to another stage do not occur. Reproductive individuals (R) of a given
282 mass also produce $\phi_1/2$ female juveniles (J), *i.e.*, half of the total number of recruits. Female
283 recruits are distributed across z mass classes by the end of summer, given by ϕ_2 . The
284 mathematical descriptions of the IPMs for the winter and summer seasons are provided in
285 Supporting Material S2. Our population model assumes that past conditions affecting individuals
286 are captured by the current mass distribution and are propagated through time, allowing us to
287 assess trait- and stage-mediated demographic processes (Ozgul et al., 2010).

288 We numerically integrated the summer and winter IPMs using the ‘midpoint rule’
289 (Easterling et al., 2000) with upper and lower integration limits of 7.8 (472 g) and 17.1 (5000 g),
290 respectively. To avoid unintended eviction of individuals from the model (*i.e.*, for a given mass
291 class z , the sum of the probabilities to transition to $z^* < 1$), we applied a constant correction (*i.e.*,
292 equally redistributing evicted individuals among all z^*) when constructing the IPMs as suggested
293 in Merow *et al.*, (2014) (see also Williams et al., 2012). For each stage-specific IPM, we chose a
294 bin size of 100 (*i.e.*, dividing masses into 100 classes), as further increasing the bin size did not
295 significantly improve the precision of estimates of the long-term population growth rate. The

296 IPMs we constructed accurately reproduced observed population dynamics from 1976-2016
297 (Supporting Material S2).

298

299 *Sensitivity of population dynamics to seasonal demographic processes: prospective*
300 *perturbations*

301 Changes in population dynamics in response to changes in environmental fluctuations are
302 determined by the response of demographic processes to the environment and, in turn, of
303 population dynamics to demographic processes (Maldonado-Chaparro et al., 2018). To explore
304 these two sources of variation in the long-term fitness of the marmot population, we first
305 quantified the proportional change in the demographic processes (Table 1) to changes in Q_y , *i.e.*,
306 $\partial(\log \rho)/\partial Q_y$, where ρ is a demographic process. We calculated these elasticities for different
307 values of Q_y (from -1 to 1), increasing each value by 0.01 and keeping mass at its stage-specific
308 average and ε_Y fixed to 0. To assess the effect of parameter uncertainty on our estimates, we
309 repeated these calculations for a sample of 1000 parameter values drawn from the posterior
310 distribution (Paniw et al., 2017).

311 We next assessed which demographic processes most affected the stochastic population
312 fitness under observed (1976-2016) environmental fluctuations. We used a simulation of 100,000
313 years to assess the stochastic population growth rate, $\log \lambda_S$, a measure of fitness (see Supporting
314 Material S3 for details; see section below for short-term viability simulations). During the
315 simulation, we calculated the elasticity of $\log \lambda_S$ to changes in the 40-year observed mean (e_S^μ)
316 and standard deviation (e_S^σ) of stage-specific demographic processes; we adapted the approach
317 described in Ellner *et al.* (2016; chapter 7) to evaluate the relative effects of these changes on \log
318 λ_S (see S3 for details). The two elasticities quantify the strength of selection pressures on lower-

319 level vital rates in stochastic environments (Haridas & Tuljapurkar, 2005; Rees & Ellner, 2009).
320 We repeated the elasticity calculations for a sample of 100 parameter values from the posterior
321 distribution.

