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

Authors: Maria Paniw[1,2], Dylan Childs[3], Kenneth B Armitage[4], Daniel T Blumstein[5,6], Julien Martin[7,8], Madan K. Oli[9], Arpat Ozgul[1]

1 - Department of Evolutionary Biology and Environmental Studies, University of Zurich, Winterthurerstrasse 190, CH-8057 Zurich, Switzerland, arpat.ozgul@ieu.uzh.ch
2 - Ecological and Forestry Applications Research Centre (CREAF), Campus de Bellaterra (UAB) Edifici C, ES-08193 Cerdanyola del Vallès, Spain, m.paniw@creaf.uab.cat
3 - Department of Animal and Plant Sciences, University of Sheffield, Western Bank, Sheffield S10 2TN, UK, d.childs@sheffield.ac.uk
4 - Ecology & Evolutionary Biology Department, The University of Kansas, Lawrence, KS 66045-7534, USA, marmots@ku.edu
5 - Department of Ecology and Evolutionary Biology, University of California Los Angeles, Los Angeles, CA 90095, USA, marmots@ucla.edu
6 - The Rocky Mountain Biological Laboratory, Crested Butte, CO 81224, USA
7 - School of Biological Sciences, University of Aberdeen, Aberdeen, AB24 2TZ, UK,
8 – Department of Biology, University of Ottawa, Ottawa, K1N 9A7, Canada, *julien.martin@uottawa.ca*
9 - Department of Wildlife Ecology, University of Florida, Gainesville, FL 32611, USA, olim@ufl.edu

*Corresponding author: Tel.: +34-671-246-338; email: maria.paniw@ieu.uzh.ch; ORCID ID: 0000-0002-1949-4448

ABSTRACT

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

INTRODUCTION

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

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

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

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

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

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

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

METHODS

Study species

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

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

Seasonal demographic rates and trait transitions

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

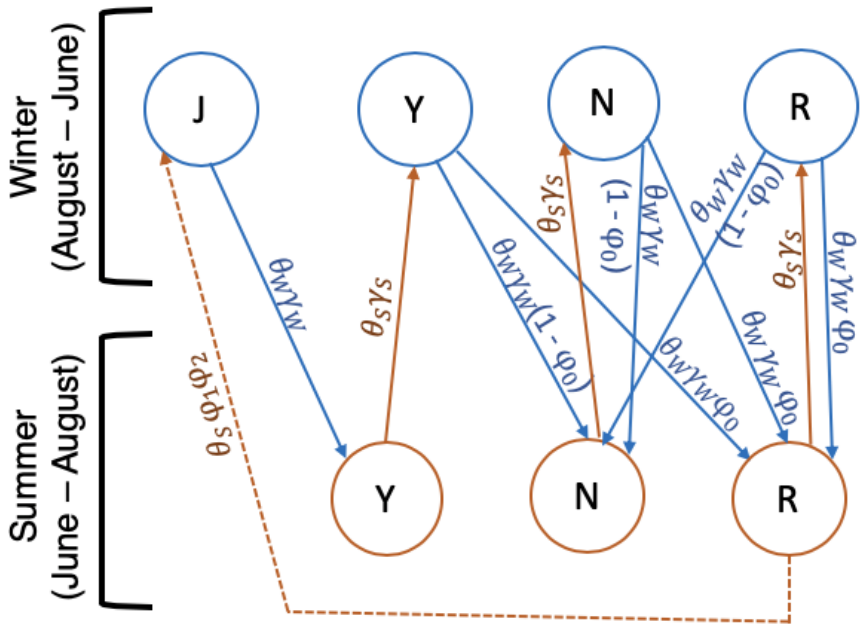

Figure 1: Seasonal life-cycle transitions modelled for yellow-bellied marmots. The two seasons correspond to the main periods of mass loss (winter) and gain (summer). Solid and dashed arrows represent discrete-time stage transitions and recruitment, respectively. Transitions among winter (W) and summer (S) stages (marked by arrows in different colors) depend on demographic rates (survival [θ], reproduction [φ_0], and recruitment [φ_1]) and trait transitions (mass change [γ], and offspring mass [φ_2]). Stages are: juveniles, J, yearlings, Y, non-reproductive adults, N, and reproductive adults, R. All stage-specific demographic rates and trait transitions are modeled using generalized linear mixed effects models in a Bayesian framework and include body mass and a common latent variable representing environmental quality as covariates.

