Climate variability and density-dependent population dynamics: Lessons from a simple High Arctic ecosystem

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Eco-configurations are still puzzled by the diverse population dynamics of herbivorous small mammals that range from high-amplitude, multiannual cycles to stable dynamics. Theory predicts that this diversity results from combinations of climatic seasonality, weather stochasticity, and density-dependent food web interactions. The almost ubiquitous 3- to 5-y cycles in boreal and arctic climates may theoretically result from bottom-up (plant–herbivore) and top-down (predator–prey) interactions. Assessing, empirically, the roles of such interactions and how they are influenced by environmental stochasticity has been hampered by food web complexity. Here, we take advantage of a uniquely simple High Arctic food web, which allowed us to analyze the dynamics of a graminivorous vole population not subjected to top-down regulation. This population exhibited high-amplitude, noncyclic fluctuations—partly driven by weather stochasticity. However, the predominant driver of the dynamics was overcompensatory density dependence in winter that caused the population to frequently crash. Model simulations showed that the seasonal pattern of density dependence would yield regular 2-y cycles in the absence of stochasticity. While such short cycles have not yet been observed in mammals, they are theoretically plausible if graminivorous vole populations are deterministically bottom-up regulated. When incorporating weather stochasticity in the model simulations, cyclicality became disrupted and the amplitude was increased—akin to the observed dynamics. Our findings contrast with the 3- to 5-y population cycles that are typical of graminivorous small mammals in more complex food webs, suggesting that top-down regulation is normally an important component of such dynamics.

Seasonality | Bottom-up regulation

Theory suggests that contrasting population dynamics result from details in the pattern of density dependence, including its strength, whether it acts instantly or with a delay, and how it interacts with deterministic (seasonal) and stochastic (weather) components of the prevailing or changing climate (1–5). Studies of small rodents have contributed much to elucidating the different facets of density-dependent and density-independent population dynamics (2, 4). A central topic has been what sort of density dependence yields the high-amplitude, multiannual population cycles for which voles and lemmings have been so renowned (6–9). Based on time series analyses, delayed density dependence is considered to be a main determinant of population cycles (see refs. 4, 10 for reviews), although overcompensatory direct density dependence appears to be an alternative in some settings (11). As rodent cycles are most prevalent in northern ecosystems with profound climatic seasonality (refs. 6, 7, 12, but see refs. 13, 14), several studies have emphasized that annual density dependence ought to be decomposed into its seasonal components (15–17)—both to accurately account for the density-dependent structure that underlies the observed dynamics and to identify the season-specific biotic mechanisms that cause density dependence. Considering seasonal dynamics is also crucial to assessing the role of climatic change and weather stochasticity because both differ between summer and winter (15, 18). The role of climate forcing is now also emphasized by the recent collapses and dampening of population cycles in several ecosystems that appear to be associated with ongoing climate change (15, 19, 20).

Linking density dependence to the biotic mechanisms that causally generate the diversity of population dynamics patterns seen in small mammals has proved to be challenging. Most rodent populations are imbedded in complex food webs and, hence, simultaneously subjected to a multitude of biotic interactions that could cause the different facets of density-dependent population growth. For instance, density dependence may result from both top-down and bottom-up trophic interactions as well as intrinsic population mechanisms (11, 21, 22). While field experiments have helped pinpoint some mechanisms (23–27), they have been too short term to be conclusive with respect to what generates different patterns of multiannual population dynamics.