322 *Population viability under changes in environmental quality*

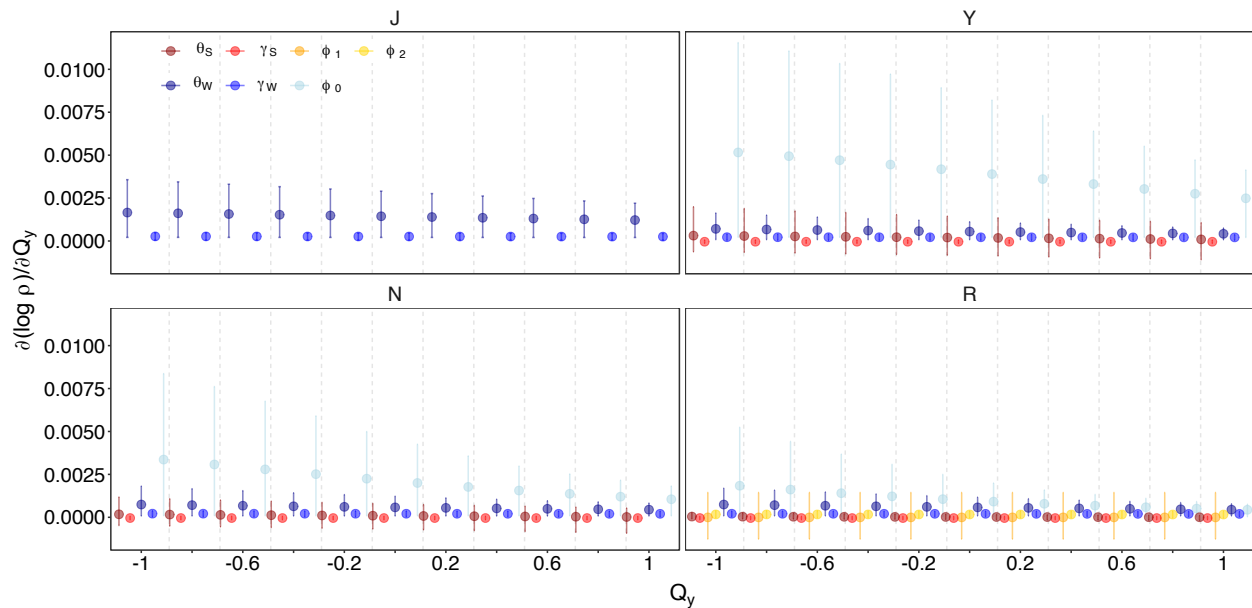
323 To assess how the combined effects of (i) seasonal demographic responses to environmental
324 fluctuations and (ii) population sensitivity to seasonal demography impact population viability,
325 we simulated population dynamics under environmental change. We ran 200 independent
326 simulations each projecting population dynamics for 50 years. The projections were based on
327 several scenarios of changes in the distribution of environmental quality, Q_y , corresponding to
328 changes in the average and standard deviation of winter length and harshness as well as
329 unobserved environmental drivers. We first created base simulations (*i.e.*, no environmental
330 change) where Q_y was picked from a normal distribution with $\mu_Q = 0$ and $\sigma_Q = 1$ across all
331 demographic processes. This was appropriate, as we found no indication of temporal
332 autocorrelation in Q_y (Supporting Material S1). Next, we approximated random future
333 fluctuations in Q_y under different average environmental conditions. To do so, we sampled Q_y
334 from a normal distribution fixing the average environmental quality ($\mu_Q = -1, -0.5, 0.5, 1$) and its
335 variation ($\sigma_Q = 0.6, 1.2$) over the 50 years of projections. We then explored how a trend in μ_Q
336 would affect viability and mass distribution. To do so, we decreased the four μ_Q by 0.01 in each
337 year of the projections, keeping σ_Q unaltered. We also explored population-level effects of future
338 increases in the temporal autocorrelation in Q_y , as detailed in Supporting Material S4. All
339 simulations were repeated for a random sample of 1,000 parameters from the posterior
340 distribution to account for parameter uncertainty.

341 For all environmental-change scenarios, we recorded the probability of quasi-extinction
 342 across the 200 simulations. Quasi-extinction was defined conservatively as the number of non-
 343 juvenile individuals (*i.e.*, yearlings and non-reproductive and reproductive individuals) in the
 344 population to be < 4 , which corresponded to 10 % of their lowest observed number.