We determined demographic rates (survival, reproduction, and recruitment) for two discrete growing seasons: winter (August - June) and summer (June - August) (Fig. 1), delineating the main periods of mass loss and gain, respectively (Maldonado-Chaparro et al., 2017). We assumed that females that permanently disappeared from a colony had died. This measure of apparent survival may overestimate the death of yearlings in the summer, which disperse from their natal colonies (Van Vuren & Armitage, 1994). At the same time, the intensive trapping protocol ensured a high capture probability of yearlings (Oli & Armitage, 2004), decreasing the discrepancies between their apparent and true survival.

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

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

Modelling covariation in demographic processes – latent-variable approach

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

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

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

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

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

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

Demographic process	Function	Likelihood distribution
Winter (W):		
Survival (θ_W)	$\text{logit}(\theta_W) = \alpha_{0\theta W} + \alpha_{a\theta W}[\text{stage}] + \beta_{z\theta W} \times z + \beta_{q\theta W} \times Q_y[\text{year}] + \varepsilon_{y\theta W}[\text{year}]$	$B(\theta_W)$
Mass next (z_W^*)	$z_W^* = \alpha_{0z*W} + \alpha_{az*W}[\text{stage}] + (\beta_{zz*W} + \beta_{zaz*W}[\text{stage}]) \times z + \beta_{qz*W} \times Q_y[\text{year}] + \varepsilon_{yz*W}[\text{year}]$	$\aleph(z_W^*, \tau_{z*W})$
Reproduction (φ_0)	$\text{logit}(\varphi_0) = \alpha_{0\varphi 0} + \alpha_{a\varphi 0}[\text{stage}] + \beta_{z\varphi 0} \times z + \beta_{q\varphi 0} \times Q_y[\text{year}] + \varepsilon_{y\varphi 0}[\text{year}]$	$B(\varphi_0)$
Summer (S):		
Survival (θ_S)	$\text{logit}(\theta_S) = \alpha_{0\theta S} + \alpha_{a\theta S}[\text{stage}] + \beta_{z\theta S} \times z + \beta_{q\theta S} \times Q_y[\text{year}] + \varepsilon_{y\theta S}[\text{year}]$	$B(\theta_S)$
Mass next (z_S^*)	$z_S^* = \alpha_{0z*S} + \alpha_{az*S}[\text{stage}] + (\beta_{zz*S} + \beta_{zaz*S}[\text{stage}]) \times z + \beta_{qz*S} \times Q_y[\text{year}] + \varepsilon_{yz*S}[\text{year}]$	$\aleph(z_S^*, \tau_{z*S})$
Number of recruits (φ_1)	$\log(\varphi_1) = \alpha_{0\varphi 1} + \beta_{z\varphi 1} \times z + \beta_{q\varphi 1} \times Q_y[\text{year}] + \varepsilon_{y\varphi 1}[\text{year}]$	$P(\varphi_1)$
Juvenile mass (z_J^*)	$z_J^* = \alpha_{0z*J} + \beta_{zz*J} \times z_M + \beta_{qz*J} \times Q_y[\text{year}] + \varepsilon_{yz*J}[\text{year}]$	$\aleph(z_J^*, \tau_{z*J})$

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

The latent variable, Q_y, effectively captured the covariation among the demographic processes (Supporting Material S1); therefore, using one latent variable across both seasons was sufficient. Our GLMMs showed a strong effect of Q_y on winter but not summer demographic processes. This effect was positive for all winter demographic processes, as evidenced by the positive β_q (Table S1.1). The β_q for demographic processes in the summer, however, were comparatively

small and were not significantly different from 0 (95 % posterior C.I.s overlapped 0). The positive β_q indicate that Q_y effectively estimates the overall annual environmental quality or suitability, capturing both biotic and abiotic processes. A positive value of Q_y then depicts an environmental condition at a given time point that increases winter survival and probability of reproducing and decreases mass loss (Hindle et al., 2018). The variation in Q_y was in part explained by environmental variables measured at the study site, but was unrelated to population density (Supporting Material S1). Negative values of Q_y were associated with longer and more severe winters and a higher snow cover, while positive Q_y indicated warmer winters and springs. However, as the environmental variables explained < 50 % of the variation in Q_y, the latent variable captures multivariate, partly unobserved biotic and abiotic processes into a simple, univariate measure of how bad ($Q_y < 0$) or good ($Q_y > 0$) environmental conditions are likely to affect marmot demography.