Here, we apply an approach that has proved useful for unraveling the effects of density dependence and weather stochasticity in herbivorous large mammals (e.g., refs. 28–30), namely to target populations that are found in exceptionally simple biotic settings. Hence, our study targets a High Arctic population of the graminivorous (grass-eating) East European vole (Microtus levis) in a food web that lacks significant top-down regulation (i.e., predation). By combining statistical analyses of long-term, high-quality live-trapping data with simulations of a population model parameterized from these data, we 1) estimate the seasonal density dependence and resultant population dynamics (e.g., cyclic or noncyclic) that emerge in such a simple biotic setting and 2) assess how climatic population fluctuations | Tundra ecosystem | Trophic interactions | Seasonality | Bottom-up regulation

Significance

Whether the renowned population cycles of small mammals in northern food webs are driven by bottom-up (plant–herbivore) or top-down (predator–prey) interactions is still a debated question but crucial to our understanding of their ecological functions and response to climate change. A long-term study of a graminivorous vole population in an exceptionally simple High Arctic food web allowed us to identify which population dynamics features are present without top-down regulation. Unique features were high-amplitude, noncyclic population fluctuations driven by a combination of stochastic weather events and season-specific density dependence likely arising from plant–herbivore interactions. That such features are not present in more complex food webs points to the importance of top-down regulation in small mammal populations.

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seasonality and weather stochasticity in terms of rain-on-snow (ROS) events in winter impinge on such density-dependent population dynamics. Finally, we point out how the insights from this unique case study shed light on the longstanding puzzle about what generates population cycles and how ongoing climate change may influence these cycles.

Results

Annual Population Dynamics. The 1991 to 2007 time series of Eastern European vole densities from the Ridge trapping area was characterized by high-amplitude population fluctuations with 2 to 4 y between subsequent crash years (Fig. 1). The population dynamics appeared to be stationary (i.e., there was no evidence for temporal trends in mean or variance in densities over the 18 y). The amplitude of the fluctuations (s-index = 0.57) was within the range found in population time series of cyclic Arctic lemming populations (31). However, in contrast to most other Arctic populations, first- ($r = -0.39$) and second-order ($r = 0.02$) autocorrelation coefficients showed no evidence of cyclic dynamics.

Seasonal Density Dependence. Population densities in the Core trapping area showed the same annual pattern as densities in the Ridge area (cross-correlation of vole densities with the Ridge trapping area, $r = 0.92$; Fig. 1). Within the summer season, monthly population growth rates were all positive (Figs. 2 and 3A). This suggests that vole population densities remained below carrying capacity in summer. Still, there was evidence for negative density dependence in the monthly population growth rates (Table 1 and Fig. 3A). Detailed demographic analyses suggested that the density dependence in population growth in summer was partly due to density-dependent survival in both sexes (SI Appendix, Table S2 and Fig. S2) and strong density-dependent regulation of adult male densities (SI Appendix, Fig. S2).

In contrast, winter monthly population growth rates were negative in many of the years. This difference between seasons was due to maximum population growth rate in winter (at low population densities and ROS) being one-third the maximum population growth rate in summer ($\beta_0$ in Table 1), while the effect size of density was similar in the two seasons ($\beta_2$ in Table 1). In addition, negative population growth rates in winter were associated with high levels of ROS in the winter (Table 1 and Fig. 3B). Overall, the data suggest that density dependence in winter led to strong overcompensatory population regulation (i.e., negative realized growth).

Simulated Population Dynamics. Simulation of the deterministic version of the seasonal density-dependent model (Eqs. 2, 5, and 6) generated stable 2-y vole cycles. These 2-y cycles were relatively robust to changes in climate severity in winter, in that high ROS must become the norm before we expect a change to stable dynamics with a single equilibrium density (Fig. 4A). The 2-y cycle in the baseline model was also robust to changes in season lengths as climate change would have to reduce the winter length to well below 8 mo for more complex dynamics to appear (Fig. 4B). However, the signal of the 2-y cycles deteriorated rapidly with increasing levels of stochastic process error (Fig. 4C). At the observed levels of process error, as generated by stochastic variation in winter ROS, the expected second-order autocorrelation was close to zero in the model simulations ($r = 0.02$) (i.e., similar to what was estimated from the time series data). Finally, temporal variability as quantified by the s-index increased with increasing process error (Fig. 4D).