345 RESULTS

346 *Sensitivity of population dynamics to seasonal demographic processes*

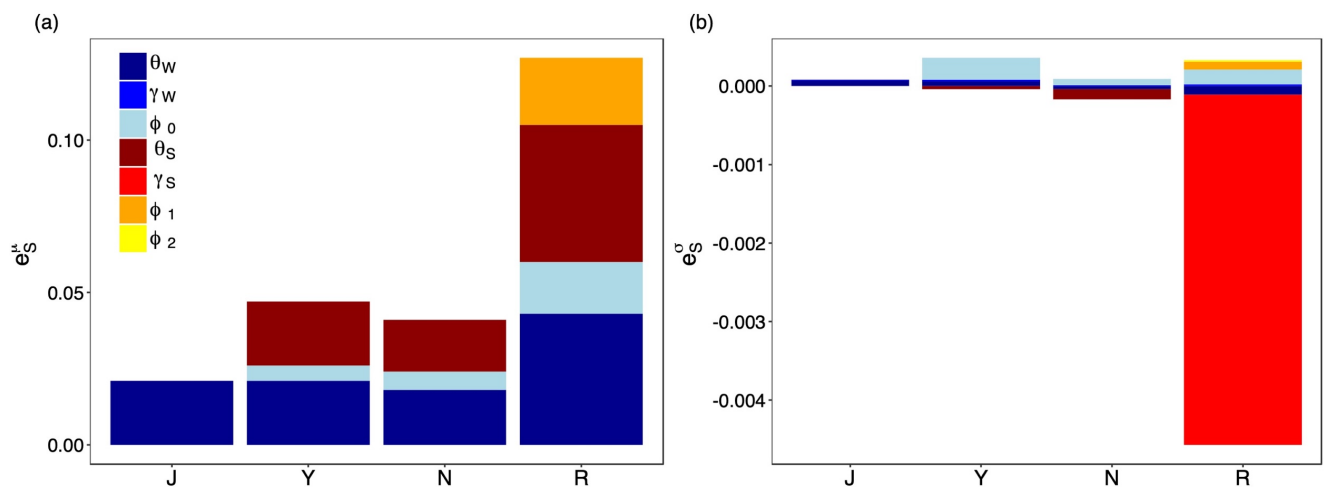
347 In accordance with the posterior distribution of β_q parameters, which did not cross 0 for winter
 348 demographic processes, only winter demographic processes were significantly affected by small
 349 changes in Q_y (Fig. 2). Among the winter demographic processes, changes in Q_y affected
 350 reproduction across stages the most, followed by survival of juveniles (Fig. 2).



351
 352 **Figure 2:** The sensitivity of seasonal demographic processes to environmental quality in marmots. Sensitivity is
 353 assessed as proportional changes in demographic processes, ρ , as environmental quality, Q_y , increases slightly. This
 354 sensitivity is measured with respect to different average values of Q_y and across four different life-cycle stages:
 355 juveniles (J), yearlings (Y), non-reproductive adults (N), and reproductive adults (R). The demographic processes
 356 include winter (W; blue color tones) and summer (S; red color tones) survival (θ) and mass change (γ); and

357 probability of reproducing (ϕ_0), recruitment (ϕ_1), and juvenile mass (ϕ_2). Points and error bars show averages \pm 95
 358 % C.I. across 1,000 posterior parameter samples obtained from the Bayesian population model.
 359

360 While environmental quality affected winter demographic processes only, our
 361 prospective perturbation analyses showed that winter and summer demography equally
 362 determine long-term population fitness. Stochastic elasticity analyses (e_S^μ and e_S^σ) showed that
 363 relative increases in the mean (μ) of winter (θ_W) and summer (θ_S) survival for reproductive
 364 adults (R), would lead to substantial relative increases of the stochastic population growth rate,
 365 $\log \lambda_S$ (Fig. 3a). Highest, positive e_S^μ were found at intermediate and large mass classes, and e_S^μ
 366 was negative for small masses when mass changes (γ) and offspring mass (ϕ_2) were perturbed
 367 (Fig. S3.1a in Supporting Material S3). This explained the overall small e_S^μ for γ and ϕ_2 summed
 368 over all mass classes (Fig. 3a). Overall, relative changes in $\log \lambda_S$ due to increases in the standard
 369 deviation of demographic processes (e_S^σ) were much smaller compared to e_S^μ (Fig. 3b) and didn't
 370 differ significantly between vital rates, as 95 % posterior C.I. crossed 0 (Fig. S3.1b).