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

Seasonal Integral Projection Models

We used the most parsimonious models of demographic processes (Table 1) to parameterize density-independent, stage-mass-structured, seasonal and environment-specific Integral Projection Models (IPMs) (Easterling et al., 2000; Ellner et al., 2016). For each stage a, the IPMs

track the number of individuals (n_a) in the mass range [z, z+dz] at time t. The fate of these individuals at time t+1 is described by a set of coupled integral equations, which differ for each season and are a function of the latent environmental variable Q_y. In the winter season, individuals can survive (θ_W) and change mass (γ_W) according to their stage, mass, and environment. Conditional on survival, juveniles (J) transition to yearlings (Y), while all other stages are distributed to either the reproductive (R) or non-reproductive (N) adult stage at the beginning of summer, depending on the stage-specific probability of reproducing (φ_0). During the summer season, individuals in stages Y, N, and R survive (θ_S) and change mass (γ_S) according to their stage and mass at the beginning of summer and according to the environment; but, in summer, transitions to another stage do not occur. Reproductive individuals (R) of a given mass also produce $\varphi_1/2$ female juveniles (J), *i.e.*, half of the total number of recruits. Female recruits are distributed across z mass classes by the end of summer, given by φ_2. The mathematical descriptions of the IPMs for the winter and summer seasons are provided in Supporting Material S2. Our population model assumes that past conditions affecting individuals are captured by the current mass distribution and are propagated through time, allowing us to assess trait- and stage-mediated demographic processes (Ozgul et al., 2010).

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

significantly improve the precision of estimates of the long-term population growth rate. The IPMs we constructed accurately reproduced observed population dynamics from 1976-2016 (Supporting Material S2).

Sensitivity of population dynamics to seasonal demographic processes: prospective perturbations

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

We next assessed which demographic processes most affected the stochastic population fitness under observed (1976-2016) environmental fluctuations. We used a simulation of 100,000 years to assess the stochastic population growth rate, log λs, a measure of fitness (see Supporting Material S3 for details; see section below for short-term viability simulations). During the simulation, we calculated the elasticity of log λs to changes in the 40-year observed mean (e_S^{μ}) and standard deviation (e_S^{σ}) of stage-specific demographic processes; we adapted the approach described in Ellner *et al*. (2016; chapter 7) to evaluate the relative effects of these changes on log

λs (see S3 for details). The two elasticities quantify the strength of selection pressures on lower-level vital rates in stochastic environments (Haridas & Tuljapurkar, 2005; Rees & Ellner, 2009). We repeated the elasticity calculations for a sample of 100 parameter values from the posterior distribution.