Discussion

Without significant top-down regulation from predators or interspecific competition, the focal study system is essentially reduced to a simple two-link food chain consisting of a multivoltine herbivore population and their graminoid food plants in a profoundly seasonal environment. Here, we have presented an empirical analysis of such an ecological system that previously has been subjected to only theoretical investigations. Such systems have been modeled mechanistically in continuous time to identify under which circumstances multiannual herbivore population cycles can be expected (32, 33). Moreover, theoreticians have thoroughly investigated the dynamical properties of phenomenological discrete-time models with seasonal density dependence (34), akin to the model we parameterized here with field data. The core insight from this theory—and indeed also our empirical study—is that the profound seasonality destabilizes the dynamics of such simple systems (35). Profound seasonality in terms of a long Arctic winter without primary production implies that the carrying capacity in summer greatly exceeds that of the winter. Such an environmental setting combined with a multivoltine life history and rapid population growth in summer allows the herbivore population to overshoot its winter carrying capacity. We provide evidence that such a situation prevails in the graminivorous East European voles on Svalbard. While density dependence frequently causes population crashes over winter, the population does not reach the higher densities needed for negative population growth to occur over summer.

Fig. 1. Picture of the Core and Ridge areas for live trapping East European voles near Grumantbyen on Svalbard (A) and time series of vole densities (B) estimated in August in the Core area (gray shape, lines, and points; years 1990 to 1996 and 2002 to 2006) and the Ridge area (black shape, lines, and points; years 1991 to 2007). Error bars represent 95% CIs.
It is well known that discrete-time models with seasonal density dependence, like our baseline model, can generate 2-y cycles (34). Interestingly, the continuous-time, plant–herbivore model analyzed by Turchin and Batzli (33)—with a graminoid-type plant regrowth function, multivoltine herbivore population dynamics, and High Arctic seasonality without stochasticity—also generates 2-y cycles, even though the seasonal dynamics is very different from what is seen in the discrete-time models and our empirical data (SI Appendix). This emphasizes the general propensity of seasonality to generate 2-y cycles in this system. Still, we are not aware that such short population cycles have ever been reported for any mammal population. Neither are we aware of any other examples of the noncyclic high-amplitude, high-frequency boom–bust vole population dynamics we observed in our High Arctic study system. In multivoltine rodents, cycle lengths typically vary between 3 and 5 y in ecosystems with profound seasonality, while populations in environments with less pronounced seasonality have typically noncyclic, low-amplitude fluctuations that often are categorized as stable dynamics (36, 37). It is commonly assumed that cycle-generating mechanisms have delayed effects on population density, such as extended periods of food depletion or predation over more than one phase (4, 10, 12, 21), while opinions differ about which mechanisms are in place (38, 39). Seasonality is in itself a source of delay in producer–consumer interactions (32) but is expected to yield, at most, 2-y cycles. Also, the rapid regrowth of graminoids (33)—even after high vole peak densities and severe winter grazing (40)—prevents longer delays that may generate the longer cycles often found in graminivorous voles. Our results also imply that delayed intrinsic mechanisms, such as stress-induced maternal effects (41), were not present. Finally, our study is consistent with the hypothesis that the almost ubiquitous guild of specialist rodent predators in boreal and Arctic food webs normally cause the delays in density dependence that generate the longer vole cycles in that the predator guild was lacking in our study system (39, 42, 43). Indeed, the unique absence of top-down regulation by specialist predators in High Arctic Svalbard is the most likely cause of the exceptional vole population dynamics observed. Experimental predator removals (24, 27, 40) have never been conducted at a sufficient spatial and temporal scale to investigate whether a similar outcome would appear in other vole populations released from top-down regulation.

A fundamental question in population ecology accentuated by global climate change is how abiotic environmental variation can modify the effect of density-dependent biotic interactions. Our study adds to previous studies showing that episodes of mild winter weather in boreal and Arctic ecosystems, through increased ROS events and changes in snow conditions, may lead to population crashes in herbivores (44–46) and disrupt population cycles (18–20, 47). Previous models have shown that climatically disrupted population cycles in multivoltine rodents readily collapse to low-amplitude fluctuations and, hence, stable population dynamics (19, 20, 47). Here, we have shown that population cycles in a very simple trophic system may also be disrupted by increasing weather stochasticity but without any dampening effect on the dynamics. Hence, our case study provides support to the general conjecture that the impact of climate change on ecological systems depends on their structure and hence can be expected to be diverse across time and space (18, 19, 29).