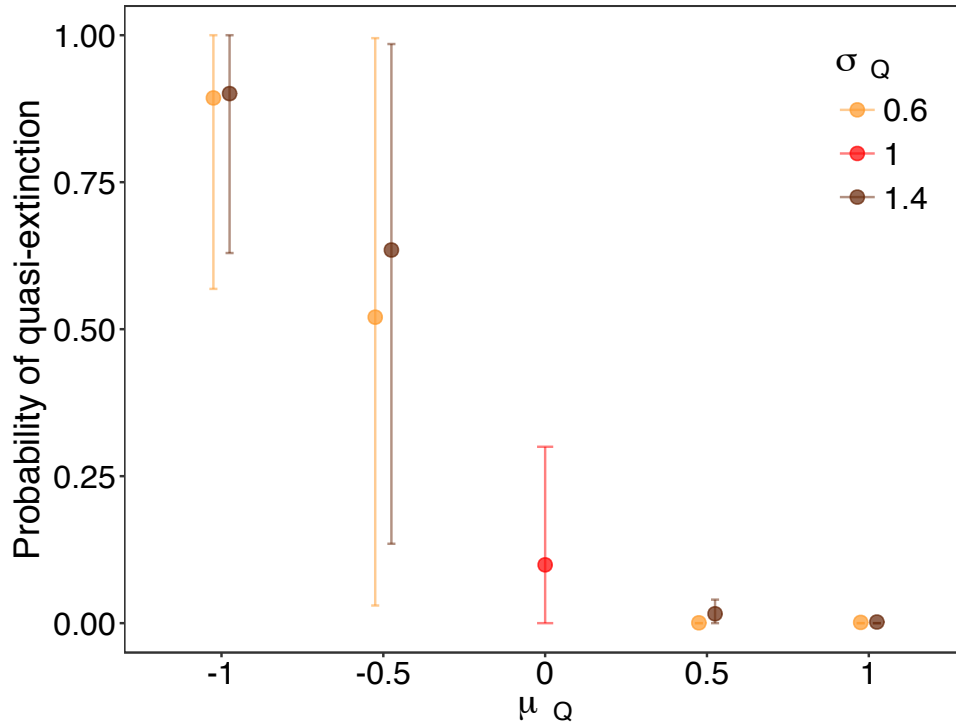
371
 372 **Figure 3** Sensitivity of the average long-term population fitness to changes in the average and variability of
 373 demographic processes modeled for the yellow-bellied marmots. The sensitivity measure is obtained analytically as
 374 elasticities (e) of the stochastic population growth rate, $\log \lambda_S$, to changes in (a) the mean (μ) and (b) standard
 375 deviation (σ) of stage-specific demographic processes summed over all mass classes. Stages are juveniles (J),

376 yearlings (Y), non-reproductive adults (N), and reproductive adults (R). Demographic processes include winter (W)
377 and summer (S) survival (θ) and mass change (γ); reproduction (ϕ_0); recruitment (ϕ_1), and offspring mass
378 distribution (ϕ_2). Elasticities were calculated at the mean posterior values of parameters obtained from the Bayesian
379 demographic model.

380

381 *Population viability under changes in environmental quality*

382 While population fitness was equally sensitive to demographic processes over winter and
383 summer, environmental fluctuations strongly affected viability through winter demography.
384 Using base simulations (*i.e.*, obtaining Q_y from a normal distribution with $\mu_Q = 0$ and $\sigma_Q = 1$),
385 the probability of quasi-extinction, at an average of 0.1 [0.0, 0.3 C.I.] across posterior
386 parameters, were relatively low. Simulations of population dynamics based on scenarios of
387 environmental change, corresponding in part to changes in winter length and harshness, resulted
388 in substantial changes to viability. Quasi-extinction decreased (0 at $\mu_Q = 1$) and increased (0.9
389 [0.6, 1.0 C.I.] at $\mu_Q = -1$), compared to base simulations, when the population experienced a high
390 and low average environmental quality (Q_y), respectively (Fig. 4). Average quasi-extinction
391 further increased and its uncertainty across posterior parameters decreased when a declining
392 trend in Q_y was simulated (Fig. S4.1). Changes in the standard deviation (Fig. 4) and
393 autocorrelation (Fig. S4.2) of Q_y had comparatively little effect on quasi-extinction.