Population viability under changes in environmental quality

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

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

RESULTS

Sensitivity of population dynamics to seasonal demographic processes

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

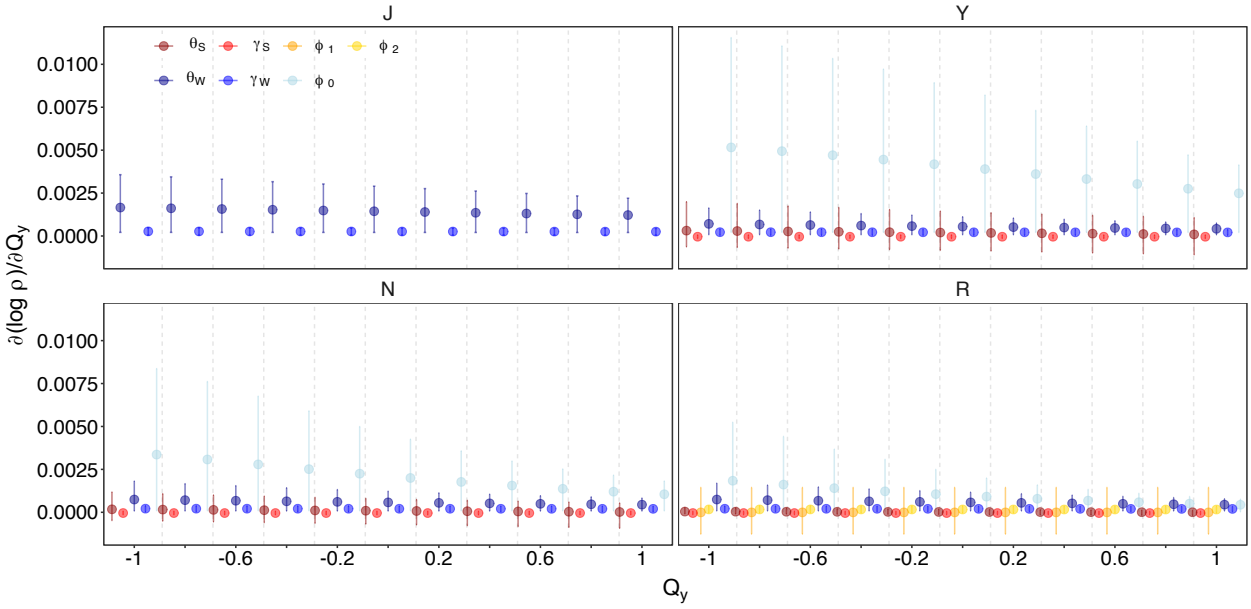

Figure 2: The sensitivity of seasonal demographic processes to environmental quality in marmots. Sensitivity is assessed as proportional changes in demographic processes, ρ, as environmental quality, Q_y, increases slightly. This sensitivity is measured with respect to different average values of Q_y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

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

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

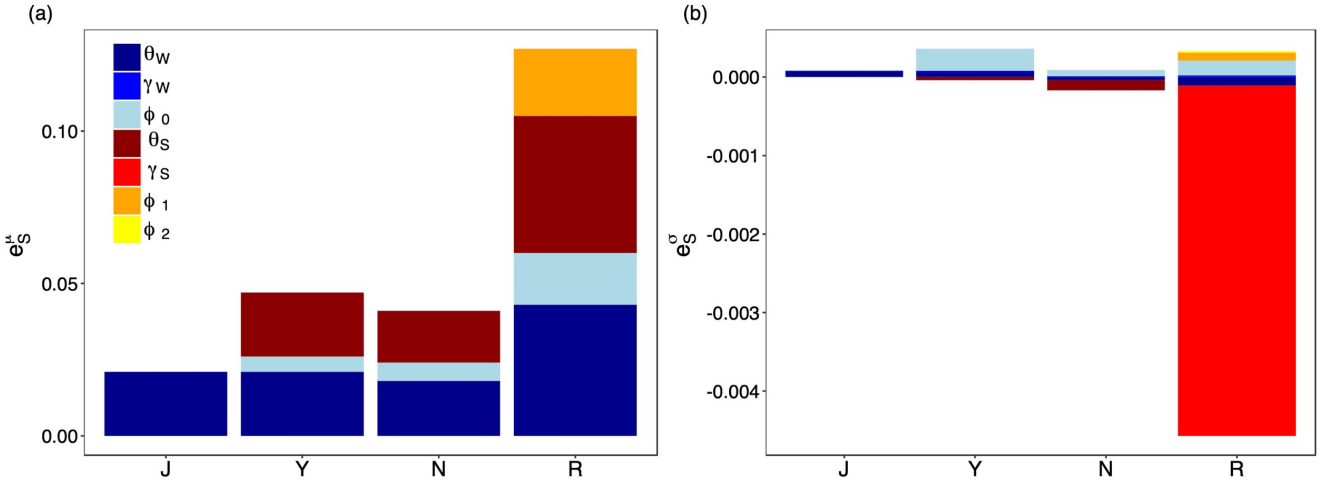

Figure 3 Sensitivity of the average long-term population fitness to changes in the average and variability of demographic processes modeled for the yellow-bellied marmots. The sensitivity measure is obtained analytically as elasticities (e) of the stochastic population growth rate, log λs, to changes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