Materials and Methods

Study Population. Our study was located at Grumant in Svalbard (78.18°N, 15.13°E). This High Arctic location is characterized by cool summers (July average: 5.9 °C) and cold winters (January average: −16.2 °C) with little precipitation (average 190 mm; period 1960 to 1990; ref. 48). Average daily air temperatures are typically above 0 °C from early to mid-June to September (data from Longyearbyen airport, ~13 km away from Grumant). Winter temperatures are much more variable than summer temperatures (48, 49). ROS (measured as amount of precipitation that fell at temperature above 1 °C from November to April) are relatively frequent stochastically
occurring weather events (30) and have been found to strongly influence the population dynamics of all year-round resident vertebrate populations in Svalbard (50).

The East European vole belongs to one of the most speciose and widespread genera (Microtus) of small mammals (51). Most Microtus species are graminivorous (grass-eating) and have multivoltine life histories (e.g., multiple generations per year). The East European vole was accidentally introduced to Svalbard in the first half of the 20th century (52). The voles have a highly restricted distribution on the archipelago, associated with seabird fertilized tundra vegetation dominated by graminoids (53).

There are no other small mammals present in Svalbard, which suggests no interspecific competition (54). The Arctic fox (Vulpes lagopus) is the only terrestrial predator present but acts as a generalist carnivore that mostly relies on large colonies of seabirds in the study area (55). Thus, the focal food web lacks the guild of specialist predators consisting of mustelids, owls, hawks, and jaegers that are almost omnipresent in the Arctic (12). East European voles reproduce quickly with females observed to be gravid as early as 17 d old (56) and may live at high population densities (>100 individuals · ha⁻¹; ref. 54). Many Microtus populations have been studied extensively on the European continent where they typically exhibit 3- to 5-y multiannual cycles (11, 13–15).

**Live Trapping.** East European voles were live trapped during the years 1990 to 2007. Here, we analyzed two datasets. The main dataset used for analysis of seasonal density dependence and demography was obtained from one of the largest and lushest vole habitat patches in Svalbard, hereafter termed Core area (Fig. 1). From 1990 to 1996, we used a trapping grid of 93 Ugglan Special multiple-capture traps encompassing 4.5 ha of the Core area, while from 2002 to 2006, the grid was made of 74 traps and encompassed 2.8 ha. Traps were separated by ~20 m and placed by burrow entrances wherever possible. Trapping of the Core area followed the robust design of Pollock (57) and consisted of three primary periods (late June/early July [hereafter termed July] = P1; early August = P2; and early September = P3), each with 6 to 10 secondary periods, except in 1990, when primary periods spanned only the first half of the summer. The primary trapping periods consisted of traps being checked at 1300 and 1900 after baiting with oats and potatoes in the morning (0700, i.e., two secondary periods per day). Traps were deactivated during the last trapping period of the day. We used a second, more long-term dataset obtained from a linear habitat on a ridge and vegetated part of the ravine at the western edge of the Core area (hereafter termed Ridge; Fig. 1) for assessing the annual population dynamics. The Ridge area was monitored with 30 traps in August during 1991 to 2007, with the same number of secondary periods as for the Core area. Captured voles were marked by toe clipping, sexed, and weighed. The study was conducted according to the regulations for research in Svalbard during the study period.

**Density Estimation.** The densities of the vole population in the Core and Ridge areas were estimated by spatially explicit capture-recapture (SECR) models with the package secr in R (58–60). Briefly, these models have the advantage of estimating capture probabilities based on the distance separating the center of activities of an individual from a trap (60). By using a two-parameter half-normal detection function, we obtained more accurate estimates of the area effectively surveyed (61). For the Core area, we used one null SECR model per trapping period per year with the Huggins parameterization to estimate density (61). For the Ridge area, annual densities were obtained for the August trapping period only. Densities of male and female adults (body mass ≥25 g) and subadults (body mass <25 g; ref. 56) were derived from this general model. Because the movement of voles in Svalbard is restricted by habitat, we used a 20-m buffer around traps to build the state-space that is used to estimate effective sampling area. The Nelder–Mead algorithm was used for optimization of the likelihood in model fitting.