394 **Figure 4:** Probability of quasi-extinction (*i.e.*, < 4 non-juveniles in the population) of yellow-bellied marmots under
 395 different scenarios of environmental change. The scenarios consisted of projecting population dynamics for 50 years
 396 fixing a different mean (μ) and standard deviation (σ) of environmental quality (Q) in all demographic processes.
 397 Points and error bars show averages \pm 95 % C.I. across 1,000 posterior parameter samples obtained from the
 398 Bayesian population model. Base simulations ($\mu_Q = 0$; $\sigma_Q = 1$) are depicted in red.

400

401 DISCUSSION

402 One important pathway through which environmental change can act on population dynamics is
 403 through seasonal direct and carry-over effects on survival, development, and reproduction
 404 (Harrison et al., 2010; Paniw et al., 2019). These effects, however, are often cryptic and therefore
 405 difficult to quantify in ecological models (Hindle et al., 2019). We use a novel, factor-analytic
 406 approach to efficiently quantify partially unobserved environment-demography relationships.
 407 This approach allows us to investigate how positive responses in several demographic processes
 408 to winter environmental conditions can drive annual population dynamics in a winter-adapted

409 mammal. The sensitivity to winter conditions occurs despite the fact that offspring are recruited
410 in summer and both summer and winter demographic processes determine population fitness. As
411 whole-year, population-level effects of environmental change can be filtered by season-specific
412 processes in the absence of density-dependent feedbacks, we highlight that the assessment of
413 such processes allows for a mechanistic understanding of population persistence (Picó et al.,
414 2002; Paniw et al., 2019).

415 In marmots, as in numerous other populations (Bassar et al., 2016; Jenouvrier et al.,
416 2018), seasonal demographic processes play an important role in life-cycle dynamics (Armitage,
417 2017). Our prospective perturbations show that changes in both mean winter and summer
418 survival of reproductive adults have the strongest effect on population fitness, confirming the
419 critical role of this life-cycle stage (Ozgul et al., 2009). At the same time, environmental
420 conditions do not affect adult survival or other demographic processes in the same way
421 throughout the year. That is, although the environment has been shown to drive particularly
422 recruitment in numerous temperate species (e.g., Bonardi et al., 2017; Nouvellet et al., 2013),
423 such effects are not evident in marmots; here, a higher annual environmental quality, which
424 increases all winter demographic processes, shows little impact on summer demography,
425 including recruitment. In turn, only these joint responses of winter demographic processes to
426 environmental quality determine population persistence under environmental change.

427 The complex, partially unmeasured environmental processes that cause positive
428 covariation in seasonal demographic processes can be effectively captured using a univariate,
429 latent measure of environmental quality. In our study, this latent quality correlated better with
430 observed annual population growth than any measured environmental variable (Supporting
431 Material S1). In part, a good quality depicts shorter and milder winters. Milder winters increase

432 food availability and the time available for vigilance, thereby decreasing predation risk (Van
433 Vuren, 2001), especially just before or after hibernation (*i.e.*, within our winter season) when
434 such risk is severe (Armitage, 2014). Predation risk in early spring also increases under high
435 snow cover, as marmots, including more experienced adult females, cannot easily retreat to their
436 burrows (Blumstein, pers. obs.). Predation is however cryptic in the system (Van Vuren, 2001).
437 Capturing the effects of unobserved environmental variation, including predation, the latent-
438 variable approach appears to be a promising alternative to modeling seasonal demographic
439 processes under limited knowledge of their drivers (Evans & Holsinger, 2012; Hindle et al.,
440 2019; Hindle et al., 2018). We note that this approach may find limited applications in cases
441 where environment-demography relationships are more complex than in the yellow-bellied
442 marmots and include negative demographic covariation (e.g., due to opposing environmental
443 effects on demographic rates or tradeoffs between these rates). However, positive covariation in
444 demographic patterns is common (Jongejans et al., 2010; Paniw et al., 2019); and, given the short
445 time series of most demographic datasets (Salguero-Gómez et al., 2015; 2016) or little
446 knowledge on the actual environmental drivers of population dynamics (van de Pol et al., 2016;
447 Teller et al., 2016), the factor-analytic approach can be particularly useful in comparative
448 studies.