379 yearlings (Y), non-reproductive adults (N), and reproductive adults (R). Demographic processes include winter (W)
380 and summer (S) survival (θ) and mass change (γ); reproduction (φ_0); recruitment (φ_1), and offspring mass
381 distribution (φ_2). Elasticities were calculated at the mean posterior values of parameters obtained from the Bayesian
382 demographic model.
383

384 *Population viability under changes in environmental quality*

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

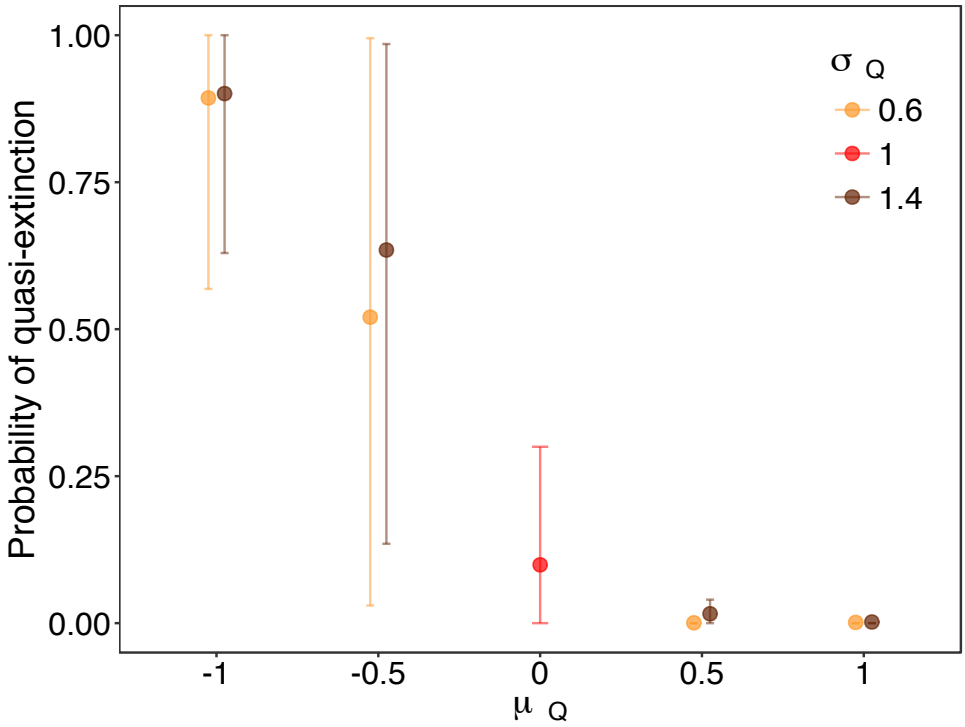

Figure 4: Probability of quasi-extinction (*i.e.*, < 4 non-juveniles in the population) of yellow-bellied marmots under different scenarios of environmental change. The scenarios consisted of projecting population dynamics for 50 years fixing a different mean (μ) and standard deviation (σ) of environmental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

DISCUSSION

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

412 mammal. The sensitivity to winter conditions occurs despite the fact that offspring are recruited
413 in summer and both summer and winter demographic processes determine population fitness. As
414 whole-year, population-level effects of environmental change can be filtered by season-specific
415 processes in the absence of density-dependent feedbacks, we highlight that the assessment of
416 such processes allows for a mechanistic understanding of population persistence (Picó et al.,
417 2002; Paniw et al., 2019).

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

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

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

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

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

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

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

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

491 Our results emphasize that declines in environmental quality in the non-reproductive
492 season alone can strongly affect annual population dynamics of a mammal highly adapted to
493 seasonal environments. Therefore, positive demographic covariation under environmental
494 change may threaten populations even if it affects demographic process to which the stochastic
495 growth rate is least sensitive, *i.e.*, processes that are under low selection pressure (Coulson et al.,
496 2005; Iles et al., 2019). Studies that focus on the effects of environmental factors on single
497 demographic processes that strongly affect both short- and long-term population dynamics may
498 therefore underestimate the important role of seasonal demographic covariation.

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

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

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