**Annual Density-Dependent Structure and Temporal Variability.** We used the annual early August SECR density estimates from the Ridge area to assess the long dynamics of the system. The presence of cyclic dynamics was assessed using autocorrelations at different time lags. We used the SD of the log10-transformed time series as a metric for the temporal variability (i.e., the amplitude) of the multiannual dynamics (s-index; ref. 62). The s-index has been used both to define cyclic dynamics (index values >0.5; ref. 63) and to compare populations across environmental gradients (31, 36, 64).

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**Table 1.** Parameter estimates for the best models for monthly population growth of Eastern European voles (Eqs. 1–5) over winter and in the summer period and estimates of the Bayesian $R^2$ for the models

| Parameter | Winter | Summer |
|-----------|--------|--------|
| $\beta_0$ | 0.23 (0.14, 0.33) | 0.64 (0.44, 0.83) |
| $\beta_0$ | -0.0040 (-0.0058, -0.0024) | -0.0051 (-0.0095, -0.0006) |
| $\rho_{\text{ROS}}$ | -0.0043 (-0.0074, -0.0014) | — |
| $\sigma_1^2$ | 0.005 (0.001, 0.020) | 0.020 (0.005, 0.058) |
| Bayesian $R^2$ | 0.92 (0.73, 0.98) | 0.40 (0.02, 0.76) |

No ROS effect was included in the model for summer population growth, giving no estimate of $\beta_{\text{ROS}}, \beta_0 = \text{coefficient for density at time } t; \rho_{\text{ROS}} = \text{coefficient for effect of ROS}; \text{and } \sigma_1^2 = \text{process error variance.}
Furthermore, we modeled process error affects equally winter and summer population growth. Estimates of process error variance in models without a ROS effects were the log normal measurement error SDs, estimated from annual observations of ROS (in millimeters) in winter (ROS_w) and residual stochastic variation in r (process error, e_r), measured by annual observations of ROS every winter using the population model with no process error (ROS = 0, e_r² = 0). Estimates of the second-order autocorrelation (C) and the amplitude of fluctuations (s-index) in autumn densities in the baseline population model for increasing values of process error variance (e_r²) and ROS = 0. (Eqs. 3.7). In C and D, the long, dashed line represents a model where process error only affects the summer population growth, the short, dashed line a model where process error only affects the winter population growth, and the solid line represents a model where process error affects equally winter and summer population growth. Estimates of process error variance in models without a ROS effects were σ_w² = 0.014 and σ_s² = 0.020 for winter and summer, respectively. Estimated second-order autocorrelation in the Ridge area in August was 0.02, while the s-index was 0.57.

Seasonal Density Dependence and Climate Effects on Population Growth. We used the following model to explore patterns of variation in population growth in summer and winter in the Core area:

\[ X_{t+1} = X_t \cdot e^{r_t + \Delta t}, \]  

where \( X_t \) is the true population density at time \( t \), \( r_t \) is the population growth rate from \( t \) to \( t+1 \), and \( \Delta t \) is the time period from \( t \) to \( t+1 \) (in months). Furthermore, we modeled \( r_t \) as a linear function of \( X_t \), weather variability measured by annual observations of ROS (in millimeters) in winter (ROS_w), and residual stochastic variation in \( r \) (process error, e_r).

\[ r_t = \beta_0 + \beta_X \cdot X_t + \beta_{\text{ROS}} \cdot \text{ROS}_w + e_t, \]  

where we assume that the process error \( e_t \sim N(0, \sigma_e^2) \) and \( \beta_0, \beta_X, \beta_{\text{ROS}} \) and \( \sigma_e^2 \) are parameters estimated by the data. Measurement error was included in the model assuming a log normal distribution for the densities estimated using the SECR model, \( D_t \), giving:

\[ D_t \sim \log\text{normal}(\log(X_t), \sigma_{\text{error}}^2). \]  