449 The seasonal effects of environmental quality on population persistence must be
450 understood in terms of the role of reproductive females in the marmot population (Ozgul et al.,
451 2009). In our simulations, shorter and less severe winters (*i.e.*, a good winter quality), would
452 result in more reproductive females in the summer (Armitage et al., 2003). In turn, summer
453 survival and recruitment by these females are important to long- and short-term demography
454 (Ozgul et al., 2009; Maldonado-Chaparro et al., 2018), but are not driven by environmental

455 conditions. That is, although predation affects individuals in summer (Van Vuren, 2001), its
456 effects are strongest on juveniles and yearlings, while adult females are little affected (Ozgul et
457 al., 2006). At the same time, as is the case in other socially complex mammals (Morris et al.,
458 2011), reproduction in yellow-bellied marmots is governed primarily by social interactions, in
459 particular the behavior of dominant adult females (Armitage, 2010; Blumstein & Armitage,
460 1998). Even under optimal summer conditions, the reproductive output of the population may
461 not increase as dominant females suppress reproduction in younger subordinates and therefore
462 regulate the size of colonies (Armitage, 1991). Dominant females, in addition, may skip
463 reproduction themselves if they enter hibernation with a relatively low mass (Armitage, 2017).
464 Thus, the necessity of meeting the physiological requirements of hibernation profoundly affects
465 life-history traits of yellow-bellied marmots that are expressed during the active season.

466 Unlike the effects of seasonal survival and reproduction, trait transitions between seasons
467 had a smaller effect on annual population dynamics, even if winter mass changes were mediated
468 by environmental quality. These relatively small effects are likely due to the fact that marmots
469 compensate for winter mass loss with increased growth in the summer, creating a zero-net effect
470 on annual trait change (Maldonado-Chaparro et al., 2017; 2018). Although the strength of
471 compensatory effects may differ within seasons or among life-history stages (Monclús et al.,
472 2014), such effects are common in rodents and other species that have a short window for mass
473 gain (Morgan & Metcalfe, 2001; Orizaola et al., 2014), and highlight how assessing seasonal
474 dynamics can provide a mechanistic understanding of population-level global-change effects
475 (Bassar et al., 2016).

476 Under environmental change, the persistence of marmots was mostly affected by changes
477 in mean environmental quality, whereas changes in the variance and temporal autocorrelation of

478 the mean showed little effects. This supports previous conclusions that yellow-bellied marmots
479 are partly buffered against increases in environmental variation (Maldonado-Chaparro et al.,
480 2018; Morris et al., 2008) or autocorrelation (Engen et al., 2013). Further support for
481 demographic buffering comes from the fact that changes in the mean environmental quality most
482 strongly affected those demographic processes to which the stochastic population growth rate
483 was least sensitive, *i.e.*, yearlings gaining reproductive status. It is well known that in species
484 where vital rates of adults are relatively buffered, juveniles are much more sensitive to
485 environmental variation (Gaillard & Yoccoz, 2003; Jenouvrier et al., 2018). Our results indicate
486 that demographic buffering (Pfister, 1998; Morris et al., 2008) likely persists across the seasonal
487 environments and different masses for a high-altitude specialist.

488 Our results emphasize that declines in environmental quality in the non-reproductive
489 season alone can strongly affect annual population dynamics of a mammal highly adapted to
490 seasonal environments. Therefore, positive demographic covariation under environmental
491 change may threaten populations even if it affects demographic process to which the stochastic
492 growth rate is least sensitive, *i.e.*, processes that are under low selection pressure (Coulson et al.,
493 2005; Iles et al., 2019). Studies that focus on the effects of environmental factors on single
494 demographic processes that strongly affect both short- and long-term population dynamics may
495 therefore underestimate the important role of seasonal demographic covariation.

496 Most species inhabit seasonal environments. Under global environmental change, it may
497 therefore be critical to understand how seasonal patterns mediate persistence of natural
498 populations. Novel methods such as the factor analytic approach allow researchers to overcome
499 some challenges associated with more mechanistic approaches assessing population responses to

500 environmental change, and we encourage more seasonal demographic analyses across different
501 taxa.

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