The log normal measurement error SDs, \( \sigma_{\text{error}} \), were calculated from the estimates of the SE of \( D_t \) (SE(D_t)) obtained in the SECR analysis:

\[ \sigma_{\text{error}} = \log\left((\text{SE}(D_t)/D_t)^2 + 1\right). \]  

The model (Eqs. 1–3) was fitted in JAGS version 4.2.0 (65). Point estimates of \( r_t \) and associated 95% credibility intervals presented in the figures were obtained by fitting a model for \( r_t \) with time fitted as a factor (i.e., \( r_t = \beta_t \)). In addition to parameter estimates and associated 95% credibility intervals, we report estimates of Bayesian MSE \( R^2 \) for the models (66). The Bayesian \( R^2 \) was calculated as the mean of \( R^2 = \text{var(estimated values)}/\text{var(observations)} \), where \( i \) is the index of draws in Markov Chains Monte-Carlo sampling, \( \text{var(estimated values)} = \text{var(fit}_i)/\text{var(estimated values)} \), and \( \text{var(observations)} = \sigma^2_{\text{error}} \).

Analyses of population growth were done separately for the ~2.5-mo summer period (June/July to September) and for the ~9.5-mo winter period (September to June/July) as we expected the population dynamics to differ substantially in these two seasons. In the winter, we expected the amount of ROS to affect population growth (51, 67).

Summer population growth could be estimated from the change in densities from June/July (t) to August (t + 1) and from August (t + 1) to September (t + 2). Differences between these periods were investigated by fitting period as a factor in the model for population growth (Eq. 2). Population growth rates were not estimated for the summer of 1996, when there were no voles captured in the study area, and 2002, when there were no voles captured in the first and second primary trapping periods and an estimate of one vole per hectare (three voles caught) in the third primary period. The very low density estimates in September 2002 and the absence of voles in 1996 implied adoption of methodological adjustments that are detailed in the SI Appendix to allow growth rate estimates over the associated winters.

The timing of the primary trapping periods differed somewhat in 1990 to 1991 from subsequent years. In 1990, all the trapping was early in the season and the time period from primary period 1 to primary period 3 was only 0.7 mo. We therefore only used data from primary period 1 and 3 to estimate...
population growth to get a time period that was more similar to the other years ($\Delta t = 1.2$ to $1.7$ mo). In 1991, it was only 2 wk between primary period 1 and 2 and we used only estimates from primary period 2 and 3 in analyses.

**Model Simulations of Multiannual Population Dynamics.** We simulated the annual population dynamics linking summer and winter population growth using

$$X_{s,t} = X_{s} * e^{\beta_{s} \Delta t}$$

and

$$X_{w,t+1} = X_{w} * e^{\beta_{w} \Delta t}$$

where $X_{s}$ and $X_{w}$ is population density in the spring and autumn in year ($t$), respectively, $\beta_{s}$ and $\beta_{w}$ are population growth rate (month$^{-1}$) in summer and winter, respectively, and $\Delta t$ and $\Delta t_{w}$ are the time periods of the summer and winter seasons, respectively ($\Delta t_{s} + \Delta t_{w} = 12$).

Using Eq. 2, $r_{w}$ were modeled with parameters estimated from the data (Table 1). Our baseline deterministic model included only density dependence ($\bar{r}_{w} = \beta_{s} + \beta_{w} \times X_{s} j$) and assumed 3 mo of summer and 9 mo of winter. The sensitivity of population dynamics to changes in parameter values were evaluated using bifurcation diagrams and analyses of autocorrelation.

**Data Availability.** Capture-recapture data of the voles (all data used in the manuscript) have been deposited in Dryad data repository and are accessible with the following URL: https://doi.org/10.5061/dryad.hhmgknkhd (68).

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